On some symmetric multidimensional continued fraction algorithms

Pierre Arnoux [*]   Sébastien Labbé [†]

Abstract

We compute explicitly the density of the invariant measure for the Reverse algorithm which is absolutely continuous with respect to Lebesgue measure, using a method proposed by Arnoux and Nogueira. We also apply the same method on the unsorted version of Brun algorithm and Cassaigne algorithm. We illustrate some experimentations on the domain of the natural extension of those algorithms. For some other algorithms, which are known to have a unique invariant measure absolutely continuous with respect to Lebesgue measure, the invariant domain found by this method seems to have a fractal boundary, and it is unclear that it is of positive measure.

Keywords: Multidimensional continued fractions algorithms. Invariant measure. Natural extension. Fractal.

2010 Mathematics Subject Classification: 37A45, 37C40.

1   Introduction

Many continued fraction algorithms, in one or several dimensions, have a unique ergodic measure which is absolutely continuous with respect to Lebesgue measure (the so-called Gauss measure); this measure plays an important role in the dynamic of the algorithm, and it is interesting, when possible, to give an explicit closed formula for the invariant density.

It is easy to check if a given function δ is an invariant density for a map F, since it must be solution of the functional equation $\delta(x) = \sum_{y;F(y)=x} \frac{\delta(y)}{J_F(y)}$, where J_F is the jacobian of F; but it is generally difficult to find an explicit solution to this equation. One method, as proposed by Arnoux and Nogueira [AN93], is to build a geometric model of the natural extension of F, since it is often easier to find invariant densities for invertible maps. Another advantage of this method is that it gives a suspension flow on a manifold which can sometimes be linked to other areas of mathematics; for example, for Farey continued fraction, this flow can be recognized as the geodesic flow on the modular surface; for Rauzy induction on interval exchange maps, it is a Teichmüller flow.

The basic idea is as follows. A continued fraction algorithm can often be presented as a piecewise-linear map F on the positive cone of $\mathbb{R}^d$ (or a subcone of it), given locally as $\mathbf{x} \mapsto M^{-1}.\mathbf{x}$, where M is a positive matrix. We can try to build a model $\widetilde{F}$ for the natural extension as a skew product on a cone of $\mathbb{R}^d \times \mathbb{R}^d$, $(\mathbf{x}, \mathbf{a}) \mapsto (M^{-1}.\mathbf{x}, M^\top.\mathbf{a})$. We can look for a cone on which this map

---

is one-to-one, up to a set of measure 0; contraction arguments (A generalization of Hutchinson's theorem, see [AS15]) ensure, in most interesting cases, that such a set is unique, and must exist (but it could be of measure 0).

If such a set D can be found, then $\widetilde{F} : D \to D$ is by construction a bijection which preserves Lebesgue measure and the scalar product $\mathbf{x}.\mathbf{a}$. Furthermore, the flow φ_t defined on D by $\varphi_t(\mathbf{x}, \mathbf{a}) = (e^t\mathbf{x}, e^{-t}\mathbf{a})$ also preserves the measure and the scalar product, and commutes with $\widetilde{F}$. Let D_1 be the subset of D defined by $\mathbf{x}.\mathbf{a} = 1$, and Ω be the quotient $D_1/\widetilde{F}$; the flow ϕ_t projects to the quotient, and we can construct a natural extension of the projective map f associated with F as first return map of the flow to a section.

In this paper, we use this method to find the invariant density for two multidimensional continued fraction algorithms, the Reverse algorithm and the Brun algorithm, and we explain the problems encountered to apply this method to some other algorithms. Finding the set D is the crucial part of the method. A set D is known for Brun algorithm and is not known for the Hurwitz nor the Jacobi-Perron algorithm. Experimentations seem to show that it has a fractal structure for Jacobi-Perron and Arnoux-Rauzy-Poincaré algorithms.

In Section 2, as an example, we show how to compute the invariant density for the unsorted Farey map. In Section 3, we give the general argument. In Section 4, we apply it to the Reverse algorithm, in Section 5 to the Cassaigne algorithm and in Section 6, to the Brun algorithm. In Section 7, we explain how numerical computations can show in some cases the invariant set, but give more complicated results (fractal sets) in other cases. In Section 8, we make some remarks about sorted and unsorted algorithms, acceleration and choices of coodinates, to explain why one can find many variants of the continued fraction algorithms, and we illustrate that on the classical continued fraction. We also give the explicit computation of the invariant density for Brun's algorithm in any dimension.

2 The unsorted Farey map

Consider the positive cone $\Lambda = \{(x,y) \in \mathbb{R}^2 \mid 0 < x, 0 < y\}$ and define

$$F : (x,y) \mapsto \begin{cases} (x, y - x) & \text{if } x < y, \\ (x - y, y) & \text{if } x > y. \end{cases}$$

Remark 1. *The map F is well-defined, except on a set of Lebesgue measure 0. There is no canonical way to define it on the diagonal $x = y$; we could choose a convention, but since we will be interested in the ergodic properties of F, this is irrelevant. This will occur in all the examples of the paper, and we will no more remark about it.*

Since the map F is homogeneous of degree 1 (it commutes with positive homotheties), there is an associated projective map f, which can be defined on the unit simplex. This projective version of F on $\Delta = \{(x,y) \in \Lambda \mid x + y = 1\}$ is defined as $f : \mathbf{x} \mapsto \frac{F(\mathbf{x})}{\|F(\mathbf{x})\|_1}$. If we take the first variable x as coordinate on the unit simplex, f can be defined as (see Figure 1)

$$\begin{aligned} f : \quad [0,1] \quad &\to \quad [0,1] \\ x \quad &\mapsto \quad \begin{cases} \frac{x}{1-x} & \text{if } x < \frac{1}{2}, \\ 2 - \frac{1}{x} & \text{if } x > \frac{1}{2}. \end{cases} \end{aligned}$$
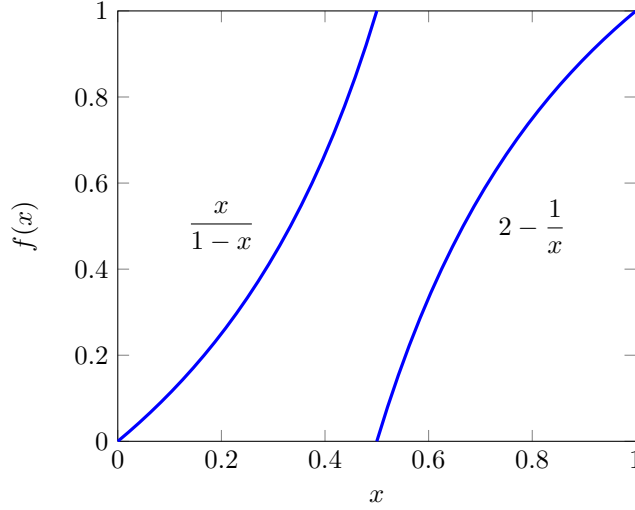
Figure 1: The graph of the unsorted Farey map.

Remark 2. *Except for a set of measure 0 (the points (x, y) such that $\frac{x}{y}$ is rational), the positive orbit $\{F^n(x, y)\}$ is defined for all $n \in \mathbb{N}$, and the norm $|x+y|$ is strictly decreasing along the orbit; the same proof shows that the positive orbit $\{f^n(x)\}$ is well defined for all $n \in \mathbb{N}$ if and only if x is irrational.*

Below we compute an invariant measure of f, using a method that also works for some multi-dimensional continued fractions algorithms. Note that neither F nor f are bijection. For instance, we have $F(2, 5) = F(5, 3) = (2, 3)$:
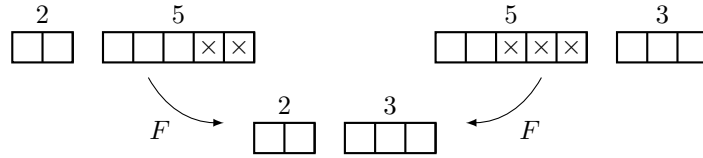
One way to transform F into a bijection is to stack the removed part and pile up the result in the following way, as done in [AN93]:
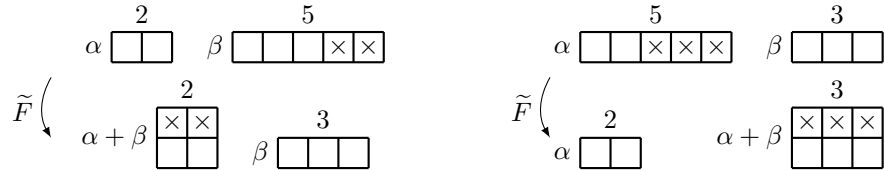
We thus constructed the following function

$$\widetilde{F} : (x, y, \alpha, \beta) \mapsto \begin{cases} (x, y - x, \alpha + \beta, \beta) & \text{if } x < y, \\ (x - y, y, \alpha, \alpha + \beta) & \text{if } x > y. \end{cases}$$

which is an explicit model of the *natural extension* of F. It is a bijection on the domain $D = \{(x, y, \alpha, \beta) \in \mathbb{R}^2 \times \mathbb{R}^2 \mid x > 0, y > 0, \alpha > 0, \beta > 0\}$; indeed, if we define $\Lambda_1 = \{(x, y) \in \mathbb{R}^2 \mid 0 <$

3

$x < y\}$ and $\Lambda_2 = \{(x,y) \in \mathbb{R}^2 \mid 0 < y < x\}$, we have $\widetilde{F}(\Lambda_1 \times \Lambda) = \Lambda \times \Lambda_2$ and $\widetilde{F}(\Lambda_2 \times \Lambda) = \Lambda \times \Lambda_1$; Since Λ_1, Λ_2 is a measurable partition of Λ, $\widetilde{F}$ sends a partition of D to another.

Both branches of $\widetilde{F}$ can be written linearly as the action by a 4×4 matrix of the form

$$\begin{pmatrix} M^{-1} & 0 \\ 0 & M^{\top} \end{pmatrix} \quad \text{where} \quad M = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \quad \text{if } x < y \quad \text{and} \quad M = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \quad \text{if } x > y.$$

Hence $\widetilde{F}$ preserves Lebesgue measure, since its jacobian is 1. It also preserves the form $x\alpha + y\beta$, that is, the total area of the rectangles.

We want to find a natural extension and an invariant measure for the projective map. For this, we introduce the flow φ_t:

$$\varphi_t : (x, y, \alpha, \beta) \mapsto (xe^t, ye^t, \alpha e^{-t}, \beta e^{-t}).$$

This flow has jacobian 1, thus preserves Lebesgue measure, and also preserves the form $x\alpha + y\beta$. It commutes with $\widetilde{F}$, that is, $\varphi_t(\widetilde{F}(\mathbf{x})) = \widetilde{F}(\varphi_t(\mathbf{x}))$.

Define now a subset D_1 of codimension 1 of D by $D_1 = \{(x, y, \alpha, \beta) \in D \mid x\alpha + y\beta = 1\}$, and a surface $\Sigma \in D_1$ by $\Sigma = \{(x, y, \alpha, \beta) \in D_1 \mid x + y = 1\}$.

D_1 is invariant by $\widetilde{F}$ and φ_t; $\widetilde{F}$ acts without periodic points since it decreases strictly $x + y$. The surface Σ is a section for ϕ_t, since, for any $(x, y, \alpha, \beta) \in D_1$, there exists a unique $\tau = -\log(x + y)$ such that $\varphi_\tau(x, y, \alpha, \beta) \in \Sigma$.

Since φ_t commutes with $\widetilde{F}$, we can consider the induced flow on $D_1/\widetilde{F}$ and the first return map of the flow to the section Σ. By construction, the effect of this map on the first coordinates is the effect of the map F followed by a renormalization, that is, the projective map f.

We consider the change of coordinates $(x, y, \alpha, \beta) \mapsto (x, \alpha - \beta, -\log(x + y), x\alpha + y\beta) = (x, \alpha', \tau, e)$; computation shows that the jacobian is 1, so Lebesgue measure is given in these coordinates as $dx \, d\alpha' \, d\tau \, de$. The domain D_1 is given by $e = 1$, and the surface of section Σ by $\tau = 0$, hence the first return map $\widetilde{f}$ must leave invariant the measure $dx \, d\alpha'$. Indeed, a straightforward computation shows that $\widetilde{f}$ is given by

$$\widetilde{f} : \quad \Sigma \quad \to \quad \Sigma$$
$$(x, \alpha') \quad \mapsto \quad \begin{cases} \left(\frac{x}{1-x}, 1 - x + \alpha'(1-x)^2\right) & \text{if } x < \frac{1}{2}, \\ \left(2 - \frac{1}{x}, \alpha' x^2 - x\right) & \text{if } x > \frac{1}{2}. \end{cases}$$

and this map has jacobian 1.

