

# A computer scientist's perspective on approximation of IFS invariant sets and measures with the random iteration algorithm—proofs and examples

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**Abstract**—We present proofs of the theorems and lemmas demonstrated previously in our paper [1]. We also display some visual examples of minimal absorbing sets and their basins of attractions generated by  $\delta$ -roundoffs of two-dimensional linear contractions as well as visualizations of DIFS stationary probability measures.

**Keywords**—IFS; Discrete Space; Markov Chain; Approximation; Invariant Set; Invariant Measure

## I. PROOFS OF THEOREMS AND LEMMAS

**I**N this section we demonstrate proofs of the theorems and lemmas we presented previously in [1].

**Proof of Theorem II.1.** (c) For each  $\tilde{x} \in C$  and each  $\Lambda_i$  from the family of absorbing sets, there is  $N(i) \in \mathbb{N}$  such that for all  $m \geq N(i)$ ,  $\tilde{w}^{om}(\tilde{x}) \in \Lambda_i$ . Hence, for all  $m \geq \max_{1 \leq i \leq K} N(i)$ ,  $\tilde{w}^{om}(\tilde{x}) \in \bigcap_i \Lambda_i$ .

(d) For any  $\tilde{x} \in C$ , there is  $N \in \mathbb{N}$  such that  $\{\tilde{w}^{oi}(\tilde{x})\}_{i \geq N} \subset \Lambda$ . Hence,  $\tilde{w}(\{\tilde{w}^{oi}(\tilde{x})\}_{i \geq N}) \subset \tilde{w}(\Lambda)$ , or equivalently  $\{\tilde{w}^{oi}(\tilde{x})\}_{i \geq N+1} \subset \tilde{w}(\Lambda)$ .

(e) Let  $\tilde{x} \in \Lambda$ . The set  $\Lambda$  is absorbing in  $C$ , so for each  $\tilde{x} \in \Lambda$  there exists  $M(\tilde{x}) \in \mathbb{N}$  such that  $B(\tilde{x}) := \{\tilde{w}^{oi}(\tilde{x})\}_{i \geq M(\tilde{x})} \subset \Lambda$ . By definition,  $\tilde{w}(B(\tilde{x})) \subset B(\tilde{x})$ . Therefore, the set  $B = \bigcup_{\tilde{x} \in \Lambda} B(\tilde{x})$  has the property that  $B \subset \Lambda$  and  $\tilde{w}(B) \subset B$ . Clearly,  $B$  is also absorbing in  $C$ , because for any  $\tilde{x} \in C$  there is  $N \in \mathbb{N}$  such that  $\tilde{w}^{oN}(\tilde{x}) \in \Lambda$ , so for all  $i \geq M(\tilde{x}) + N$ ,  $\tilde{w}^{oi}(\tilde{x}) \in B$ .

(f) Let  $\tilde{x} \in D$ . Since  $C$  is an absorbing set in  $D$ , there is  $i \in \mathbb{N}$  such that  $\tilde{w}^{oi}(\tilde{x}) \in C$ . Since  $\Lambda$  is an absorbing set in  $C$ , there exists  $N \in \mathbb{N}$  such that for all  $k \geq N$ ,  $\tilde{w}^{o(i+k)}(\tilde{x}) \in \Lambda$ .  $\square$

**Proof of Theorem II.2.** First we show that  $\mathcal{M}$  includes all periodic points in  $C$ . We have to show that given any periodic point  $\tilde{x} \in C$ ,  $\tilde{x}$  belongs to every absorbing set  $\Lambda_\alpha$  for  $\tilde{w}$ . On the contrary let us assume that  $\tilde{x} \notin \Lambda_\alpha$  for some  $\alpha \in \mathcal{I}$ . Since  $\tilde{x}$  is a periodic point, there exists  $k \in \mathbb{N}$  such that  $\tilde{w}^{ok}(\tilde{x}) = \tilde{x}$ , and thus  $\tilde{w}^{o(\alpha k)}(\tilde{x}) = \tilde{x}$  for any  $\alpha \in \mathbb{N}$ . It follows that for any  $N \in \mathbb{N}$  there is  $i \geq N$  such that  $\tilde{w}^{oi}(\tilde{x}) = \tilde{x} \notin \Lambda_\alpha$ , which contradicts the assumption of  $\Lambda_\alpha$  being an absorbing set for  $\tilde{w}$ .

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Therefore, each absorbing set  $\Lambda_\alpha$  has to include all periodic points in  $C$ , and hence they are included in the intersection of the sets.

Now we show that if  $\tilde{x} \in \mathcal{M}$ , then  $\tilde{x}$  must be periodic. On the contrary, let us assume that  $\tilde{x} \in \mathcal{M}$  and  $\tilde{x}$  is non-periodic, which implies that  $\tilde{w}^{oi}(\tilde{x}) \neq \tilde{x}$  for all  $i \in \mathbb{N}$ . Let  $\Lambda$  be any absorbing set such that  $\tilde{x} \in \Lambda$ . We will show that  $\Lambda \setminus \{\tilde{x}\}$  is an absorbing set in  $C$  and thus  $\tilde{x}$  cannot be in  $\mathcal{M}$ . Since  $\Lambda$  is absorbing in  $C$ , we get that for any  $\tilde{y} \in C$  there is  $N(\tilde{y}) \in \mathbb{N}$  so that  $\tilde{w}^{oi}(\tilde{y}) \in \Lambda$  for all  $i \geq N(\tilde{y})$ . Therefore, if  $\tilde{y}$  is such that  $\tilde{w}^{oi}(\tilde{y}) \neq \tilde{x}$  for all  $i \in \mathbb{N}$ , then for all  $i \geq N(\tilde{y})$ ,  $\tilde{w}^{oi}(\tilde{y}) \in \Lambda \setminus \{\tilde{x}\}$ . Hence  $\Lambda \setminus \{\tilde{x}\}$  absorbs the orbits  $\{\tilde{w}^{oi}(\tilde{y})\}_{i \in \mathbb{N}}$  that do not intersect  $\{\tilde{x}\}$ . Now, because  $\tilde{x}$  is non-periodic,  $\{\tilde{w}^{oi}(\tilde{x})\}_{i \in \mathbb{N}}$  does not intersect  $\{\tilde{x}\}$ , and thus  $\tilde{w}^{oi}(\tilde{x}) \in \Lambda \setminus \{\tilde{x}\}$  for all  $i \geq N(\tilde{x})$ . Therefore, for any  $\tilde{y} \in C$  such that  $\tilde{w}^{ok}(\tilde{y}) = \tilde{x}$  for a certain  $k \in \mathbb{N}$ , we get that  $\tilde{w}^{oi}(\tilde{y}) \in \Lambda \setminus \{\tilde{x}\}$  for all  $i \geq N(\tilde{x}) + k$ . Hence  $\Lambda \setminus \{\tilde{x}\}$  also absorbs the orbits that do intersect  $\{\tilde{x}\}$ . Therefore,  $\Lambda \setminus \{\tilde{x}\}$  is an absorbing set in  $C$ , and it follows that  $\tilde{x}$  is not in  $\mathcal{M}$ , so we have arrived at a contradiction, which completes the proof.  $\square$