We can describe more precisely the surface Σ by (see Figure 2):

$$\Sigma = \{(x, \alpha', \tau, e) \mid \tau = 0, e = 1, \alpha' + \beta > 0, \beta > 0, x\alpha' + \beta = 1\}$$
$$= \{(x, \alpha', 0, 1) \mid \alpha' + 1 - x\alpha' > 0, 1 - x\alpha' > 0\}$$
$$= \{(x, \alpha', 0, 1) \mid \frac{1}{x-1} < \alpha' < \frac{1}{x}\}$$

Therefore the invariant density of f is

$$\delta(x) = \int_{\frac{1}{x-1}}^{\frac{1}{x}} 1 \, d\alpha' = \frac{1}{x} - \frac{1}{x-1} = \frac{1}{x(1-x)}.$$

One easily check that this density is invariant; a straightforward computation shows that, if $y_1 = \frac{x}{1+x}$ and $y_2 = \frac{1}{2-x}$ are the preimages of the point x, we have $\delta(x) = \frac{\delta(y_1)}{f'(y_1)} + \frac{\delta(y_2)}{f'(y_2)}$.
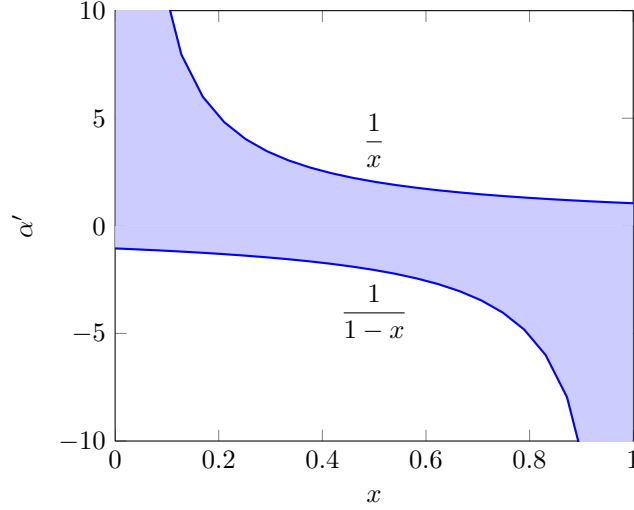
4

Figure 2: The domain of the natural extension of the unsorted Farey map.

Remark 3. *That density is unbounded; moreover, the total mass of the measure is infinite. The reason is that we have two indifferent fixed points at both extremities of the interval, because we have used the additive algorithm. We could accelerate the map around the fixed points by using the multiplicative algorithm, and recover the classical finite Gauss measure (more exactly, a variant of it, since we use the unsorted map).*

Remark 4. *There is one more structure: the map $\widetilde{F}$ and the flow φ_t leave invariant the symplectic form $dx \wedge d\alpha + dy \wedge d\beta$, and the flow φ_t is the hamiltonian flow associated with the hamiltonian $x\alpha + y\beta$. In fact, the set $D/\widetilde{F}$ can be identified to the tangent bundle of the modular surface, given in coordinates, and φ_t is the geodesic flow.*

3 Constructing the measure from the natural extension

One can extend the method of the previous section to a large range of continued fraction algorithm. We recall the method proposed by Arnoux and Nogueira [AN93] to find heuristics for a geometric model of the natural extension of a multidimensional continued fraction algorithm, and an explicit formula for the invariant density (Gauss measure) in some cases.

Let Λ be a subcone of the positive cone $\mathbb{R}^d_+$ (Two interesting cases are when $\Lambda = \mathbb{R}^d_+$ and when $\Lambda = \{(x_1, x_2, \ldots, x_d) \in \mathbb{R}^d \mid 0 < x_1 < x_2 < \ldots < x_d\}$ consists of sorted entries). A *Multidimensional Continued Fraction (MCF) algorithm* is a function

$$F: \begin{array}{ccc} \Lambda & \to & \Lambda \\ \mathbf{x} & \mapsto & M(\mathbf{x})^{-1} \cdot \mathbf{x}. \end{array}$$

where M is an homogeneous function of degree 0, piecewise constant on subcones, that associates to each $\mathbf{x} \in \Lambda$ an invertible matrix $M(\mathbf{x})$. Classical references for MCF algorithms are [Sch00, Bre81]. In the cases we consider, the entries of $M(\mathbf{x})$ are nonnegative integers.

To the function F, we can associate its projective version f, defined in a canonical way from the piecewise linear map on the projective space $P(\Lambda)$. For explicit computation, one should take

coordinates, and a nice way to do that is to consider the codimension 1 compact domain in Λ, $\Delta = \{\mathbf{x} \in \Lambda \mid \|\mathbf{x}\| = 1\}$ for some norm $\|\cdot\|$; it is explicitly given by:

$$
\begin{array}{rccc}
f: & \Delta & \to & \Delta \\
& \mathbf{x} & \mapsto & \frac{F(\mathbf{x})}{\|F(\mathbf{x})\|}.
\end{array}
$$

Even if there is no canonical coordinates on $P(\Lambda)$, and hence no canonical Lebesgue measure, there is a natural Lebesgue measure class, since all changes of coordinates are piecewise projective. In many cases, there is a unique invariant (up to a constant) measure in this class, the Gauss measure, given by an invariant density. We are interested by explicit formulas for this invariant density, and for this, we try to build a geometric model for the natural extension of f with good properties.

A potential geometric model for the *natural extension* $\widetilde{F}$ of the piecewise linear function F can be defined as:

$$
\begin{array}{rccc}
\widetilde{F}: & \Lambda \times \mathbb{R}_+^d & \to & \Lambda \times \mathbb{R}_+^d \\
& \begin{pmatrix} \mathbf{x} \\ \mathbf{a} \end{pmatrix} & \mapsto & \begin{pmatrix} M(\mathbf{x})^{-1} & 0 \\ 0 & M(\mathbf{x})^\top \end{pmatrix} \begin{pmatrix} \mathbf{x} \\ \mathbf{a} \end{pmatrix}
\end{array}
$$

To normalize the vector $\mathbf{x}$ after the application of $\widetilde{F}$, we define the flow φ_t as

$$
\begin{array}{rccc}
\varphi_t: & \Lambda \times \mathbb{R}_+^d & \to & \Lambda \times \mathbb{R}_+^d \\
& \begin{pmatrix} \mathbf{x} \\ \mathbf{a} \end{pmatrix} & \mapsto & \begin{pmatrix} e^t & 0 \\ 0 & e^{-t} \end{pmatrix} \begin{pmatrix} \mathbf{x} \\ \mathbf{a} \end{pmatrix}
\end{array}
$$

for all $t \in \mathbb{R}$. Note that the following properties are verified:

- φ_t and $\widetilde{F}$ are well defined, since the matrix $M(x)$ is nonnegative and the definition set is a product of cones;

- φ_t and $\widetilde{F}$ both have jacobian 1;

- φ_t and $\widetilde{F}$ both preserve the scalar product $\langle \mathbf{x}, \mathbf{a} \rangle$;

- φ_t commutes with $\widetilde{F}$: $\varphi_t \circ \widetilde{F} = \widetilde{F} \circ \varphi_t$.

Remark 5. *The map $\widetilde{F}$ has no reason to be a bijection, and the main problem is to find a conical domain $D \subset \Lambda \times \mathbb{R}_+^d$ such that $\widetilde{F}$ is a bijection on D.*

From now on, we suppose that we have found such a domain D of positive measure. Then φ_t and $\widetilde{F}$ preserve Lebesgue measure on D, since they are bijections with jacobian 1. But this measure is not very useful, because not only it has infinite mass, but its projection to Λ has infinite density. We want to restrict the domain.

Definition 6. *We define $D_1 = \{(\mathbf{x}, \mathbf{a}) \in D \mid \langle \mathbf{x}, \mathbf{a} \rangle = 1\}$*

We can define a natural Lebesgue measure on D_1 by considering coordinates on D_1 complemented by $\langle \mathbf{x}, \mathbf{a} \rangle$ to get a global coordinate system on D, and disintegrating Lebesgue measure. The functions $\widetilde{F}$ and φ_t are well-defined on D_1 and both preserve locally Lebesgue measure on D_1. Since they are bijective, they also preserve it globally.

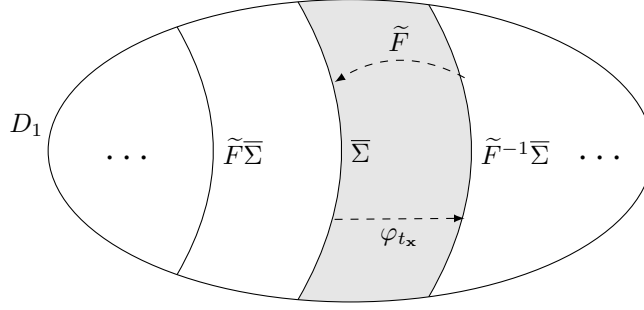**Definition 7.** *We define $\Omega = D_1 / \widetilde{F}$, the set of orbits of the map $\widetilde{F}$.*

Figure 3: The function $\widetilde{F}$ applied iteratively on the section $\overline{\Sigma} \subset D_1$. The gray region corresponds to a fundamental domain of the action of $\widetilde{F}$ on D_1 and can be identified to Ω.

Since φ_t and $\widetilde{F}$ commute, we can make φ_t act on the quotient space Ω. We can assimilate Ω to a fundamental domain of the action of $\widetilde{F}$ on D_1 (this is well defined, since, except for the fixed points of F which have measure 0, $\widetilde{F}$ decreases the norm of $\mathbf{x}$).

Definition 8. *We define* $\overline{\Sigma} = \{(\mathbf{x}, \mathbf{a}) \in D_1 \mid \|\mathbf{x}\| = 1\}$*, and denote by* Σ *its projection to* Ω*.*

Remark 9. *The surface* $\overline{\Sigma}$ *is a section for the flow* φ_t*: for any* $(\mathbf{x}, \mathbf{a}) \in D_1$*, there is a unique* $t = -\log(\|\mathbf{x}\|)$*, depending only on* $\mathbf{x}$*, such that* $\varphi_t(\mathbf{x}, \mathbf{a}) \in \overline{\Sigma}$*.*

For any $(\mathbf{x}, \mathbf{a}) \in \Sigma$, define $t_{\mathbf{x}}$ to be the first return time of $(\mathbf{x}, \mathbf{a})$ to Σ, that is, $t_{\mathbf{x}} = -\log \|F(\mathbf{x})\|$. We can define a fundamental domain for the action of $\widetilde{F}$ as

$$\{\varphi_t(\mathbf{x}, \mathbf{a}) \mid (\mathbf{x}, \mathbf{a}) \in \overline{\Sigma}, 0 \leq t < t_{\mathbf{x}}\}$$

We consider the first return map $\widetilde{f} : \Sigma \to \Sigma$ of the flow on Ω. This map can be seen in the fundamental domain as the composition of $\widetilde{F}$ with the flow: $\widetilde{f} = \varphi_{t_{\mathbf{x}}} \circ \widetilde{F}(\mathbf{x}, \mathbf{a})$.

Since $\widetilde{F}$ preserves Lebesgue measure on D_1, there is a well-defined Lebesgue measure on Ω, which is preserved by ϕ_t; since Σ is a section, we can define a transverse invariant measure μ on Σ in the usual way: if m is Lebesgue mesure on Ω and $U \subset \Sigma$ is a measurable set, we define

$$\mu(U) = \lim_{t \to 0} \frac{m\{\phi_s(u) \mid u \in U, 0 \leq s \leq t\}}{t}$$

From the definition, we have $\widetilde{f}(\mathbf{x}, \mathbf{a}) = (f(\mathbf{x}), M(\mathbf{x})^\top \mathbf{a} \cdot \|F(\mathbf{x})\|)$, and $\widetilde{f}$ can be projected on the application f. Hence we can find an invariant measure for f equivalent to Lebesgue measure by partial integration on the section Σ.