**Proof of Theorem II.4.** By Theorem II.1 (b) there is at least one absorbing set in  $C$ . Moreover,  $C$  is bounded and thus finite, so there is at most a finite number of absorbing sets in  $C$ . Hence, the the conclusion follows from Theorem II.1 (c).  $\square$

**Proof of Theorem II.5.** By Theorem II.2,  $\mathcal{M}[\tilde{w}, C]$  consists of all periodic points of  $\tilde{w}$  in  $C$ . Naturally, the points remain periodic in every superset of  $C$ , in this instance the set  $D$ . Hence, every absorbing set in  $D$  has to include all periodic points in  $D$  (otherwise the set would not be absorbing in  $D$ ), and we get that the intersection of all absorbing sets in  $D$  is nonempty and include  $\mathcal{M}[\tilde{w}, C]$ . Moreover,  $\mathcal{M}[\tilde{w}, C]$  is an absorbing set in  $C$ , and by the assumption of the theorem,  $C$  is an absorbing set in  $D$ , so from Theorem II.1 (f) it follows that  $\mathcal{M}[\tilde{w}, C]$  is also an absorbing set in  $D$ , and thus  $\mathcal{M}[\tilde{w}, C]$  include the intersection of all absorbing sets in  $D$ . Since, as we have shown,  $\mathcal{M}[\tilde{w}, C]$  is also included in the intersection, this completes the proof.  $\square$

**Proof of Theorem II.7.** Let  $\tilde{x}$  be any point of  $\mathcal{D}^n(\delta)$ . The mapping  $w$  is a contraction on  $(\mathbb{R}^n, d)$ , so  $\lim_{i \rightarrow \infty} w^{oi}(\tilde{w}) =$



$x_f$  or equivalently

$$\forall \varepsilon > 0, \exists N \in \mathbb{N}, \forall i \geq N, d(w^{oi}(\tilde{x}), x_f) < \varepsilon.$$

As a consequence, for a given  $\varepsilon > 0$ , there exists  $N \in \mathbb{N}$  such that for any  $i \geq N$  we get that

$$\begin{aligned} d(\tilde{w}^{oi}(\tilde{x}), x_f) &\leq d(\tilde{w}^{oi}(\tilde{x}), w^{oi}(\tilde{x})) + d(w^{oi}(\tilde{x}), x_f) \\ &< \theta(1 - \lambda)^{-1} + \varepsilon \end{aligned}$$

on the basis of inequality (II.2) in [1]. Hence, for all  $i \geq N$ ,  $\tilde{w}^{oi}(\tilde{x}) \in \Lambda(x_f, r_0 + \varepsilon)$ , and because  $\tilde{x}$  is any point of  $\mathcal{D}^n(\delta)$ , we get that, for every  $\varepsilon > 0$ ,  $\Lambda(x_f, r_0 + \varepsilon)$  is an absorbing set for  $\tilde{w}$ .

Now we prove that  $\Lambda(x_f, r_0 + \varepsilon)$  also owns the absorbing property for  $\varepsilon = 0$ . Let  $\Lambda^C(x_f, r_0) = \mathcal{D}^n(\delta) \setminus \Lambda(x_f, r_0)$ . Since  $\Lambda^C(x_f, r_0)$  is countable, the minimum  $\varepsilon = \min\{d(\tilde{y}, x_f) : \tilde{y} \in \Lambda^C(x_f, r_0)\}$  exists and  $\varepsilon > r_0$ . Therefore,  $\Lambda(x_f, r_0 + \varepsilon/2)$  does not include any point from  $\Lambda^C(x_f, r_0)$ , and thus  $\Lambda(x_f, r_0 + \varepsilon/2) = \Lambda(x_f, r_0)$ .

The last thing to we show is that  $\tilde{w}$  maps  $\Lambda(x_f, r_0)$  into itself. Let  $\tilde{y} \in \Lambda(x_f, r_0)$ . We have

$$\begin{aligned} d(\tilde{w}(\tilde{y}), x_f) &\leq d(\tilde{w}(\tilde{y}), w(\tilde{y})) + d(w(\tilde{y}), x_f) \\ &\leq \theta + \lambda d(\tilde{y}, x_f) \\ &\leq \theta(1 + \lambda(1 - \lambda)^{-1}) = \theta(1 - \lambda)^{-1} \end{aligned}$$

because  $d(\tilde{w}(\tilde{y}), w(\tilde{y})) \leq \theta$  by the definition of a  $\delta$ -roundoff of a mapping (Def. II.2 in [1]). Therefore,  $\tilde{w}(\tilde{y}) \in \Lambda(x_f, r_0)$ , which completes the proof.  $\square$

**Proof of Corollary II.8.** On the basis of Theorem II.7,  $\Lambda(x_f, r_0)$  is bounded and  $\tilde{w}$  maps it into itself, so from Theorem II.4 there exists  $\mathcal{M}[\tilde{w}, \Lambda(x_f, r_0)]$ . But, by Theorem II.7,  $\Lambda(x_f, r_0)$  is an absorbing set for  $\tilde{w}$  in  $\mathcal{D}^n(\delta)$ , and thus  $\mathcal{M}[\tilde{w}] = \mathcal{M}[\tilde{w}, \Lambda(x_f, r_0)]$  by Theorem II.5.  $\square$

**Proof of Theorem III.2.** (a) Suppose  $C$  is a closed class, so for every  $i \in \{1, \dots, N\}$ ,  $\tilde{w}_i(C) \subset C$ . Hence  $\bigcup_{i=1}^N \tilde{w}_i(C) \subset C$ . Now, suppose that  $\bigcup_{i=1}^N \tilde{w}_i(C) \subset C$ . The mappings  $w_i$  map  $C$  into itself, so for any  $\tilde{x} \in C$  and every finite sequence  $i_1, \dots, i_m$  of indices from  $\{1, \dots, N\}$ , we have  $w_{i_m} \circ \dots \circ w_{i_1}(\tilde{x}) \in C$ . Therefore, if  $\tilde{y} \notin C$ , then  $\tilde{y}$  is not accessible from  $C$ , and thus  $C$  is a closed class. (b) Let  $\tilde{x} \in C$ , where  $C$  is a communication class. Then  $\tilde{x}$  is accessible from any state in  $C$ , and hence there exist at least one  $\tilde{y} \in C$  and  $i \in \{1, \dots, N\}$  such that  $\tilde{w}_i(\tilde{y}) = \tilde{x}$ . Hence,  $\tilde{x} \in \tilde{w}_i(C)$ , and as a consequence  $\bigcup_{i=1}^N \tilde{w}_i(C) \supset C$ .  $\square$

**Lemma.** Let  $\{S; \tilde{w}_1, \dots, \tilde{w}_N; p_1, \dots, p_N\}$  be a DIFS. Suppose that for at least one DIFS mapping  $\tilde{w}_i$  there exists a minimal absorbing set  $\mathcal{M}[\tilde{w}_i, S]$ . Let  $\mathcal{A}_k$  be a recurrent communication class of the associated Markov chain  $\{\tilde{X}_k\}$ . If there exists  $\tilde{x} \in S$  such that both  $\tilde{x} \in \mathcal{A}_k$  and  $\tilde{x} \in \mathcal{B}[\mathcal{M}_j[\tilde{w}_i, S]]$ , then  $\mathcal{A}_k \supset \mathcal{M}_j[\tilde{w}_i, S]$ .

*Proof.* Suppose that  $\tilde{x} \in S$  satisfies the assumptions above. Since  $\tilde{x}$  is in the basin of attraction of  $\mathcal{M}_j[\tilde{w}_i, S]$ , by Def. II.4 there is  $m \in \mathbb{N}$  such that  $\tilde{w}_i^{om}(\tilde{x}) \in \mathcal{M}_j[\tilde{w}_i, S]$ . Moreover,  $\mathcal{M}_j[\tilde{w}_i, S]$  is a periodic orbit for  $w_i$  in  $S$  (Corollary II.3 in [1]). Therefore, writing  $p \in \mathbb{N}$  for the period of  $\mathcal{M}_j[\tilde{w}_i, S]$ , we

get that for any  $y \in \mathcal{M}_j[\tilde{w}_i, S]$ , there is a certain  $k < m + p$  such that

$$\Pr(\tilde{X}_k = \tilde{y} \mid \tilde{X}_0 = \tilde{x}) \geq \prod_{l=1}^k p_i(\tilde{w}_i^{ol}(\tilde{x})) > 0,$$

because  $p_i(\cdot)$  is strictly positive over  $S$ . Hence, every point of  $\mathcal{M}_j[\tilde{w}_i, S]$  is accessible from  $\tilde{x}$ . Since, by assumption,  $\tilde{x}$  is also in  $\mathcal{A}_k$  and the set is a closed class, as a result we get that  $\mathcal{A}_k \supset \mathcal{M}_j[\tilde{w}_i, S]$  as required.  $\square$

**Proof of Theorem III.3.** (a) Let  $\mathcal{A}_k \in \mathcal{A}_{\mathcal{F}}$  and let  $\mathcal{M}[\tilde{w}_i, S] \in \mathcal{M}_{\mathcal{F}}$ . The basins of attraction  $\mathcal{B}[\mathcal{M}_j[\tilde{w}_i, S]]$ ,  $j \in \{1, \dots, \mathcal{M}_{\#}[\tilde{w}_i, S]\}$  forms a countable partition of  $S$ . Therefore,  $\mathcal{A}_k$  is a subset of a union of a certain number of the basins,  $\mathcal{A}_k = \mathcal{A}_k \cap \bigcup_j \mathcal{B}[\mathcal{M}_j[\tilde{w}_i, S]]$ . On the basis of the above lemma, if  $\mathcal{A}_k \cap \mathcal{B}[\mathcal{M}_j[\tilde{w}_i, S]] \neq \emptyset$ , then  $\mathcal{A}_k \supset \mathcal{M}_j[\tilde{w}_i, S]$ . Hence, each set in  $\mathcal{A}_{\mathcal{F}}$  includes at least one component of the set  $\mathcal{M}[\tilde{w}_i, S]$ . Moreover, for any  $\mathcal{A}_m \in \mathcal{A}_{\mathcal{F}}$  such that  $\mathcal{A}_m \supset \mathcal{M}_j[\tilde{w}_i, S]$ , we have  $\mathcal{A}_k = \mathcal{A}_m$ , because the recurrent communication classes are disjoint. Hence, for any  $\mathcal{M}[\tilde{w}_i, S] \in \mathcal{M}_{\mathcal{F}}$ , the number of sets in  $\mathcal{A}_{\mathcal{F}}$  cannot exceed the number of the components of  $\mathcal{M}[\tilde{w}_i, S]$ , which completes this part of the proof.

(b) If  $\mathcal{A}_{\mathcal{F}}$  is empty, the conclusion of the theorem follows trivially. Assume that  $\mathcal{A}_{\mathcal{F}}$  is nonempty. The basins of attractions of the components of the set  $\mathcal{M}[\tilde{w}_i, S]$  constitute a countable partition of  $S$ , so by the above lemma, each  $\mathcal{A}_k \in \mathcal{A}_{\mathcal{F}}$  includes at least one of the components of  $\mathcal{M}[\tilde{w}_i, S]$ . But by the assumption of the theorem, all the components belong to the basin of attraction  $\mathcal{B}[\mathcal{M}_k[\tilde{w}_j, S]]$  and thus, again by the lemma above,  $\mathcal{M}_k[\tilde{w}_j, S]$  is a subset of every set  $\mathcal{A}_k \in \mathcal{A}_{\mathcal{F}}$ . Since  $\mathcal{A}_{\mathcal{F}}$  is a family of disjoint sets,  $\mathcal{A}_{\mathcal{F}}$  consists of a single set as required.  $\square$

**Proof of Theorem III.6.** First we show that the DIFS transformations map  $S$  into itself. Suppose that  $\tilde{x}$  is any point of  $S$ . We need to show that, for any  $i \in \{1, \dots, N\}$ ,  $\tilde{w}_i(\tilde{x}) \in S$ . Using the triangle inequality along with the Lipschitz continuity of  $w_i$ 's and the definition of a  $\delta$ -roundoff of a mapping (Def. II.2 in [1]), we get, for any  $i \in \{1, \dots, N\}$  and  $\varepsilon > 0$ , that

$$\begin{aligned} d(\tilde{w}_i(\tilde{x}), o) &\leq d(w_i(\tilde{x}), o) + d(\tilde{w}_i(\tilde{x}), w_i(\tilde{x})) \\ &\leq d(w_i(\tilde{x}), x_f^{(i)}) + d(x_f^{(i)}, o) + \theta \\ &\leq \lambda_{max} d(\tilde{x}, x_f^{(i)}) + r_{max} + \theta \\ &\leq \lambda_{max} (d(\tilde{x}, o) + d(x_f^{(i)}, o)) + r_{max} + \theta \\ &< \lambda_{max} (\alpha r_{max} + \theta(1 - \lambda_{max})^{-1} + \varepsilon + r_{max}) \\ &\quad + r_{max} + \theta \\ &= \alpha r_{max} + \lambda_{max} (\theta(1 - \lambda_{max})^{-1} + \varepsilon) \\ &< \alpha r_{max} + \theta(1 - \lambda_{max})^{-1} + \varepsilon \end{aligned}$$

as required.