Proposition 10. [AN93] *If the measure of* Σ *is positive, by choosing a coordinate system* $(\mathbf{x}, \mathbf{a})$ *in which the invariant measure is written* $d\mathbf{x} \, d\mathbf{a}$*, the invariant measure of* f *can be computed by the formula:*

$$\delta(\mathbf{x}) = \int_{\{\mathbf{a} : (\mathbf{x}, \mathbf{a}) \in \Sigma\}} 1 \, d\mathbf{a}$$

We now describe an explicit change of variables that is useful to compute explicitly all these measures in the case of the norm $\|\mathbf{x}\| = \sum_{i=1}^d |x_i|$. Define $y_i = x_i$ for $i < d$, and $\tau = -\log(\sum_{i=1}^d |x_i|)$; define $b_i = a_i - a_d$ for $i < d$ and $e = \sum_{i=1}^d a_i x_i$. A straightforward computation shows that the change of coordinates $(\mathbf{x}, \mathbf{a}) \mapsto (\mathbf{y}, \mathbf{b}, \tau, e)$ has jacobian 1. The domain D_1 is defined by $e = 1$, so the invariant measure on Ω is given by $\prod_{i=1}^{d-1} dy_i \prod_{i=1}^{d-1} db_i \, d\tau$; the section Σ is defined by $\tau = 1$; since the coordinate τ is, up to a sign, the time coordinate of the flow, the invariant transverse measure for the flow, that is, the invariant measure for $\widetilde{f}$, is given explicitly by $\prod_{i=1}^{d-1} dy_i \prod_{i=1}^{d-1} db_i$.

7

Remark 11. *We have made arbitrary choices. The first one was the choice of a coordinates for the projective space $P(\Lambda)$; we could choose another section, for example $x_d = 1$ (we could also take another norm, but it seems in general more convenient to choose an affine section). The second one, which is related, is the choice of the section Σ. The flow φ_t and the space Ω are more intrinsic; different choices lead to different forms of the same continued fraction; this explains why we can find a large range of formulas for apparently similar continued fractions, based on a different choice of coordinates.*

4   The Reverse algorithm

The Arnoux-Rauzy algorithm is a partial algorithm, defined on the positive cone by subtracting the two smallest coordinates from the largest one. It is only defined if the largest coordinate is larger than the sum of the two smaller coordinates, that is, on the complement of the set of coordinates which satisfy the triangular inequality. The set of points with infinite orbits, known as the Rauzy gasket [AS13], has measure 0, which make it unusable to find rational approximations of most points in the positive cone.

It can be completed in various ways by defining it on this missing set; for example, on can consider the simplest map which sends the central cone onto the positive cone, $(x, y, z) \mapsto (-x + y + z, x - y + z, x + y - z)$.

More formally, we can use the notations of the previous section, and define Reverse algorithm on the following partition of $\Lambda = \{(x, y, z) \in \mathbb{R}^3 \mid x > 0, y > 0, z > 0\}$ up to a set of measure zero:

$$\Lambda_1 = \{(x, y, z) \in \Lambda \mid x > y + z\},$$
$$\Lambda_2 = \{(x, y, z) \in \Lambda \mid y > x + z\},$$
$$\Lambda_3 = \{(x, y, z) \in \Lambda \mid z > x + y\},$$
$$\Lambda_4 = \{(x, y, z) \in \Lambda \mid x < y + z, y < x + z, z < x + y\}$$

The name *Reverse* comes from the fact that its effect, as a projective map acting on the central triangle Λ_4 of the unit simplex, is an homothety of ratio 2, followed by a central symmetry which reverses the vector with respect to the center of the simplex.

We define the four matrices:

$$A_1 = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, A_2 = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}, A_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix}, A_4 = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix}.$$

and the matrix function $M : \Lambda \to GL(3, \mathbb{Q})$ such that $M(\mathbf{x}) = A_i$ if and only $\mathbf{x} \in \Lambda_i$. Recall that this matrix function defines the functions $F(\mathbf{x}) = M(\mathbf{x})^{-1}\mathbf{x}$, $f(\mathbf{x}) = F(\mathbf{x})/\|F(\mathbf{x})\|_1$ and $\widetilde{F}(\mathbf{x}, \mathbf{a}) = (M(\mathbf{x})^{-1}\mathbf{x}, M(\mathbf{x})^\top \mathbf{a})$.

We show in the left of figure 4 the trace on the unit simplex of the partition of Λ; each branch of the map F sends Λ_i to all of Λ.

Explicitly, the function $\widetilde{F}$ is given by

$$\widetilde{F} : (x, y, z, \alpha, \beta, \gamma) \mapsto \begin{cases} (x - y - z, y, z, \alpha, \alpha + \beta, \alpha + \gamma) & \text{if } (x, y, z) \in \Lambda_1, \\ (x, y - x - z, z, \alpha + \beta, \beta, \beta + \gamma) & \text{if } (x, y, z) \in \Lambda_2, \\ (x, y, z - x - y, \alpha + \gamma, \beta + \gamma, \gamma) & \text{if } (x, y, z) \in \Lambda_3, \\ (-x + y + z, x - y + z, x + y - z, \frac{\beta + \gamma}{2}, \frac{\alpha + \gamma}{2}, \frac{\alpha + \beta}{2}) & \text{if } (x, y, z) \in \Lambda_4. \end{cases}$$
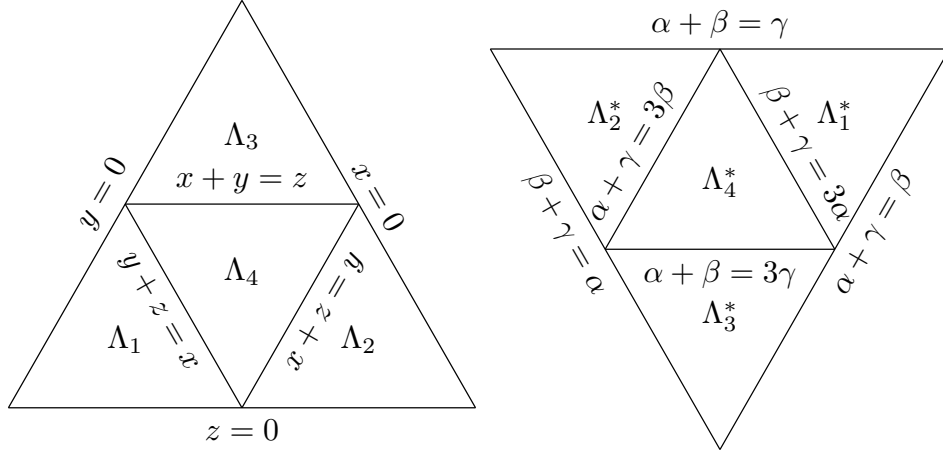
Figure 4: Left: the trace on the unit simplex of the partition of Λ into Λ_i for $i \in \{1, 2, 3, 4\}$. Right: the partition of Λ^* into Λ_i^* for $i \in \{1, 2, 3, 4\}$.

One can easily find a cone Λ^* such that $\widetilde{F}$ is a bijection on $D = \Lambda \times \Lambda^*$: numerical experimentations show that after few iterations of $\widetilde{F}$, (α, β, γ) belongs to the subset of $\Lambda_4 \subset \Lambda$ of triples which satisfy the triangular inequality. Indeed, for any $(\alpha, \beta, \gamma) \in \Lambda$, $(\frac{\beta+\gamma}{2}, \frac{\alpha+\gamma}{2}, \frac{\alpha+\beta}{2})$ satisfies the triangular inequality; and if $(\alpha, \beta, \gamma) \in \Lambda_4$, then $(\alpha, \alpha + \beta, \alpha + \gamma)$ also satisfies the triangular inequality. Since almost all orbits of F cross Λ_4, with probability 1, after a finite time, the second part of the coordinates of $\widetilde{F}^n(\mathbf{x}, \mathbf{a})$ enters Λ_4 and never escapes.

Hence we define $\Lambda^* = \Lambda_4$, and a partition of Λ^* by

$$\Lambda_1^* = \{(\alpha, \beta, \gamma) \in \Lambda^* \mid \alpha < \frac{\alpha + \beta + \gamma}{4}\},$$

$$\Lambda_2^* = \{(\alpha, \beta, \gamma) \in \Lambda^* \mid \beta < \frac{\alpha + \beta + \gamma}{4}\},$$

$$\Lambda_3^* = \{(\alpha, \beta, \gamma) \in \Lambda^* \mid \gamma < \frac{\alpha + \beta + \gamma}{4}\},$$

$$\Lambda_4^* = \{(\alpha, \beta, \gamma) \in \Lambda^* \mid \alpha, \beta, \gamma > \frac{\alpha + \beta + \gamma}{4}\}.$$

Lemma 12. $\{\Lambda_1^*, \Lambda_2^*, \Lambda_3^*, \Lambda_4^*\}$ *is a partition of Λ^*, up to a set of measure 0.*

Proof. It is clear from the definition that Λ_4^* is disjoint from the other 3, since it is defined by opposite inequalities. Consider a point in the intersection of Λ_1^* and Λ_2^*; we have $\alpha, \beta < \frac{\alpha+\beta+\gamma}{4}$, hence $\gamma > \frac{\alpha+\beta+\gamma}{2}$, which is incompatible with the triangular inequality. The same proof is valid for the two other intersections, hence these four sets are disjoint.

The equality $\alpha = \frac{\alpha+\beta+\gamma}{4}$ defines a set of codimension 1 and measure 0, and similarly for the two other ones. For any other point in Λ^*, either all of α, β, γ are larger than $\frac{\alpha+\beta+\gamma}{4}$, which defines λ_4^*, or one (and only one, by the previous paragraph) is smaller, which defines one of the other simplices; hence they form a partition, up to a set of measure 0. We show in the right of figure 4 the trace on the unit simplex of this partition. $\square$

As an immediate consequence, we obtain a partition of D:

9

Corollary 13. *Up to a set of measure 0, we can write the set $D = \Lambda \times \Lambda^*$ as a disjoint union*

$$D = \coprod_{i=1}^{4} \Lambda_i \times \Lambda^* = \coprod_{i=1}^{4} \Lambda \times \Lambda_i^*$$

Hence, to prove that $\widetilde{F}$ is a bijection on D, since each branch of $\widetilde{F}$ is a non-degenerate linear map, it suffices to prove:

Lemma 14. *We have $\widetilde{F}(\Lambda_i \times \Lambda^*) = \Lambda \times \Lambda_i^*$.*

Proof. We check directly by computation that $F(\Lambda_i) = \Lambda$. We want to show that if $\mathbf{x} \in \Lambda_i$, then $A_i^\top \mathbf{a} \in \Lambda_i^*$. We use the notation $(x', y', z', \alpha', \beta', \gamma') = \widetilde{F}(x, y, z, \alpha, \beta, \gamma)$ and we proceed case by case.

If $\mathbf{x} \in \Lambda_1$, then $\alpha' + \beta' + \gamma' = 3\alpha + \beta + \gamma > 4\alpha = 4\alpha'$. Therefore $\widetilde{F}(x, y, z, \alpha, \beta, \gamma) \in \Lambda \times \Lambda_1^*$. A similar proof applies for Λ_2 and Λ_3.

If $\mathbf{x} \in \Lambda_4$, then $\alpha' + \beta' + \gamma' = \alpha + \beta + \gamma < 2\alpha + 2\beta = 4\gamma'$; a similar proof applies to α' and β', which proves that $\widetilde{F}(x, y, z, \alpha, \beta, \gamma) \in \Lambda \times \Lambda_4^*$.

We have proved inclusion; to prove equality, it is enough to show that the matrix $A_i^\top$ sends the extreme points of Λ^* to the extreme points of Λ_i^*, which is done by computation. $\square$
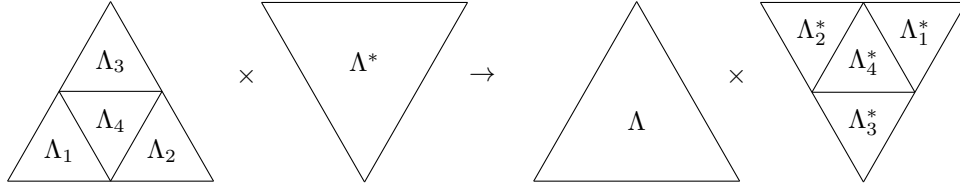

Figure 5: The function $\widetilde{F} : \Lambda_i \times \Lambda^* \to \Lambda \times \Lambda_i^*$ is one-to-one and onto for each $i \in \{1, 2, 3, 4\}$.