Now we show that the Markov chain associated with the DIFS possesses a set  $\mathcal{A}$  of recurrent states. By Theorem III.2 (a) in [1], the set  $S$  is a closed class and, moreover,  $S$  is by definition finite, so due to the second part of Theorem

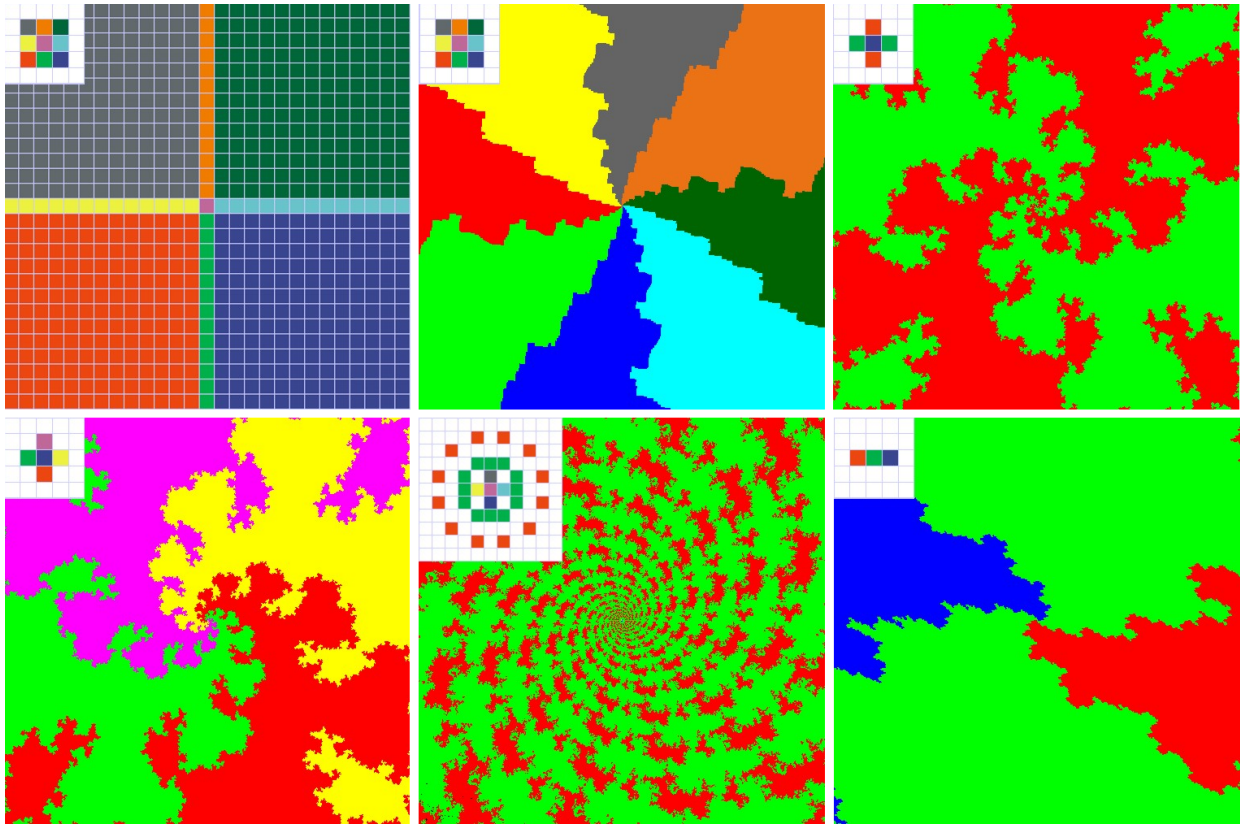


Fig. 1. Examples of minimal absorbing sets and their basins of attraction generated by  $\delta$ -roundoffs of two-dimensional linear contractions. From left to right, top to bottom, the first five pictures are derived from similarities with scaling factors and rotations respectively: 0.6 and  $0^\circ$ , 0.6 and  $5^\circ$ , 0.6 and  $150^\circ$ , 0.6 and  $30^\circ$ , 0.9 and  $30^\circ$ . The last picture is derived from a linear mapping specified by the matrix  $\begin{bmatrix} 0.5 & 0.3 \\ -0.1 & 0.4 \end{bmatrix}$ .

III.1 in [1], the Markov chain possesses at least one nonempty recurrent communication class within  $S$ , and hence  $\mathcal{A} \neq \emptyset$ .

To complete the proof we need to show that  $\mathcal{A} \subset S$ . By the assumption of the theorem,  $\tilde{w}_i$  are roundoffs of contractions, so by Corollary II.8 in [1], for each  $\tilde{w}_i$ , there is a finite minimal absorbing set  $\mathcal{M}[\tilde{w}_i]$  (with respect to the whole space  $\mathcal{D}^n(\delta)$ ). Now, fix some  $i \in \{1, \dots, N\}$  and observe that if  $\tilde{x} \in \mathcal{A}$ , then  $\tilde{x}$  is located in one of the basins of attraction of  $\mathcal{M}[\tilde{w}_i]$ . Therefore, for each  $\tilde{x} \in \mathcal{A}$ , there is a state  $\tilde{y} \in \mathcal{M}[\tilde{w}_i]$  accessible from  $\tilde{x}$ , and thus also  $\tilde{y} \in \mathcal{A}$ , because  $\mathcal{A}$  is a closed class. In addition,  $\mathcal{A}$  is composed of communication classes  $\mathcal{A}_k$ . From this we conclude that any set which is a closed class and, at the same time, includes  $\mathcal{M}[\tilde{w}_i]$  has to contain  $\mathcal{A}$  too. The set  $S$  is a closed class, so to finish the proof it suffices to show that  $S \supset \mathcal{M}[\tilde{w}_i]$ . On the basis of Theorem II.7 in [1], for every point  $\tilde{x} \in \mathcal{M}[\tilde{w}_i]$ ,  $d(\tilde{x}, x_f^{(i)}) \leq \theta(1 - \lambda_i)^{-1}$ , and hence for any  $\varepsilon > 0$ ,

$$\begin{aligned} d(\tilde{x}, o) &\leq d(x_f^{(i)}, o) + d(\tilde{x}, x_f^{(i)}) \\ &\leq r_{max} + \theta(1 - \lambda_{max})^{-1} \\ &< \alpha r_{max} + \theta(1 - \lambda_{max})^{-1} + \varepsilon, \end{aligned}$$

because  $\alpha \geq 1$ . This completes the proof.  $\square$

**Proof of Corollary III.7.** The only thing to show is that for every  $\tilde{X}_0 = \tilde{x} \in \mathcal{D}^n(\delta)$ ,  $\Pr(\exists i \in \mathbb{N} : X_i \notin \mathcal{T}) = 1$ . Using Theorem III.6 in [1], for any  $\tilde{x} \in \mathcal{T}$  one can construct a finite closed class  $S$  so that  $\tilde{x} \in S$  and  $\mathcal{A} \subset S$ . Therefore,

the "escape-from-transient-class" conclusion follows from the second part of Theorem III.1.  $\square$

**Proof of Lemma III.8.** The mutual existence of the orbits  $\{\tilde{x}_k = \tilde{w}_{i_k}(\tilde{x}_{k-1})\}$  and  $\{x_k = w_{i_k}(x_{k-1})\}$  of the chain  $\tilde{X}$  and  $X$ , respectively, is trivially provided by the strict positivity of probability functions  $p_i(\cdot)$  and probability weights  $q_i$ . Therefore, all we need to show is that all points of the orbits satisfy inequality (III.6) in [1]. The proof is by induction. For  $k = 1$ , we have

$$\begin{aligned} d(x_1, \tilde{x}_1) &\leq d(x_1, w_{i_1}(\tilde{x}_0)) + d(w_{i_1}(\tilde{x}_0), \tilde{x}_1) \\ &\leq d(w_{i_1}(\tilde{x}_0), \tilde{w}_{i_1}(\tilde{x}_0)) \leq \theta, \end{aligned}$$

where the last inequality follows from the definition of a  $\delta$ -roundoff of a mapping (Def. II.2 in [1]). Now assume that inequality III.6 in [1] is true for  $k$ . Then

$$\begin{aligned} d(x_{k+1}, \tilde{x}_{k+1}) &\leq d(x_{k+1}, w_{i_{k+1}}(\tilde{x}_k)) + d(w_{i_{k+1}}(\tilde{x}_k), \tilde{x}_{k+1}) \\ &\leq d(w_{i_{k+1}}(x_k), w_{i_{k+1}}(\tilde{x}_k)) \\ &\quad + d(w_{i_{k+1}}(\tilde{x}_k), \tilde{w}_{i_{k+1}}(\tilde{x}_k)) \\ &\leq \lambda_{max} d(x_k, \tilde{x}_k) + \theta \\ &\leq \lambda_{max} (1 - \lambda_{max})^{-1} \theta + \theta \\ &= \theta(1 - \lambda_{max})^{-1} \end{aligned}$$

as required.  $\square$

**Proof of Theorem III.9.** The Hausdorff distance between  $\mathcal{A}_k^+$  and  $A_\infty$  is  $h(\mathcal{A}_k^+, A_\infty) = \inf\{\varepsilon \geq 0 : A_\infty \subset$

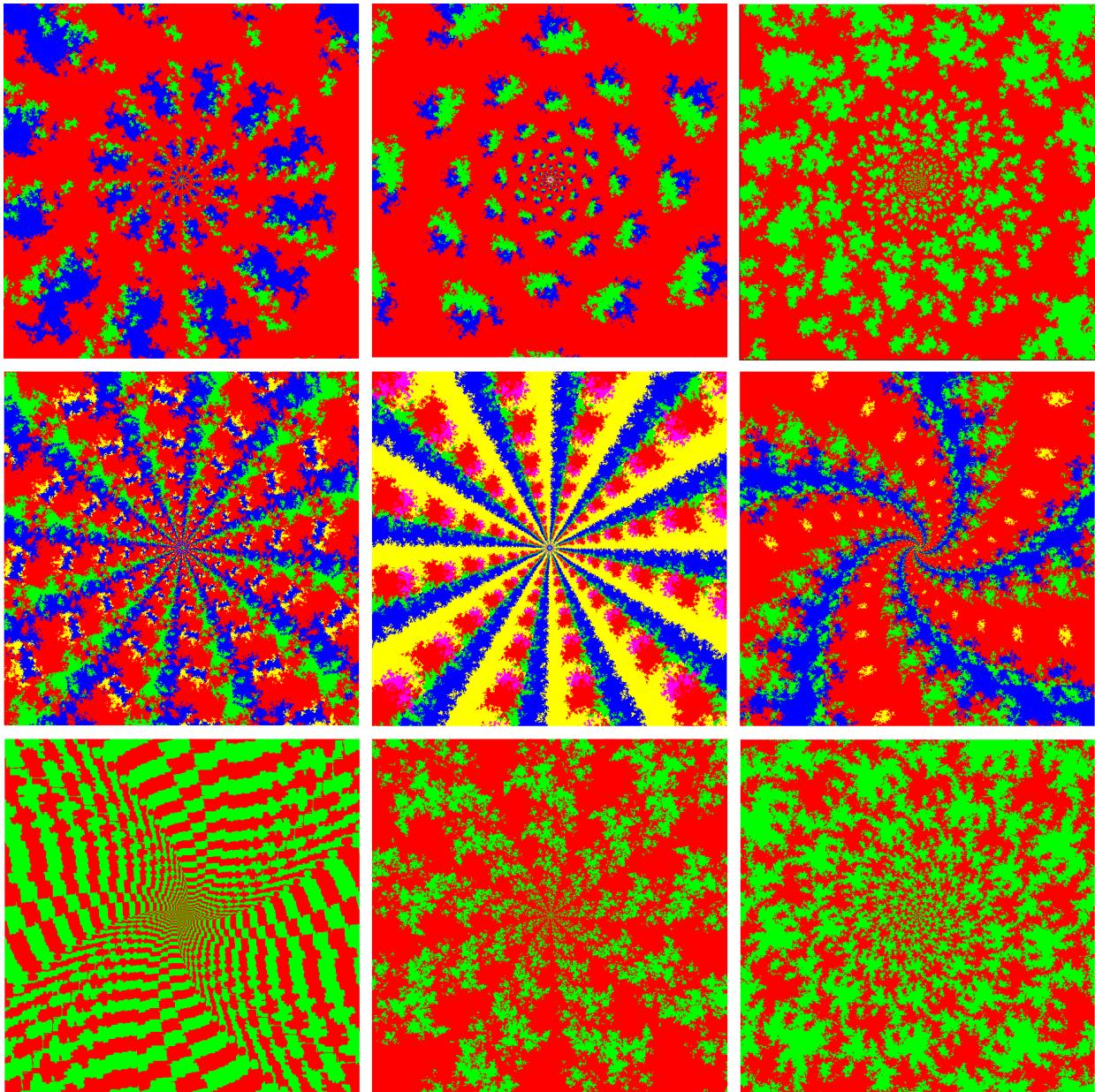


Fig. 2. Examples of basins of attractions generated by  $\delta$ -roundoffs of two-dimensional affine maps being contractions with respect to the "weighted" maximum metric  $d_\infty^{(p)}(x, y) = \max(p|x_1 - y_1|, |x_2 - y_2|)$ ,  $p > 1$ .