We now apply the results of the previous section. Recall that

$$D_1 = \{(x, y, z, \alpha, \beta, \gamma) \in D \mid x\alpha + y\beta + z\gamma = 1\}.$$

and the surface of section Σ is defined by

$$\Sigma = \{(x, y, z, \alpha, \beta, \gamma) \in \mathbb{R}_+^6 \mid x\alpha + y\beta + z\gamma = 1, x + y + z = 1, \alpha + \beta > \gamma, \alpha + \gamma > \beta, \beta + \gamma > \alpha\}.$$

It is convenient to take the coordinates on D_1 defined in the previous section: We keep variables x and y and change the coordinates $\alpha' = \alpha - \gamma$, $\beta' = \beta - \gamma$, $e = x\alpha + y\beta + z\gamma$, $\tau = -\log(x+y+z)$; it is readily checked that the jacobian is one. Since the domain D_1 is defined by $e = 0$, and coordinate τ is the return time of the flow to Σ, we get

$$\Sigma = \{(x, y, \alpha', \beta', \tau, e) \in \mathbb{R}_+^6 \mid \tau = 0, e = 1, \alpha' + \beta' + \gamma > 0, \alpha' + \gamma > \beta', \beta' + \gamma > \alpha'\}$$

Furthermore, the invariant measure for the return map $\widetilde{f}$ of the flow to Σ is $dx\,dy\,d\alpha'\,d\beta'$. From this, we obtain:

Proposition 15. *The density function of the invariant measure of $f : \Delta \to \Delta$ for the Reverse algorithm is*

$$\frac{1}{(1-x)(1-y)(1-z)}.$$

10

Proof. To obtain this density and according to Proposition 10, it suffices to integrate the measure $d\alpha'\, d\beta'$ for a fixed x, y. Given $(x, y) \in \mathbb{R}^2$, the set of admissible $(\alpha', \beta') \in \mathbb{R}^2$ satisfies

$$\begin{cases} x\alpha' + y\beta' + \gamma = 1, \\ \alpha' + \beta' + \gamma > 0, \\ \alpha' + \gamma > \beta', \\ \beta' + \gamma > \alpha', \end{cases} \quad \text{or equivalently} \quad \begin{cases} (1-x)\alpha' + (1-y)\beta' + 1 > 0, \\ (1-x)\alpha' + 1 > (1+y)\beta', \\ (1-y)\beta' + 1 > (1+x)\alpha'. \end{cases}$$

This domain is the interior of a triangle of vertices $\left(\frac{1}{x-1}, 0\right)$, $\left(0, \frac{1}{y-1}\right)$, $\left(\frac{1}{x+y}, \frac{1}{x+y}\right)$ shown below.
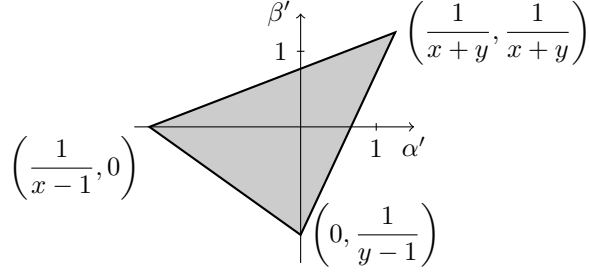
Lemma 16. *Let $a, b, c \in \mathbb{R}$. The area of a triangle of vertices $(a, 0)$, $(0, b)$ and (c, c) is $\frac{1}{2}|ab - bc - ac| = \frac{1}{2}|abc|\left|\frac{1}{c} - \frac{1}{a} - \frac{1}{b}\right|$.*

This lemma is proved by an easy determinant computation; it shows that the triangle we consider has area

$$\frac{1}{2}\left(\frac{1}{(x+y)(x-1)(y-1)}\right)|(x+y) - (x-1) - (y-1)| = \frac{1}{(x+y)(1-x)(1-y)}.$$

$\square$

Remark 17. *This density is not bounded: it tends to infinity at the three extreme points of the simplex. The reason is that these points are indifferent fixed points for the algorithm. This is a common feature for an additive continued fraction algorithm. However, and oppositely to what happens in dimension 1, the total mass of that density is bounded; its value is*

$$\int_0^1 \int_0^{1-x} \frac{1}{(x+y)(1-x)(1-y)} dy dx = \int_0^1 \frac{2\log(x)}{(x+1)(x-1)} dx = \frac{\pi^2}{4}$$

5 The Cassaigne algorithm

This algorithm was suggested by Julien Cassaigne (DynA3S meeting, LIAFA, Paris, October 12th, 2015)[1] as a way to generate words of low factor complexity $(p(n) = 2n+1)$ in with arbitrary letter frequencies on a three-letter alphabet. An interesting aspect of this algorithm is that there are always only two branches.

More formally, we can use the notations of the previous section, and define Cassaigne algorithm on the following partition of $\Lambda = \{(x_1, x_2, x_3) \in \mathbb{R}^3 \mid x_1 > 0, x_2 > 0, x_3 > 0\}$ up to a set of measure zero:

$$\Lambda_a = \{(x_1, x_2, x_3) \in \Lambda \mid x_1 > x_3\},$$
$$\Lambda_b = \{(x_1, x_2, x_3) \in \Lambda \mid x_3 > x_1\}.$$

---

[1] http://www.liafa.univ-paris-diderot.fr/dyna3s/Oct2015

We define two matrices:

$$C_a = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \quad \text{and} \quad C_b = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{pmatrix}.$$

We define the matrix function $M : \Lambda \to GL(3, \mathbb{Q})$ such that $M(\mathbf{x}) = A_i$ if and only $\mathbf{x} \in \Lambda_i$ for $i \in \{a, b\}$. Recall that this matrix function defines the functions $F(\mathbf{x}) = M(\mathbf{x})^{-1}\mathbf{x}$, $f(\mathbf{x}) = F(\mathbf{x})/\|F(\mathbf{x})\|_1$ and $\widetilde{F}(\mathbf{x}, \mathbf{a}) = (M(\mathbf{x})^{-1}\mathbf{x}, M(\mathbf{x})^\top \mathbf{a})$. We show in the left of Figure 6 the trace on the unit simplex of the partition of Λ; each branch of the map F sends Λ_i to all of Λ for $i \in \{a, b\}$.
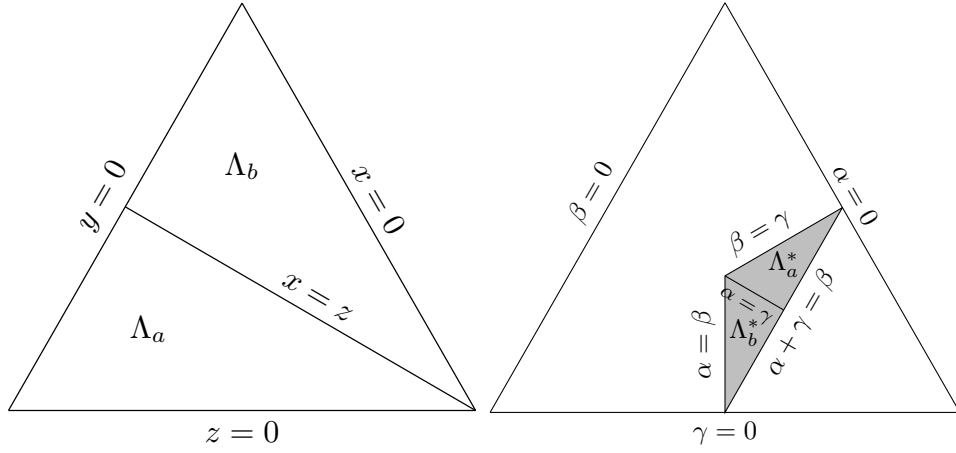

Figure 6: Left: the trace on the unit simplex of the partition of Λ into Λ_a and Λ_b. Right: the partition of Λ^* into Λ_a^* and Λ_b^*.

Explicitly, the function $\widetilde{F}$ is given by

$$\widetilde{F} : (x, y, z, \alpha, \beta, \gamma) \mapsto \begin{cases} (x - z, z, y, \alpha, \alpha + \gamma, \beta) & \text{if } (x, y, z) \in \Lambda_1, \\ (y, x, z - x, \beta, \alpha + \gamma, \gamma) & \text{if } (x, y, z) \in \Lambda_2. \end{cases}$$

One can easily find a cone Λ^* such that $\widetilde{F}$ is a bijection on $D = \Lambda \times \Lambda^*$: numerical experimentations show that after few iterations of $\widetilde{F}$, (α, β, γ) belongs to the subset of $\Lambda^* \subset \Lambda$

$$\Lambda^* = \{(\alpha, \beta, \gamma) \in \mathbb{R}^3 \mid \max\{\alpha, \gamma\} < \beta < \alpha + \gamma\}.$$

We define a partition (see the right of Figure 6) of Λ^* by

$$\Lambda_a^* = \{(\alpha, \beta, \gamma) \in \Lambda^* \mid \alpha < \gamma\},$$
$$\Lambda_b^* = \{(\alpha, \beta, \gamma) \in \Lambda^* \mid \alpha > \gamma\}.$$

Lemma 18. $\{\Lambda_a^*, \Lambda_b^*\}$ *is a partition of* Λ^*, *up to a set of measure 0.*

Proof. Trivial. ☐

As an immediate consequence, we obtain a partition of D:

Corollary 19. *Up to a set of measure 0, we can write the set $D = \Lambda \times \Lambda^*$ as a disjoint union*

$$D = \coprod_{i \in \{a,b\}} \Lambda_i \times \Lambda^* = \coprod_{i \in \{a,b\}} \Lambda \times \Lambda_i^*$$

Hence, to prove that $\widetilde{F}$ is a bijection on D, since each branch of $\widetilde{F}$ is a non-degenerate linear map, it suffices to prove:

Lemma 20. *We have $\widetilde{F}(\Lambda_a \times \Lambda^*) = \Lambda \times \Lambda_a^*$ and $\widetilde{F}(\Lambda_b \times \Lambda^*) = \Lambda \times \Lambda_b^*$.*

Proof. We check directly by computation that $F(\Lambda_i) = \Lambda$. We want to show that if $\mathbf{x} \in \Lambda_i$ and $\mathbf{a} \in \Lambda^*$, then $C_i^\top \mathbf{a} \in \Lambda_i^*$. It is enough to show that the matrix $C_i^\top$ sends the extreme points of Λ^* to the extreme points of Λ_i^* for $i \in \{a, b\}$. This is done by the following matrix computation

$$C_a^\top \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 1 & 1 & 1 \end{pmatrix} \quad \text{and} \quad C_b^\top \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 1 \\ 0 & 1 & 1 \end{pmatrix}$$
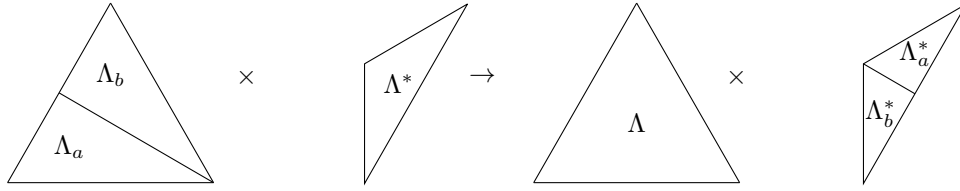
and this proves the equalities. $\qquad\square$

Figure 7: The function $\widetilde{F} : \Lambda_i \times \Lambda^* \to \Lambda \times \Lambda_i^*$ is one-to-one and onto for each $i \in \{1, 2, 3, 4\}$.