$N(\mathcal{A}_k^+, \varepsilon)$  and  $\mathcal{A}_k^+ \subset N(A_\infty, \varepsilon)$ , where  $N(A, \varepsilon) := \{x \in \mathbb{R}^n : d(x, a) < \varepsilon, a \in A\}$  denotes the (open)  $\varepsilon$ -neighbourhood of a set  $A$ . First we show that for any  $\varepsilon > 0$ ,  $A_\infty \subset N(\mathcal{A}_k^+, \theta(1 - \lambda_{max})^{-1} + \varepsilon)$ . We need to show that for any  $a \in A_\infty$ , there is a certain  $\tilde{x} \in \mathcal{A}_k^+$  such that  $d(a, \tilde{x}) < \theta(1 - \lambda_{max})^{-1} + \varepsilon$ . Let  $a$  be any point of  $A_\infty$ . The attractor is the support of the IFS invariant measure  $\pi$ , so for any  $\varepsilon > 0$ ,  $\pi(B(a, \varepsilon)) > 0$ . Moreover, by Elton's ergodic theorem (Eq. III.4 in [1])  $\pi$  is ergodic, and hence, for any initial point  $x_0 \in \mathbb{R}^n$ , almost every orbit  $\{x_i\}_{i=0}^\infty$  of the Markov chain generated by the IFS visits  $B(a, \varepsilon)$  infinitely often. Hence, for any  $x_0 \in \mathbb{R}^n$ , there is a finite sequence of indices  $i_m, \dots, i_1 \in \{1, \dots, N\}$  such that  $d(w_{i_m} \circ \dots \circ w_{i_1}(x_0), a) < \varepsilon$ . On that basis, putting  $x_0 \in \mathcal{A}_k^+$  and using the previous lemma,

we get that there is a finite sequence  $\mathbf{i} = (i_m, \dots, i_1)$  of indices such that

$$\begin{aligned} d(\tilde{w}_\mathbf{i}(x_0), a) &\leq d(\tilde{w}_\mathbf{i}(x_0), w_\mathbf{i}(x_0)) + d(w_\mathbf{i}(x_0), a) \\ &< \theta(1 - \lambda_{max})^{-1} + \varepsilon, \end{aligned}$$

where  $w_\mathbf{i}(\cdot) := w_{i_m} \circ \dots \circ w_{i_1}(\cdot)$ . But  $\mathcal{A}_k^+$  is a closed class, and thus  $\tilde{w}_\mathbf{i}(x_0) \in \mathcal{A}_k^+$ . Hence,  $A_\infty \subset N(\mathcal{A}_k^+, \theta(1 - \lambda_{max})^{-1} + \varepsilon)$  as required.

Now we show that for any  $\varepsilon > 0$ ,  $\mathcal{A}_k^+ \subset N(A_\infty, \theta(1 - \lambda_{max})^{-1} + \varepsilon)$ . Let  $\tilde{x}$  be any point of  $\mathcal{A}_k^+$ . Since  $\mathcal{A}_k^+$  is a recurrent class,  $\tilde{x}$  is a recurrent state and thus there is a finite sequence of indices  $\mathbf{i} = (i_m, \dots, i_1) \in \{1, \dots, N\}^m$  such that  $\tilde{w}_\mathbf{i}(\tilde{x}) = \tilde{x}$ , and hence for any  $j \in \mathbb{N}$ ,  $\tilde{w}_\mathbf{i}^{oj}(\tilde{x}) = \tilde{x}$ . Now let  $a \in A_\infty$ . Since the IFS maps  $w_i$  are contractions, we have,

for any  $j \in \mathbb{N}$ ,

$$d(w_1^{\circ j}(\tilde{x}), w_1^{\circ j}(a)) \leq \lambda_{max}^{m \cdot j} d(\tilde{x}, a)$$

and hence for any  $\varepsilon > 0$ , there is  $M \in \mathbb{N}$  such that

$$d(w_1^{\circ M}(\tilde{x}), w_1^{\circ M}(a)) \leq \lambda_{max}^M d(\tilde{x}, a) < \varepsilon,$$

because  $\lambda_{max} \in [0, 1)$ . On the basis of above, using the previous lemma we conclude that

$$\begin{aligned} d(\tilde{x}, w_1^{\circ M}(a)) &= d(\tilde{w}_1^{\circ M}(\tilde{x}), w_1^{\circ M}(a)) \\ &\leq d(\tilde{w}_1^{\circ M}(\tilde{x}), w_1^{\circ M}(\tilde{x})) + d(w_1^{\circ M}(\tilde{x}), w_1^{\circ M}(a)) \\ &< \theta(1 - \lambda_{max})^{-1} + \varepsilon. \end{aligned}$$

But  $w_1^{\circ M}(a) \in A_\infty$ , because  $w_i$ 's map  $A_\infty$  into itself. It follows that  $\mathcal{A}_k^+ \subset N(A_\infty, \theta(1 - \lambda_{max})^{-1} + \varepsilon)$  as required.

Since both  $A_\infty \subset N(\mathcal{A}_k^+, \theta(1 - \lambda_{max})^{-1} + \varepsilon)$  and  $\mathcal{A}_k^+ \subset N(A_\infty, \theta(1 - \lambda_{max})^{-1} + \varepsilon)$  for any  $\varepsilon > 0$ , we get that the infimum in the Hausdorff distance  $h(\mathcal{A}_k^+, A_\infty)$  is bounded from above by the value of  $\theta(1 - \lambda_{max})^{-1}$ . This completes the proof.  $\square$

**Proof of Corollary III.10.** By Theorem III.6 in [1], for any  $\delta > 0$ , the set of all recurrent states of the associated Markov chain is nonempty and finite (and thus compact), and so are the set's subsets  $\mathcal{A}_k^+(\delta)$ . Therefore, for any  $\delta > 0$ ,  $\mathcal{A}_k^+(\delta)$  is an element of  $\mathcal{H}(\mathbb{R}^n)$ , the family of all nonempty and compact subsets of  $\mathbb{R}^n$ . By a standard argument for hyperbolic IFSs,  $A_\infty \in \mathcal{H}(\mathbb{R}^n)$  too. Since  $\text{diam}_d(C_\delta) \rightarrow 0$  as  $\delta \rightarrow 0$ , and the Hausdorff distance  $h$  is a metric on  $\mathcal{H}(\mathbb{R}^n)$ , the conclusion follows from inequality III.7 in [1].  $\square$

**Proof of Theorem III.11.** In [2] Peruggia showed that a very similar conclusion holds under the assumption that the fixed point of one of the IFS mappings coincides with  $\mathbf{0} \in \mathcal{D}^2(\delta)$ , the zero vector of the discrete (pixel) space, which naturally stays intact while changing the value of the discretization parameter  $\delta$  (cf. [2], Theorem 4.38, pp. 129–131). Although our theorem does not impose such a restriction, the proof is founded on similar arguments as those used in the proof by Peruggia.

First, observe that the summation on the left hand side of Eq. III.8 in [1] can be restricted to the support of the measure  $\pi_k(\delta)$ ,  $\text{supp}(\pi_k(\delta)) = \mathcal{A}_k^+(\delta)$ , and, as we pointed out earlier, the corresponding Markov chain  $\{\tilde{X}_i^\delta(\tilde{x}_0) : \tilde{X}_0^\delta(\tilde{x}_0) = \tilde{x}_0 \in \mathcal{A}_k^+(\delta)\}$  generated by the DIFS (for fixed  $\delta$ ) is (Birkhoff's) ergodic on  $\mathcal{A}_k^+(\delta)$ . In addition, by Elton's ergodic theorem, the IFS invariant measure  $\pi_\infty$  is ergodic for the Markov chain  $\{X_i(x_0) : X_0 = x_0 \in \mathbb{R}^n\}$  generated by the IFS on  $\mathbb{R}^n$ . Putting these facts together, we get that for any  $\delta > 0$ , with probability one,

$$\begin{aligned} &\left| \sum_{\tilde{x} \in \mathcal{A}_k^+(\delta)} f(\tilde{x}) \pi_k(\delta)\{\tilde{x}\} - \int_{\mathbb{R}^n} f(x) d\pi_\infty \right| \\ &= \lim_{m \rightarrow \infty} \frac{1}{m} \left| \sum_{i=0}^{m-1} (f(\tilde{X}_i^\delta(\tilde{x}_0)) - f(X_i(\tilde{x}_0))) \right| \quad (\text{I.1}) \\ &\leq \lim_{m \rightarrow \infty} \frac{1}{m} \sum_{i=0}^{m-1} |f(\tilde{X}_i^\delta(\tilde{x}_0)) - f(X_i(\tilde{x}_0))| \end{aligned}$$

where  $\tilde{x}_0 \in \mathcal{A}_k^+(\delta)$ . But  $\tilde{X}_i^\delta(\tilde{x}_0) = \tilde{w}_{I_i}(\tilde{X}_{i-1}^\delta(\tilde{x}_0))$  and  $X_i(\tilde{x}_0) = w_{I_i}(X_{i-1}(\tilde{x}_0))$ , that is, both Markov chains are driven by the same sequence  $\{I_i\}_{i \in \mathbb{N}}$  of the i.i.d. random variables  $I_i$  distributed as  $[p_1, \dots, p_N]$ . Hence, from Lemma III.8 in [1],

$$d(X_i(\tilde{x}_0), \tilde{X}_i^\delta(\tilde{x}_0)) \leq \theta(1 - \lambda_{max})^{-1} \quad (\text{I.2})$$

for every  $i \in \mathbb{N}$ , where  $\lambda_{max}$  is the maximum contractivity factor of the IFS mappings  $w_i$ . Now crucial is the observation, which will be shown to be true at the end of the proof, that there exists a compact set  $E \subset \mathbb{R}^n$  such that, for every  $\delta \in (0, R)$ , where  $R > 0$  is a certain real number,  $E \supset \{X_i(\tilde{x}_0)\}$  and  $E \supset \{\tilde{X}_i^\delta(\tilde{x}_0)\}$  for any  $\tilde{x}_0 \in \mathcal{A}_k^+(\delta)$ , that is, none of the chains for  $\delta \in (0, R)$  moves out of  $E$ . Then, on the basis of the Heine–Cantor theorem,  $f$ 's are uniformly continuous on  $E$ . Since the right-hand side of inequality (I.2) converges to 0 as  $\delta \rightarrow 0$ , from this we conclude that for any  $\varepsilon > 0$ , there is  $\delta(\varepsilon) \in (0, R)$  such that for any  $\delta \in (0, \delta(\varepsilon))$ ,  $|f(\tilde{X}_i^\delta(\tilde{x}_0)) - f(X_i(\tilde{x}_0))| < \varepsilon$  for all  $i \in \mathbb{N}$ . Therefore,

$$\lim_{\delta \rightarrow 0} \left( \lim_{m \rightarrow \infty} \frac{1}{m} \sum_{i=0}^{m-1} |f(\tilde{X}_i^\delta(\tilde{x}_0)) - f(X_i(\tilde{x}_0))| \right) = 0$$

and hence, taking limits as  $\delta \rightarrow 0$  on both sides of inequality (I.1), we get

$$\lim_{\delta \rightarrow 0} \sum_{\tilde{x} \in \mathcal{A}_k^+(\delta)} f(\tilde{x}) \pi_k(\delta)\{\tilde{x}\} = \int_{\mathbb{R}^n} f(x) d\pi_\infty.$$

Since the sets  $\mathcal{A}_k^+(\delta)$  are the supports of the measures  $\pi_k(\delta)$ , the summation on the left-hand side of the above formula equals the summation over the whole space  $\mathcal{D}^n(\delta)$ , and thus we have arrived at the conclusion of the theorem.

The remaining thing to show is the existence of a compact set  $E$ , in which the Markov chains  $\{X_i(\tilde{x}_0)\}$  and  $\{\tilde{X}_i^\delta(\tilde{x}_0)\}$  reside for any  $\tilde{x}_0 \in \mathcal{A}_k^+(\delta)$  and  $\delta \in (0, R)$ , so as to assure the uniform continuity of the functions  $f$ . To this end, we can apply the following construction: Fix  $R > 0$  and observe that by inequality III.7 in [1], for any  $\delta \in (0, R)$ ,  $\mathcal{A}_k^+(\delta) \subset N(A_\infty, r_0)$ ,  $r_0 = \frac{1}{2} \text{diam}_d(C_R)(1 - \lambda_{max})^{-1}$ , and because  $A_\infty$  is compact, so is the closure  $\bar{N}(A_\infty, r_0)$  of the neighbourhood. Hence, for all  $\delta \in (0, R)$  and  $\tilde{x}_0 \in \mathcal{A}_k^+(\delta)$ , all Markov chains  $\{\tilde{X}_i^\delta(\tilde{x}_0)\}$  reside in  $\bar{N}(A_\infty, r_0)$  (because  $\mathcal{A}_k^+(\delta)$ 's are closed classes). Next we extend  $\bar{N}(A_\infty, r_0)$  so as to additionally encompass all Markov chains  $\{X_i(\tilde{x}_0)\}$  for  $\delta \in (0, R)$  and  $\tilde{x}_0 \in \mathcal{A}_k^+(\delta)$ . Due to inequality (I.2), it is easily done by doubling the radius of  $\bar{N}(A_\infty, r_0)$ , so the required compact set is  $\bar{N}(A_\infty, 2r_0)$ . This completes the proof.  $\square$

## II. MINIMAL ABSORBING SETS—EXAMPLES

In Fig. 1 we present examples of minimal absorbing sets and their basins of attractions generated by  $\delta$ -roundoffs of two-dimensional affine contractions with respect to the Euclidean metric. In turn, Fig. 2 shows some examples of basins of attractions of affine contractions with respect to the "weighted" maximum metric defined as  $d_\infty^{(p)}(x, y) := \max(p|x_1 - y_1|, |x_2 - y_2|)$ , with weight  $p > 1$  as a parameter. It is worth noting what intricate dynamics an iteration of a

single affine map can exhibit when quantized and viewed in a discrete space.