We now apply the results of the previous section. Recall that

$$D_1 = \{(x, y, z, \alpha, \beta, \gamma) \in D \mid x\alpha + y\beta + z\gamma = 1\}.$$

and the surface of section Σ is defined by

$$\Sigma = \{(x, y, z, \alpha, \beta, \gamma) \in \mathbb{R}_+^6 \mid x\alpha + y\beta + z\gamma = 1, x + y + z = 1, \alpha < \beta, \gamma < \beta, \beta < \alpha + \gamma\}.$$

It is convenient to take the coordinates on D_1 defined in the previous section: We keep variables x and y and change the coordinates $\alpha' = \alpha - \gamma$, $\beta' = \beta - \gamma$, $e = x\alpha + y\beta + z\gamma$, $\tau = -\log(x+y+z)$; it is readily checked that the jacobian is one. Since the domain D_1 is defined by $e = 0$, and coordinate τ is the return time of the flow to Σ, we get

$$\Sigma = \{(x, y, \alpha', \beta', \tau, e) \in \mathbb{R}_+^6 \mid \tau = 0, e = 1, \alpha' < \beta', 0 < \beta', \beta' < \alpha' + \gamma\}$$

Furthermore, the invariant measure for the return map $\widetilde{f}$ of the flow to Σ is $dx\,dy\,d\alpha'\,d\beta'$. From this, we obtain:

Proposition 21. *The density function of the invariant measure of $f : \Delta \to \Delta$ for the Cassaigne algorithm is*

$$\frac{1}{(1-x)(1-z)}.$$

Proof. To obtain this density and according to Proposition 10, it suffices to integrate the measure $d\alpha'\, d\beta'$ for a fixed x, y. Given $(x, y) \in \mathbb{R}^2$, the set of admissible $(\alpha', \beta') \in \mathbb{R}^2$ satisfies

$$
\begin{cases}
x\alpha' + y\beta' + \gamma = 1, \\
\alpha' < \beta', \\
0 < \beta', \\
\beta' < \alpha' + \gamma,
\end{cases}
\quad \text{or equivalently} \quad
\begin{cases}
\alpha' < \beta', \\
0 < \beta', \\
(1-x)\alpha' + 1 > (1+y)\beta'.
\end{cases}
$$

This domain is the interior of a triangle of vertices $\left(\frac{1}{x-1}, 0\right)$, $(0, 0)$, $\left(\frac{1}{x+y}, \frac{1}{x+y}\right)$ shown below.
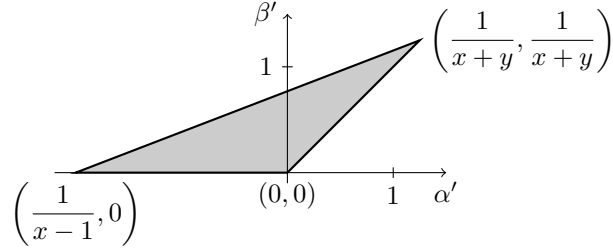
The area of the triangle is

$$
\frac{1}{2}\left(0 - \frac{1}{x-1}\right)\left(\frac{1}{x+y}\right) = \frac{1}{2(1-x)(x+y)}
$$

$\square$

Remark 22. *The total mass of that density is bounded; its value is*

$$
\int_0^1 \int_0^{1-x} \frac{1}{2(1-x)(x+y)}\, dy\, dx = \int_0^1 \frac{\log(x)}{2(x-1)}\, dx = \frac{\pi^2}{12}
$$

6   The Brun algorithm

The Brun algorithm is arguably the simplest multidimensional continued fraction algorithm; it is defined on the positive cone $\Lambda = \{(x_1, x_2, x_3) \in \mathbb{R}^3 \mid x_1 > 0, x_2 > 0, x_3 > 0\}$ by taking the second largest coordinate from the largest one. Formally, we have 6 cases, and consider the following partition (up to a set of measure zero):

$$
\Lambda_\pi = \{(x_1, x_2, x_3) \in \Lambda \mid x_{\pi(1)} < x_{\pi(2)} < x_{\pi(3)}\},
$$

for $\pi \in \mathcal{S}_3$ where $S_3 = \{123, 132, 231, 213, 312, 321\}$ is the set of permutations of $\{1, 2, 3\}$. We define the six elementary matrices:

$$
B_{123} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix}, B_{132} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}, B_{231} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix},
$$

$$
B_{213} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix}, B_{312} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, B_{321} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.
$$

and the matrix function $M : \Lambda \to GL(3, \mathbb{Q})$ is defined according to the above partition:

$$
M : \mathbf{x} \mapsto B_\pi \quad \text{if and only if} \quad \mathbf{x} \in \Lambda_\pi.
$$

14

As in Section 3, the matrix function $M(\mathbf{x})$ allows to define the functions

$$F(\mathbf{x}) = M(\mathbf{x})^{-1}\mathbf{x}, \quad f(\mathbf{x}) = \frac{F(\mathbf{x})}{\|F(\mathbf{x})\|_1} \quad \text{and} \quad \widetilde{F}(\mathbf{x}, \mathbf{a}) = (M(\mathbf{x})^{-1}\mathbf{x}, M(\mathbf{x})^{\top}\mathbf{a}).$$

The six branches of the map F are not full; that is, their image is not the complete cone. It will be useful to define the subcones Θ_π by

$$\Theta_\pi = \{(x_1, x_2, x_3) \in \Lambda \mid x_{\pi(1)} < x_{\pi(2)}\}.$$

Remark 23. *Each Θ_π is a union of three Λ_σ, for all permutation σ such that $\sigma^{-1}\pi(1) < \sigma^{-1}\pi(2)$; a direct computation shows that $F(\Lambda_\pi) = \Theta_\pi$, so that the Λ_π form a Markov partition for F, which can be proved to be generating.*

We want to apply the techniques of Section 3; however, as experiments show, the domain for the natural extension can not be written as a global product, but as a disjoint union of products. Define for each permutation $\pi \in \mathcal{S}_3$,

$$\Lambda_\pi^* = \{(\alpha_1, \alpha_2, \alpha_3) \in \Lambda \mid \alpha_{\pi(1)} < \alpha_{\pi(3)} < \alpha_{\pi(2)}\}$$
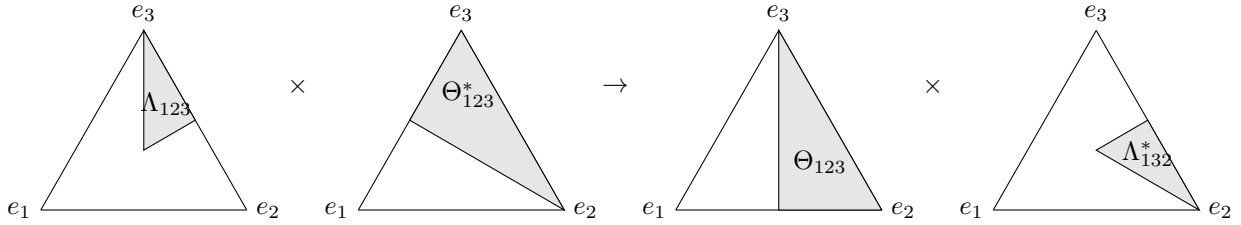$$\Theta_\pi^* = \{(\alpha_1, \alpha_2, \alpha_3) \in \Lambda \mid \alpha_{\pi(1)} < \alpha_{\pi(3)}\}$$

Figure 8: The function $\widetilde{F} : \Lambda_\pi \times \Theta_\pi^* \to \Theta_\pi \times \Lambda_\pi^*$ is one-to-one and onto for each $\pi \in \mathcal{S}_3$.

This is just a convenient notation; if we denote by $\rho = (123)$ the transpositon exchanging 2 and 3, we have $\Lambda_\pi^* = \Lambda_{\pi\rho}$ and $\Theta_\pi^* = \Theta_{\pi\rho}$, and the previous remark applies: each Θ_π^* is the disjoint union of three sets Λ_σ^*. We first prove a lemma.

Lemma 24. *We have:*

$$\bigcup_{\pi \in \mathcal{S}_3} (\Lambda_\pi \times \Theta_\pi^*) = \bigcup_{\pi \in \mathcal{S}_3} (\Theta_\pi \times \Lambda_\pi^*)$$

up to a set of measure 0, where both unions are disjoint.

Proof. Since all Λ_π are disjoint, the first union must be disjoint, and since all Λ_π^* are disjoint, the second union is disjoint.

By the previous remark, each Θ_π^* is the disjoint union of three sets Λ_σ, for those $\sigma \in S_3$ such that $\sigma^{-1}\pi(1) < \sigma^{-1}\pi(3)$. This equality is possible if, and only if, $\sigma(3) = \pi(3)$, or $\sigma(3) = \pi(2)$ and $\sigma(2) = \pi(3)$. Hence the first set is the union of all the 18 products $\Lambda_\sigma \times \Lambda_\pi$, for all pair (σ, π) satisfying this condition.

The second set can be written $\bigcup_{\pi \in \mathcal{S}_3} (\Theta_\pi \times \Lambda_\pi^*) = \bigcup_{\pi \in \mathcal{S}_3} (\Theta_\pi \times \Lambda_{\pi\rho}) = \bigcup_{\pi \in \mathcal{S}_3} (\Theta_{\pi\rho} \times \Lambda_\pi)$, where $\rho = (132)$; and $\Lambda_\sigma \subset \Theta_{\pi\rho}$ if and only if $\sigma^{-1}\pi\rho(1) < \sigma^{-1}\pi\rho(2)$: by definition of ρ, this is the same condition, hence the two sets are equal. □

Proposition 25. *The function $\widetilde{F}$ restricted to the domain $\Lambda_\pi \times \Theta_\pi^*$ is one-to-one and onto $\Theta_\pi \times \Lambda_\pi^*$ for each permutation $\pi \in \mathcal{S}_3$. The function $\widetilde{F}$ on the domain $D = \cup_{\pi \in \mathcal{S}_3} (\Lambda_\pi \times \Theta_\pi^*)$ is one-to-one and onto (See Figure 8).*

Proof. We first prove that $\widetilde{F}(\Lambda_\pi \times \Theta_\pi^*) = \Theta_\pi \times \Lambda_\pi^*$; by symmetry, it is enough to prove it for $\pi = (1, 2, 3)$. Let $(x_1, x_2, x_3, \alpha_1, \alpha_2, \alpha_3) \in \Lambda_\pi \times \Theta_\pi^*$, and let $(y_1, y_2, y_3, \beta_1, \beta_2, \beta_3) = \widetilde{F}(x_1, x_2, x_3, \alpha_1, \alpha_2, \alpha_3) = (x_1, x_2, x_3 - x_2, \alpha_1, \alpha_2 + \alpha_3, \alpha_3)$; we check that $y_1 \leq y_2$ and $\beta_1 \leq \beta_3 \leq \beta_2$, which is the condition to be in $\Theta_\pi \times \Lambda_\pi^*$. To prove equality, it is enough to consider the reciprocal map $\widetilde{F}^{-1}(y_1, y_2, y_3, \beta_1, \beta_2, \beta_3) = (y_1, y_2, y_2 + y_3, \beta_1, \beta_2 - \beta_3, \beta_3)$, and to prove that $\widetilde{F}^{-1}(\Theta_\pi \times \Lambda_\pi^*) = \Lambda_\pi \times \Theta_\pi^*$, which is done exactly the same way.

Since each branch of $\widetilde{F}$ is an invertible linear map, it is a bijection between its domain and its image; by the previous lemma, $\widetilde{F}$ is a bijection on D.

$\square$

Proposition 26. *The density function of the invariant measure of $f : \Delta \to \Delta$ for the Brun algorithm is*

$$\frac{1}{2\, x_{\pi(2)}(1 - x_{\pi(2)})(1 - x_{\pi(1)} - x_{\pi(2)})}$$

on the part $\mathbf{x} = (x_1, x_2, x_3) \in \Lambda_\pi \cap \Delta$.

Proof. In this proof, we suppose that $\pi = (1, 2, 3)$. For Brun algorithm, the surface of section for the part $\vec{x} \in \Lambda_\pi$ is

$$\Sigma = \{(x_1, x_2, x_3, \alpha_1, \alpha_2, \alpha_3) \in \mathbb{R}_+^6 \mid \sum_{i=1}^3 x_i \alpha_i = 1, \sum_{i=1}^3 x_i = 1, \mathbf{x} \in \Lambda_\pi, \mathbf{a} \in \Theta_\pi^*\}.$$

As explained in section 3, we use an explicit change of variables that is useful to compute explicitly all these measures in the case of the norm $\|\mathbf{x}\| = \sum_{i=1}^3 |x_i|$. Define $y_1 = x_1$, $y_2 = x_2$ and $\tau = -\log(\sum_{i=1}^3 |x_i|)$; define $\beta_1 = \alpha_1 - \alpha_3$, $\beta_2 = \alpha_2 - \alpha_3$ and $e = \sum_{i=1}^3 \alpha_i x_i$. A straightforward computation shows that this change of coordinates has jacobian 1. Below, we keep variables x_1 and x_2 instead of y_1 and y_2. We get that Σ is the set of $(x_1, x_2, \beta_1, \beta_2, \tau, e) \in \mathbb{R}^6$ satisfying

$$\begin{cases} \tau = 0, & \beta_1 < 0, \\ e = 1, & \beta_1 + \alpha_3 > 0, \\ x_1 < x_2, & \beta_2 + \alpha_3 > 0, \\ & \alpha_3 > 0. \end{cases}$$

The equalities $e = 1$ and $\sum_{i=1}^3 x_i = 1$ allows to substitute α_3 above. Indeed,

$$1 = e = \sum_{i=1}^3 \alpha_i x_i = (\beta_1 + \alpha_3)x_1 + (\beta_2 + \alpha_3)x_2 + \alpha_3 x_3 = \beta_1 x_1 + \beta_2 x_2 + \alpha_3(x_1 + x_2 + x_3) = \beta_1 x_1 + \beta_2 x_2 + \alpha_3.$$

So we get that Σ is described by

$$\begin{cases} \tau = 0, & \beta_1 < 0, \\ e = 1, & \beta_1 x_1 + \beta_2 x_2 - \beta_1 < 1, \\ x_1 < x_2, & \beta_1 x_1 + \beta_2 x_2 - \beta_2 < 1, \\ & \beta_1 x_1 + \beta_2 x_2 < 1. \end{cases}$$

16

From Proposition 10, the density is equal to the volume of the polytope $\{\mathbf{a} : (\mathbf{x}, \mathbf{a}) \in \Sigma\}$:

$$\delta(x_1, x_2) = \int_{\{(\beta_1,\beta_2):(x_1,x_2,0,\beta_1,\beta_2,1)\in\Sigma\}} 1 \, d\beta_1 \, d\beta_2$$

Given $(x_1, x_2) \in \mathbb{R}^2$, the polytope corresponds to the set of $(\beta_1, \beta_2) \in \mathbb{R}^2$ satisfying the four inequalities. This domain is the interior of a triangle with vertices $\left(0, \frac{1}{x_2-1}\right)$, $\left(0, \frac{1}{x_2}\right)$, $\left(\frac{1}{x_1+x_2-1}, \frac{1}{x_1+x_2-1}\right)$ shown below.
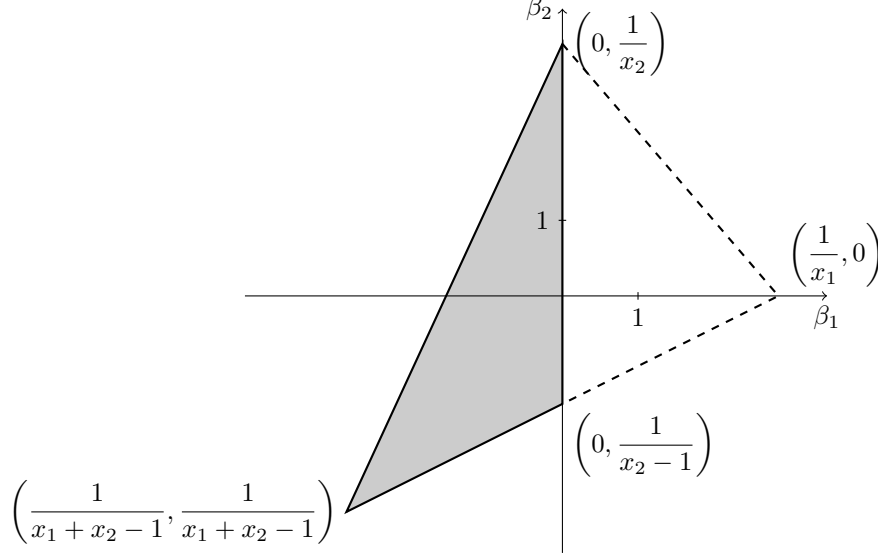
Its area is

$$\frac{1}{2}\left(\frac{1}{x_2} - \frac{1}{x_2-1}\right)\left(0 - \frac{1}{x_1+x_2-1}\right) = \frac{1}{2x_2(1-x_2)(1-x_1-x_2)} \qquad \square$$

The mass of that density for the part $x_1 < x_2 < x_3$ is:

$$\int_0^{\frac{1}{3}} \int_{x_1}^{\frac{1}{2}-\frac{x_1}{2}} \frac{1}{2x_2(1-x_2)(1-x_1-x_2)} dx_2 dx_1 = -\frac{\pi^2}{24} + \frac{1}{2}\log(3)\log(2) - \frac{1}{2}\mathrm{Li}_2(\frac{2}{3}) + \frac{3}{2}\mathrm{Li}_2(\frac{1}{3}) - \mathrm{Li}_2(-\frac{1}{3})$$

we will prove in Section 8 that this quantity is equal to $\frac{\pi^2}{24}$.

7 Experimentations: domain of the natural extension

Since the map F is linear, the trajectory of $\lambda \mathbf{x}$ is homothetic to that of $\mathbf{x}$; hence, we can always normalize $\mathbf{x}$ to $\frac{\mathbf{x}}{\|\mathbf{x}\|}$, and consider it as a point in the unit simplex. We can do the same thing for $\mathbf{a}$, and both $(\mathbf{x}, \mathbf{a}) \in \mathbb{R}^d \times \mathbb{R}^d$ and $\widetilde{F}(\mathbf{x}, \mathbf{a})$, its image under $\widetilde{F}$, can be represented graphically by a pair of points in a pair of unit simplexes. This gives four points in four distinct simplexes. To visualise the effect of $\widetilde{F}$, we draw each of these four points with the color according to which element of the partition of Λ the point $\mathbf{x}$ belongs. We could do this exercice for a set of random points $(\mathbf{x}, \mathbf{a}) \in \mathbb{R}^d \times \mathbb{R}^d$, but in order to find the domain of the natural extension, it is better to consider some orbit $(\widetilde{F}^n(\mathbf{x}, \mathbf{a}))_{n \geq 0}$ for some starting point $(\mathbf{x}, \mathbf{a}) \in \mathbb{R}^d \times \mathbb{R}^d$. Below are the pictures we get for some Multidimensional Continued Fraction algorithms.

The experimentation shows that a domain of positive measure can be expected for Brun and Reverse algorithms. The method also works for Selmer algorithm. The method does not work for
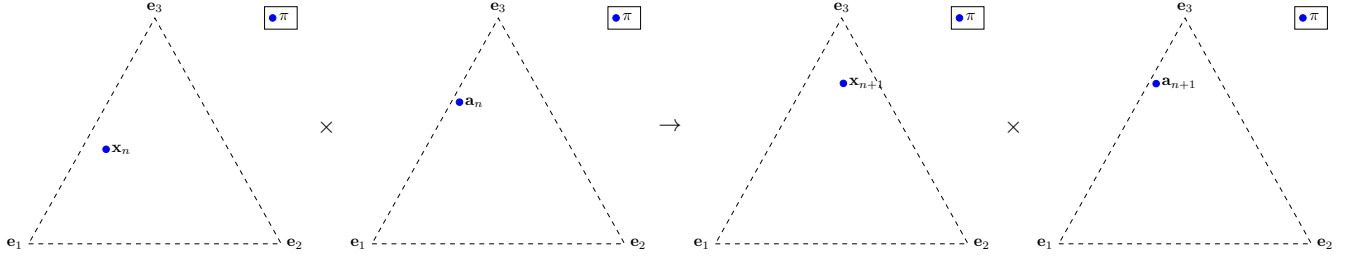
Figure 9: For some starting point $(\mathbf{x}, \mathbf{a}) \in \mathbb{R}^d \times \mathbb{R}^d$, we consider the sequence $(\mathbf{x}_n, \mathbf{a}_n) = F^n(\mathbf{x}, \mathbf{a})$ for $n \geq 0$. The points $\mathbf{x}_n$, $\mathbf{a}_n$, $\mathbf{x}_{n+1}$ and $\mathbf{a}_{n+1}$ are drawn in four different plots with a color or label π associated to the matrix $M(\mathbf{x}_n)$. This experimentation allows to view the domain of the natural extension of F.
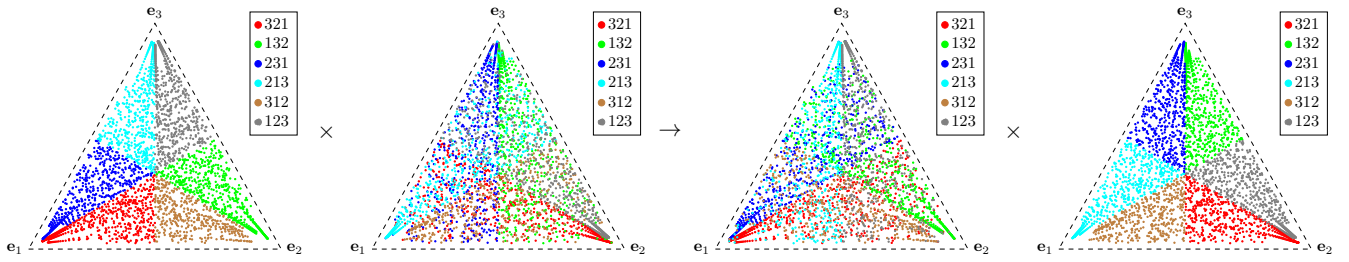

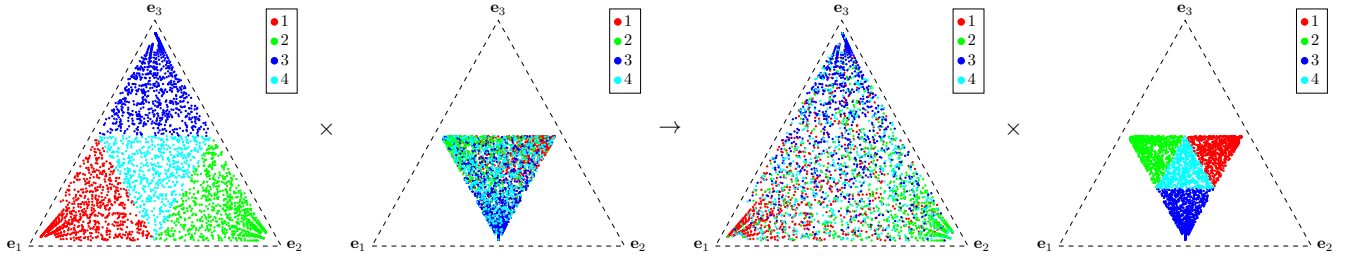Figure 10: Experimental natural extension of Brun algorithm.

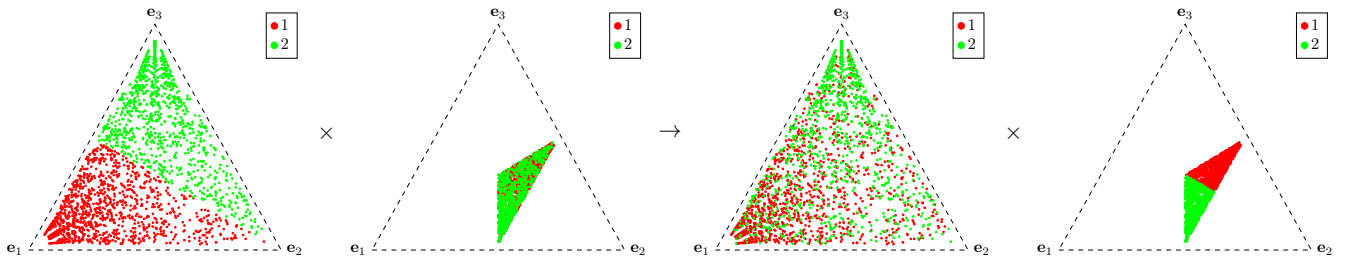Figure 11: Experimental natural extension of Reverse algorithm.

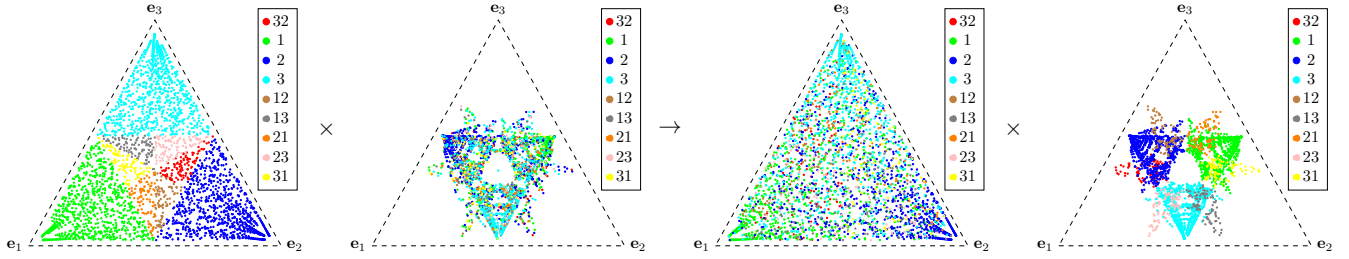Figure 12: Experimental natural extension of Cassaigne algorithm.

Figure 13: Experimental natural extension of Arnoux-Rauzy-Poincaré algorithm.

Poincaré and Meester algorithms: both do not have a finite measure preserving transformation. Note that Poincaré does have an invariant measure equivalent to Lebesgue but it is not conservative. Finally, the domain for AR-Poincaré [BL15] and for the additive version of Jacobi-Perron algorithm are fractals and defined by an IFS. Both of these algorithms are known to have an invariant measure absolutely continuous with respect to Lebesgue measure; this was shown for Jacobi-Perron algorithm by Broise and Guivarc'h [BAG01], and will will prove it for AR-Poincaré in a forthcoming paper, but we do not know their density function.