### III. DIFS STATIONARY PROBABILITY MEASURES—EXAMPLES

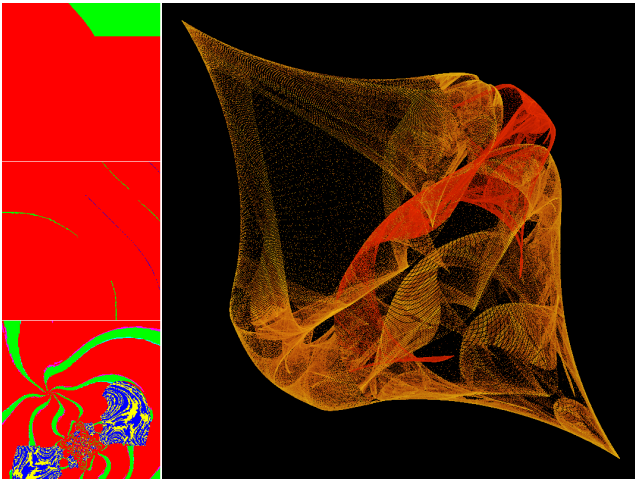


Fig. 3. An example of a DIFS stationary distribution visualized with the use of the random iteration algorithm. The DIFS is composed of three maps with minimal absorbing sets consisting of 2, 3 and 6 components, respectively. The basins of attractions of the maps are depicted on the left-hand side of the picture

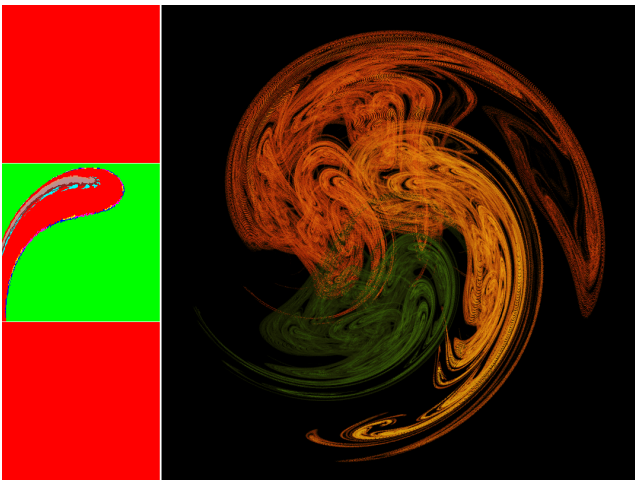


Fig. 4. Another example of a DIFS stationary distribution visualized with the use of the random iteration algorithm. The minimal absorbing sets consist of 1, 13 and 1 components, respectively

In Fig. 3–5 we present three examples of stationary probability measures of Discrete Iterated Function Systems defined on a squared subset  $C \subset \mathcal{D}^2(\delta)$  of  $1000 \times 1000$  resolution. Each DIFS consists of three mappings, which were constructed and stored in  $1000 \times 1000$  arrays. Each array  $W_i$  represented a single DIFS map  $\tilde{w}_i$  in the form of a pair of integer numbers as  $(W_i)_{kl} = \tilde{w}(k\delta, l\delta)/\delta$ . In other words, each  $W_i$  held information about a directed graph with vertices  $(k, l)$  being states of the associated Markov chain and edges generated by  $\tilde{w}_i$ , and thus the outdegree of each vertex being 1. For each

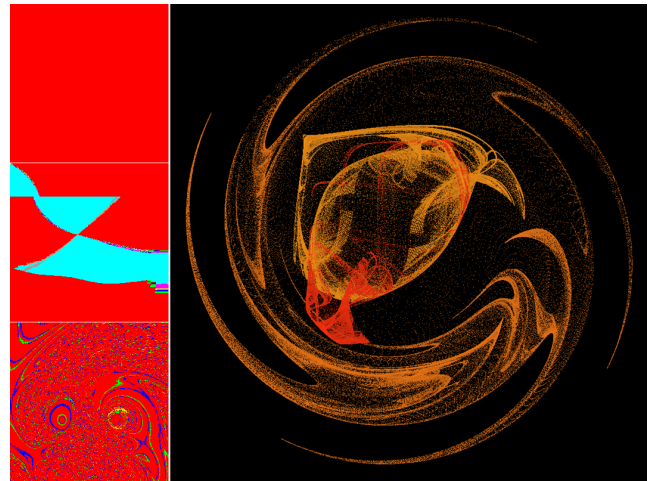


Fig. 5. One more example of a DIFS stationary distribution visualized with the aid of the random iteration algorithm. The minimal absorbing sets consist of 1, 11 and 5 components, respectively

DIFS, the arrays were initialized with three  $\delta$ -roundoffs of contractive similarities describing a Sierpiński's triangle, and then connections between the states were processed (changed) several times with the aid of nonlinear mappings (a collection of interesting transformations can be found in [3]), with care not to create a connection with a state lying outside  $C$ . Finally, the arrays (graphs) were additionally delicately "smoothed" with a  $3 \times 3$  Gaussian filter. Analogously, the distributions of the DIFS place-dependent probabilities were stored in a  $1000 \times 1000$  array  $D$  such that  $(D)_{kl}$  was the distribution at state  $(k, l)$ . The array was initiated with uniform distributions, which were then perturbed with a nonlinear mapping. The renderings were obtained by means of the random iteration algorithm that generated an orbit in subset  $C$  with the aid of the arrays  $W_i$  and  $D$ , so that given state  $(k, l)$ , the next state was determined as  $(W_i)_{kl}$  with  $i$  drawn from the distribution  $(D)_{kl}$ . Since  $W_i$ 's and  $D$  function as lookup tables, the orbit was generated in an extremely efficient manner, based only on fetches from the arrays and a pseudorandom generator. In the images the value of the measure of a state is interpreted as brightness, and the coloring reflects participation of a DIFS map in conveying a measure to a state, according to the formula  $c_{new} = (c_{old} + c_i)/2$ , where  $c_{new}$  and  $c_{old}$  are a new color and, respectively, a current color assigned to a visited state (with  $c_{old}$  initialized with black at the beginning of the algorithm), and  $c_i$  is a color assigned to the  $i$ th DIFS map.

### REFERENCES

- [1] T. Martyn, "A computer scientist's perspective on approximation of IFS invariant sets and measures with the random iteration algorithm", *International Journal of Electronics and Telecommunications*, vol. 70, no. 4, pp. 1113–1123, 2024. <https://doi.org/10.24425/ijet.2024.152514>
- [2] M. Peruggia, *Discrete Iterated Function Systems*. A. K. Peters Wellesley MA, 1993. <https://doi.org/10.5402/2012/825782>
- [3] O. S. Lawlor OS, GPU-accelerated rendering of unbounded nonlinear iterated function system fixed points, *ISRN Computer Graphics*, vol. 2012:1–17, 2012.