In Figure 14, we compute the first 2000000 iterations of an orbit in a 1024×1024 pixels picture.
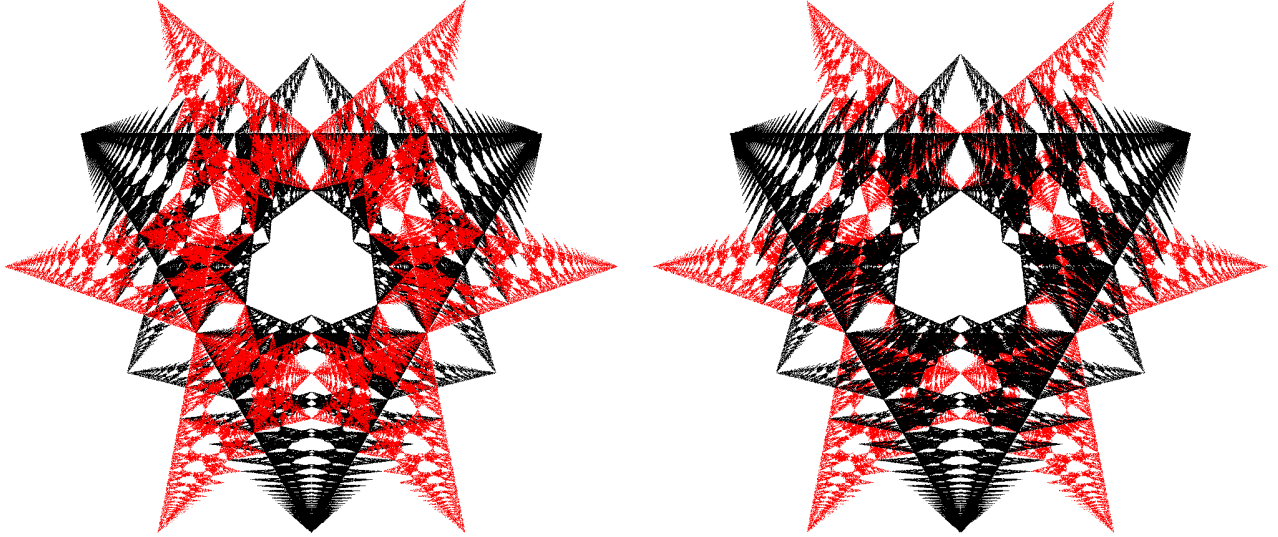

Figure 14: 2×10^6 iterations shown in a 1024×1024 pictures of the window $[-0.6, 0.6] \times [-0.6, 0.6]$. On the left, points in the six Poincaré branches are drawn last. On the right, points in the three Arnoux-Rauzy branches are drawn last.

This set presents a symmetry of order 3; it is decomposed in 9 subsets, and one can prove that there are 3 disjoint groups of three sets. It is unclear wether these sets contain open balls, have positive measure, or even dimension 2; it is also unclear if the three sets in each groups are disjoint, but if this is the case, it seems difficult to prove.
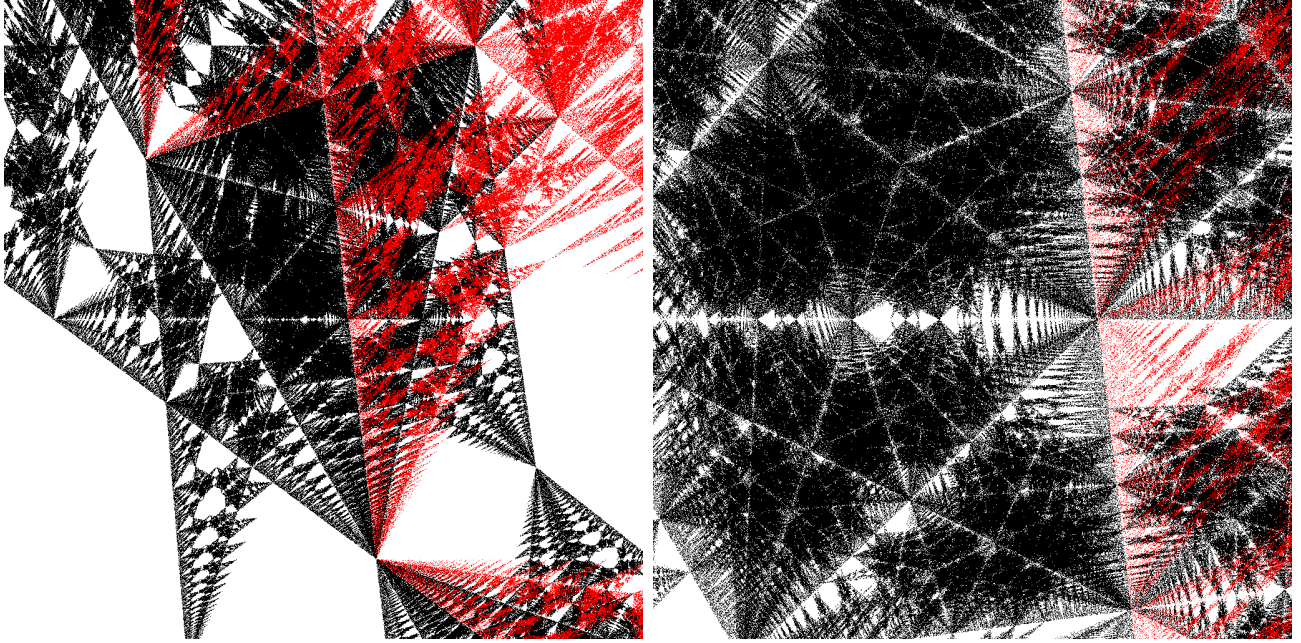
Figure 15: Left:10^8 iterations and zoom in $[0.05, 0.15] \times [0.05, 0.15]$. Right:$10^9$ iterations and zoom in $[0.09, 0.11] \times [0.09, 0.11]$.

8 Additional remarks

8.1 Invariant domain and fixed point theorem

Of course, the essential element for this construction is to find the invariant domain D. It can sometimes be proved from elementary considerations, as we have seen in the two examples above, and numerical experimentations can be useful.

When no obvious solution is found, one can use the fact that the formula for $\widetilde{f}$ is known. For fixed $\mathbf{x}$, $\widetilde{f}$ is an affine map on an hyperplane in Λ^*. In some cases, it can be proved that this collection of maps depending on $\mathbf{x}$ is uniformly contracting, and this implies that there exists a unique compact invariant set which is fixed by $\widetilde{f}$, see [AS15]. Numerical experiments seem to show that this property is still valid under much weaker conditions. This allows, as we have seen above, to guess the domain.

However, this domain is interesting only if it has strictly positive measure, which is not clear in a number of cases; even in that case, it can be very difficult to describe, as shown by the case of the japanese continued fraction, which has been very precisely studied, see for example [KSS12].

Remark also that a very well studied case, that of Rauzy induction, show the complications which can appear : if one uses exactly the procedure we describe in the present paper, one does not get a natural extension, because one gets a subset of lower dimension (given by the subspace called $H(\pi)$ by Veech, see [Vee82]), and it is necessary to add other coordinates to get the natural extension.

8.2 Symmetric algorithms

The group S_d of permutations on d elements acts in a natural way on $\mathbb{R}^d$ by $\sigma.(x_1, \ldots, x_d) = (x_{\sigma(1)}, \ldots, x_{\sigma(d)})$. We will say that a map F is *symmetric* if, for all permutations σ, $F(\sigma.\mathbf{x}) =$

$\sigma.F(\mathbf{x})$. Brun and Reverse algorithms are symmetric; a classical example of a non-symmetric algorithm is Jacobi-Perron algorithm: a permutation of coordinates gives in general a very different orbit.

When we have a symmetric algorithm defined on the positive cone, we can quotient by S_d, and restrict the algorithm to the subset defined by $0 < x_1 < x_2 < \cdots < x_d$. The original symmetric algorithm is often named the *unsorted algorithm*, and the quotient algorithm is named the *sorted algorithm*. These algorithms are essentially equivalent, since the sorted algorithm is a $d!$-to-1 factor of the unsorted one. The sorted algorithm is generally more convenient for explicit computations, in particular numerical experiments, since the relations between coordinates are known; but the unsorted algorithm, being more symmetric, gives usually simpler and more natural formulas, and is better adapted to theoretical considerations; another interesting feature is that, in classical examples, the unsorted algorithm preserves the orientation, but the sorted algorithm does not, since transpositions reverse the orientation.

This leads to many apparently distinct, but essentially equivalent, presentations for the same algorithm; this is already true for the basic Farey algorithm, which takes several forms with similar properties.

8.3   Parabolic points and acceleration

We say that a continued fraction algorithm has an indifferent fixed point if one of the branches of the projective map f associated to this algorithm has a fixed point where the Jacobian matrix is the identity (this does not depend on the choice of coordinates). Several classical algorithms have one or several indifferent fixed points; this presents the inconvenient, in dimension 1, that the invariant measure becomes infinite, and the entropy zero.

A well-known way to get rid of this problem consists in "accelerating" the algorithm; more precisely, let f_i be the branch of f which has an indifferent fixed point, and let A_i be its domain. For any x different of the fixed point, define n_x as the smallest n such the $f^n(x) \notin A_i$, and replace the branch f_i by the countable collection of branches $x \mapsto f^{n_x}(x)$. This replaces what is generally called an "additive" algorithm with a finite number of branches by a "multiplicative" algorithm, with a cointable number of branches, no indifferent fixed point, and with a finite invariant measure.

8.4   Changes of coordinates

As we said before, a continued fraction map in the usual sense is a projective map; hence there is no canonical coordinates for explicit computations, and many things depend in particular on the choice of the section Σ. As an example, we show another section for Brun's algorithm, which will allow us to compute the total mass of the invariant measure in a different way.

Instead of using the norm $\sum_{i=1..3} |x_i|$, we chose the norm $\sup_{i=1..3} |x_i|$, and we work out the proof of Proposition 26 in this setting. We suppose as before that $\pi = (1, 2, 3)$. The corresponding section, for this choice of norm, is

$$\Sigma = \{(x_1, x_2, x_3, \alpha_1, \alpha_2, \alpha_3) \in \mathbb{R}_+^6 \mid \sum_{i=1}^{3} x_i \alpha_i = 1, x_1 < x_2 < x_3 = 1, \alpha_1 < \alpha_3\}.$$

We make a different choice of coordinates here, adapted to this norm : keep x_1, x_2, α_1 and α_2; define $\tau = -\log(|x_3|)$ and $e = \sum_{i=1}^{3} \alpha_i x_i$. A straightforward computation shows that this change of coordinates has jacobian 1.

We get that Σ is the set of $(x_1, x_2, \beta_1, \beta_2, \tau, e) \in \mathbb{R}^6$ satisfying

$$\begin{cases} 0 < x_1 < x_2 < 1, & \tau = 0, \quad e = 1 \\ 0 < \alpha_1 < \alpha_3, & 0 < \alpha_2. \end{cases} \quad .$$

The equalities $e = 1$ and $x_3 = 1$ allows to substitute α_3 above, hence Σ is also described by

$$\begin{cases} 0 < x_1 < x_2 < 1, & \tau = 0, \quad e = 1 \\ \alpha_1(1 + x_1) + \alpha_2 x_2 < 1, & 0 < \alpha_1, \quad 0 < \alpha_2. \end{cases}$$

Hence the density is equal to the volume of the polytope $\{\mathbf{a} : (\mathbf{x}, \mathbf{a}) \in \Sigma\}$:

$$\delta(x_1, x_2) = \int_{\{(\alpha_1, \alpha_2) : (x_1, x_2, 0, \alpha_1, \alpha_2, 1) \in \Sigma\}} 1 \, d\alpha_1 \, d\alpha_2$$

This is again given by a triangle, this time with vertices $(0, 0)$, $(1/(1+x_1), 0)$, $(0, 1/x_2)$. Hence the invariant density is given in these coordinates by $\frac{1}{2x_2(1+x_1)}$.

The mass of that density on the triangle $0 < x_1 < x_2 < 1$ is:

$$\int_0^1 \int_0^{x_2} \frac{1}{2x_2(1+x_1)} dx_1 dx_2 = \int_0^1 \frac{\log(1+x_2)}{2x_2} \, dx_2$$

Taking the series expansion of $\log(1 + x_2)$ and integrating, we find that this quantity is equal to the series $\sum_{n=1}^{\infty} \frac{(-1)^{n+1}}{2n^2} = \frac{\pi^2}{24}$, as announced at the end of section 5. This is just a change of variable in the integral, which can be made completely explicit. Since we made the computation on a sixth of the surface, the total mass of the invariant measure for the unsorted algorithm is $\frac{\pi^2}{4}$.

The equality $-\frac{\pi^2}{24} + \frac{1}{2}\log(3)\log(2) - \frac{1}{2}\operatorname{Li}_2(\frac{2}{3}) + \frac{3}{2}\operatorname{Li}_2(\frac{1}{3}) - \operatorname{Li}_2(-\frac{1}{3}) = \frac{\pi^2}{24}$ is probably well-known among the many identities for the dilogarithm function, but we could not find it in the litterature.

8.5 Variants of the classical continued fraction

We have studied in Section 2 the unsorted Farey map. One of the advantages of this map is that it preserves orientation, and is related to matrices in $SL(2, \mathbb{R})$. However, most presentations prefer to deal with the sorted map, restricting to the cone $x < y$, and in that case it is convenient to take as section $y = 1$; this leads to the usual sorted Farey map, given by $x \mapsto \frac{x}{1-x}$ if $x < \frac{1}{2}$ and $x \mapsto \frac{1}{x} - 1$ if $x > \frac{1}{2}$.

This map has only one indifferent fixed point, at 0. Taking the acceleration of the map at this fixed point leads to the usual Gauss map, $x \mapsto \{\frac{1}{x}\}$.

Many other variants are possible, by taking different sections; they are all related to the geodesic flow on the modular surface and the group $SL(2, \mathbb{Z})$.

8.6 Higher dimensions

It is natural to generalize to higher dimensions. For the reverse algorithm, a direct generalization is not possible: the basis of the algorithm is that the complement in the unit simplex of the three simplexes of size $\frac{1}{2}$ given by the condition $x_i > \frac{1}{2}$ is also a simplex, which can be sent to the whole simplex by an homothety of ratio -2. In dimension 4, the complement, in the unit simplex, of the four simplexes $x_i > \frac{1}{2}$ is no more a simplex, but an octahedron, so we cannot extend our simple solution to that case.

But it is possible, and easy, to generalize Brun algorithm to any dimension. We define it by subtracting the second largest coordinate by the largest coordinate. We can define an invariant domain for the natural extension in the same way. Let Λ be the positive cone in $\mathbb{R}^d$, and define, for $\pi \in S_d$

$$\Lambda_\pi = \{(x_1, \ldots, x_d) \in \Lambda \mid x_{\pi(1)} < x_{\pi(2)} < \cdots < x_{\pi(d)}\}$$
$$\Theta_\pi = \{(x_1, \ldots, x_d) \in \Lambda \mid x_{\pi(1)} < x_{\pi(2)} < \cdots < x_{\pi(d-1)}\}$$
$$\Lambda_\pi^* = \{(\alpha_1, \ldots, \alpha_d) \in \Lambda \mid \alpha_{\pi(i)} < \alpha_{\pi(d)} < \alpha_{\pi(d-1)} \text{ if } i < d - 1\}$$
$$\Theta_\pi^* = \{(\alpha_1, \ldots, \alpha_d) \in \Lambda \mid \alpha_{\pi(i)} < \alpha_{\pi(d)} \text{ if } i < d - 1\}$$

The same proof as in dimension 3, with heavier notations, proves that $\widetilde{F}$ sends $\Lambda_\pi \times \Theta_\pi^*$ to $\Theta_\pi \times \Lambda_\pi^*$. It is then straightforward to compute the invariant measure; similar formulas were given in [AN93] and [HK01].

Proposition 27. *The density function of the invariant measure of $f : \Delta \to \Delta$ for the d-dimensional Brun algorithm is*

$$\delta(\mathbf{x}) = \frac{1}{(d-1)!\, x_{\pi(d-1)}} \sum_{A_1 \subset A_2 \subset \cdots \subset A_{d-1}} \prod_{k=1}^{d-1} \frac{1}{1 - \sum_{i \in A_k} x_i}$$

on the part $\mathbf{x} \in \Lambda_\pi$ where $A_1 = \{\pi(d-1)\}$, $A_{d-1} = \{1, 2, \ldots, d\} \setminus \{\pi(d)\}$ and $|A_k| = k$.

For example, taking $\pi = Id$, when $d = 2$:

$$\delta(x_1) = \frac{1}{x_1(1 - x_1)}$$

When $d = 3$:
$$\delta(x_1, x_2) = \frac{1}{2\, x_2\, (1 - x_1 - x_2)(1 - x_2)}$$

When $d = 4$:
$$\delta(x_1, x_2, x_3) = \frac{\frac{1}{(1-x_1-x_3)(1-x_3)} + \frac{1}{(1-x_2-x_3)(1-x_3)}}{6\, x_3\, (1 - x_1 - x_2 - x_3)}$$

When $d = 5$, $\delta(x_1, x_2, x_3, x_4)$ is

$$\frac{\frac{\frac{1}{(1-x_1-x_4)(1-x_4)} + \frac{1}{(1-x_2-x_4)(1-x_4)}}{1-x_1-x_2-x_4} + \frac{\frac{1}{(1-x_1-x_4)(1-x_4)} + \frac{1}{(1-x_3-x_4)(1-x_4)}}{1-x_1-x_3-x_4} + \frac{\frac{1}{(1-x_2-x_4)(1-x_4)} + \frac{1}{(1-x_3-x_4)(1-x_4)}}{1-x_2-x_3-x_4}}{24\, x_4\, (1 - x_1 - x_2 - x_3 - x_4)}$$

Proof. In this proof, we suppose that $\pi = (1, 2, \ldots, d)$. For Brun algorithm, the surface of section for the part $\mathbf{x} \in \Lambda_\pi$ is

$$\Sigma = \{(x_1, \ldots, x_d, \alpha_1, \ldots, \alpha_d) \in \mathbb{R}_+^{2d} \mid \sum_{i=1}^d x_i \alpha_i = 1, \sum_{i=1}^d x_i = 1, \mathbf{x} \in \Lambda_\pi, \mathbf{a} \in \Theta_\pi^*\}.$$

As explained in section 3, we use an explicit change of variables that is useful to compute explicitly all these measures in the case of the norm $\|\mathbf{x}\| = \sum_{i=1}^d |x_i|$. Define $y_i = x_i$ for $i < d$, and $y_d = \log(\sum_{i=1}^d |x_i|)$; define $\beta_i = \alpha_i - \alpha_d$ for $i < d$ and $\beta_d = \sum_{i=1}^d \alpha_i x_i$. A straightforward computation

23

shows that this change of coordinates has jacobian 1. Below, we keep variables x_i for $i < d$ instead of y_i. We get that Σ is the set of $(x_1, \ldots, x_{d-1}, y_d, \beta_1, \ldots, \beta_{d-1}, \beta_d) \in \mathbb{R}^{2d}$ satisfying

$$\begin{cases} y_d = 0, & \beta_i < 0 \text{ for } i < d-1, \\ \beta_d = 1, & \beta_i + \alpha_d > 0 \text{ for } i < d, \\ x_1 < \cdots < x_{d-1}, & \alpha_d > 0. \end{cases}$$

The equalities $\beta_d = 1$ and $\sum_{i=1}^d x_i = 1$ allows to substitute α_d above. Indeed,

$$1 = \beta_d = \sum_{i=1}^d \alpha_i x_i = \sum_{i=1}^{d-1}(\beta_i + \alpha_d)x_i + \alpha_d x_d = \sum_{i=1}^{d-1}\beta_i x_i + \alpha_d \sum_{i=1}^d x_i = \sum_{i=1}^{d-1}\beta_i x_i + \alpha_d.$$

So we get that Σ is described by

$$\begin{cases} y_d = 0, & \beta_i < 0 \text{ for } i < d-1, \\ \beta_d = 1, & \sum_{j=1}^{d-1}\beta_j x_j - \beta_i < 1 \text{ for } i < d, \\ x_1 < \cdots < x_{d-1}, & \sum_{j=1}^{d-1}\beta_j x_j < 1. \end{cases}$$

From Proposition 10, the density $\delta(\mathbf{x})$ is equal to the volume of the polytope $\{\mathbf{b} : (\mathbf{x}, \mathbf{b}) \in \Sigma\}$:

$$\delta(\mathbf{x}) = \int_{\{\mathbf{b}:(\mathbf{x},\mathbf{b})\in\Sigma\}} 1 \, d\mathbf{b}$$

Given $\mathbf{x} = (x_1, \ldots, x_{d-1}) \in \mathbb{R}^{d-1}$, the polytope corresponds to the set of $\mathbf{b} = (\beta_1, \ldots, \beta_{d-1}) \in \mathbb{R}^{d-1}$ satisfying the above inequalities. We conclude the proof by using Lemma 28. $\qquad\square$

Lemma 28. *Let $I \subset \{1, 2, \ldots, d\}$ and $\ell \in I$. Let $(x_1, x_2, \ldots, x_d) \in \mathbb{R}_+^d$ be positive real numbers such that $\sum_{i=1}^d x_i = 1$. Let $P_{I,\ell}$ be the $|I|$-dimensional polytope defined by the set of real vectors $(\beta_i : i \in I)$ that satisfy:*

$$\begin{cases} \beta_i < 0 \text{ for } i \in I \setminus \{\ell\}, \\ \sum_{j\in I}\beta_j x_j - \beta_i < 1 \text{ for } i \in I, \\ \sum_{j\in I}\beta_j x_j < 1. \end{cases}$$

Then

$$Vol(P_{I,\ell}) = \frac{1}{|I|(1 - \sum_{i\in I} x_i)} \left(\sum_{k\in I\setminus\{\ell\}} Vol(P_{I\setminus\{k\},\ell}) \right)$$

Proof. First notice that $\beta_i = \frac{1}{(1-\sum_{i\in I} x_i)}$ for all $i \in I$ is a solution to the second equation. This is a vertex of the polytope that we denote S. The interior of the polytope can be partionned into $|I|-1$ part p_k according to the index $k \in I \setminus \{\ell\}$ for which $\beta_k \geq \beta_i$ for all $i \in I \setminus \{\ell\}$. Note that the vertex S is also a vertex of each of these parts. In fact, we can see that each part is in fact a pyramid with top vertex S and base given by the equation $\beta_k = 0$. That base is a $(|I| - 1)$-dimensional polytope given by the equations

$$\begin{cases} \beta_i < 0 \text{ for } i \in I \setminus \{k, \ell\}, \\ \sum_{j\in I\setminus\{k\}}\beta_j x_j - \beta_i < 1 \text{ for } i \in I \setminus \{k\}, \\ \sum_{j\in I\setminus\{k\}}\beta_j x_j < 1 \end{cases}$$

which is exactly $P_{I\setminus\{k\},\ell}$ and whose volume is $Vol(P_{I\setminus\{k\},\ell})$. $\qquad\square$

Acknowledgements

References

[AN93]   Pierre Arnoux and Arnaldo Nogueira. Mesures de Gauss pour des algorithmes de fractions continues multidimensionnelles. *Ann. Sci. École Norm. Sup. (4)*, 26(6):645–664, 1993.

[AS13]   Pierre Arnoux and Štěpán Starosta. The Rauzy gasket. In Julien Barral and Stéphane Seuret, editors, *Further Developments in Fractals and Related Fields*, Trends in Mathematics, pages 1–23. Birkhäuser Boston, 2013.

[AS15]   Pierre Arnoux and Thomas A. Schmidt. Natural extensions and gauss measures for piecewise homographic continued fractions. (preprint) 2015.

[BAG01]   A. Broise-Alamichel and Y. Guivarc'h. Exposants caractéristiques de l'algorithme de Jacobi-Perron et de la transformation associée. *Ann. Inst. Fourier (Grenoble)*, 51(3):565–686, 2001.

[BL15]   V. Berthé and S. Labbé. Factor complexity of S-adic words generated by the Arnoux-Rauzy-Poincaré algorithm. *Adv. in Appl. Math.*, 63:90–130, 2015.

[Bre81]   A. J. Brentjes. *Multidimensional continued fraction algorithms*. Mathematisch Centrum, Amsterdam, 1981.

[HK01]   D. M. Hardcastle and K. Khanin. Continued fractions and the d-dimensional Gauss transformation. *Comm. Math. Phys.*, 215(3):487–515, 2001.

[KSS12]   Cor Kraaikamp, Thomas A. Schmidt, and Wolfgang Steiner. Natural extensions and entropy of α–continued fractions. *Ann. Sci. École Norm. Sup. (4)*, 25(8):2207–2243., 2012.

[Sch00]   F. Schweiger. *Multidimensional Continued Fraction*. Oxford Univ. Press, New York, 2000.

[Vee82]   William A. Veech. Gauss measures for transformations on the space of interval exchange maps. *Ann. of Math. (2)*, 115(1):201–242, 1982.