

# Linear IFSs consisting of stochastic matrices

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# AIMD (additive increase multiplicative decrease)

[1] Corless, King, Shorten, Wirth (2016)

Model of TCP: user  $j = 1, \dots, d$ ,  $d \geq 2$ , demands access to the internet,  
 $0 < \gamma_j < 1$  – coefficient of increase of  $j$ 's share,  $\sum_{j=1}^d \gamma_j = 1$ ,  
 $0 \leq \beta_j \leq 1$  – coefficient of decrease of  $j$ 's share.

$$A = \begin{bmatrix} \beta_1 & \dots & 0 \\ 0 & \dots & \beta_d \end{bmatrix} + \begin{bmatrix} \gamma_1 \\ \vdots \\ \gamma_d \end{bmatrix} [1 - \beta_1 \dots 1 - \beta_d],$$

$w_{n+1} = Aw_n$  — evolution of share at  $n$ th capacity event,  
 $c$  — capacity of connection,  
 $w_n \in H_c = \{u \in \mathbb{R}^d : \sum_{i=1}^d u_i = c\}$ .

# AIMD matrix

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$\beta_j \in [0, 1]$ ,  $\gamma_j \in (0, 1)$ ,  $\sum_{j=1}^d \gamma_j = 1$ .

[1, Lemma 3.5 p.33]:

If  $\max_{i=1, \dots, d} \beta_i < 1$ , then  $A$  is Banach contractive on

$\Delta_c = \{u \in \mathbb{R}^d : \sum_{i=1}^d u_i = c, u_i \geq 0\} \subseteq H_c$  with respect to  $\|u\|_1 = \sum_{i=1}^d |u_i|$ .

QUESTION:

What about  $\#\{i = 1, \dots, d : \beta_i = 1\} = 1$ ?

# Kantrowitz–Neumann criterion (2014) for contractivity

[2, Proposition 3.2] after improvement:

Let  $A = (a_{ij})_{i,j \in \{1, \dots, d\}}$  be a column-stochastic matrix, i.e.,  $a_{ij} \in [0, 1]$ ,  $\sum_{i=1}^d a_{ij} = 1$ ,  $i, j \in \{1, \dots, d\}$ . Then the following are equivalent:

- (i)  $A$  is a contraction w.r.t.  $\|\cdot\|_1$  on *some*  $c$ -hyperplane  $H_c$ ;
- (ii)  $A$  is a contraction w.r.t.  $\|\cdot\|_1$  on *every*  $H_c$ ;
- (iii) (positivity)  $A^T \cdot A$  has positive all entries.

# Kantrowitz–Neumann criterion (2014) for AIMD

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**Corollary:** If an AIMD matrix

$$A = \begin{bmatrix} \beta_1 & \dots & 0 \\ 0 & \dots & \beta_d \end{bmatrix} + \begin{bmatrix} \gamma_1 \\ \vdots \\ \gamma_d \end{bmatrix} [1 - \beta_1 \dots 1 - \beta_d],$$

$\beta_j \in [0, 1]$ ,  $\gamma_j \in (0, 1)$ ,  $\sum_{j=1}^d \gamma_j = 1$ , satisfies

$$\#\{i = 1, \dots, d : \beta_i = 1\} \leq 1$$

(equivalently,  $A$  has at most one trivial = unit vector column),  
then  $A$  is contractive on  $(H_c, \|\cdot\|_1)$ .

# Switched AIMD matrices $\Rightarrow$ IFS

Assume:

$A_i$  – column-stochastic matrices,

$A_i^T \cdot A_i$  has positive all entries,  $i = 1, \dots, N$ .

Then

for every  $c \in \mathbb{R}$ ,

$\mathcal{F}_c = (H_c, f_i(u) = A_i \cdot u : i = 1, \dots, N)$  is contractive w.r.t.  $\|\cdot\|_1$ .

In particular, the above holds when each matrix  $A_i$  is an AIMD matrix with at most one trivial column (unit vector).

# Switched AIMD matrices $\Rightarrow$ IFS attractor

(a)  $\mathcal{F}_c$  has an attractor  $\mathbb{A}_*(c)$ , i.e.,  $\forall$  bounded  $\emptyset \neq S \subseteq H_c$ ,

$$F^n(S) := \overline{\bigcup_{(i_1, \dots, i_n) \in \{1, \dots, N\}^n} f_{i_n} \circ \dots \circ f_{i_1}(S)} \xrightarrow[n \rightarrow \infty]{\text{Hausdorff}} \mathbb{A}_*(c);$$

$$\mathbb{A}_*(c) \equiv \mathbb{A}_*(1) \quad \forall c \neq 0;$$

(b)  $(\mathcal{F}_c, (p_i)_{i=1}^N)$  has an attractive invariant distribution  $\mu_*(c)$ , i.e.,

$$M^n(\mu) := \sum_{(i_1, \dots, i_n) \in \{1, \dots, N\}^n} p_{i_n} \cdot \dots \cdot p_{i_1} \cdot (\mu \circ f_{i_1}^{-1} \circ \dots \circ f_{i_n}^{-1}) \xrightarrow[n \rightarrow \infty]{W^*} \mu_*(c)$$

$$\forall \text{ distrib. } \mu \text{ on } H_c; \quad \sum_{i=1}^N p_i = 1, p_i > 0; \quad \text{supp } \mu_*(c) = \mathbb{A}_*(c);$$

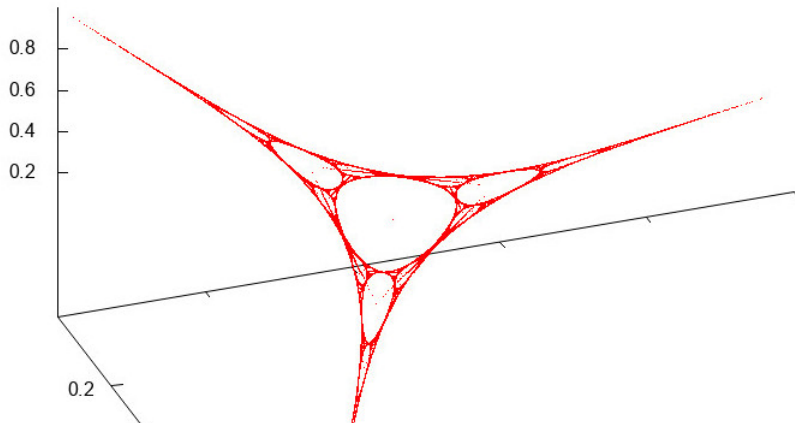
(c) (chaos game)  $\mathbb{A}_*(c) = \bigcap_{n=0}^{\infty} \overline{\{u_m : m \geq n\}}$ , where

$$\begin{cases} u_n := f_{i_n}(u_{n-1}), n \geq 1, u_0 \in H_c, \\ \{1, \dots, N\}^{\infty} \ni (i_n)_{n=1}^{\infty} - \text{disjunctive (i.e., contains all finite words).} \end{cases}$$

# AIMD triangle

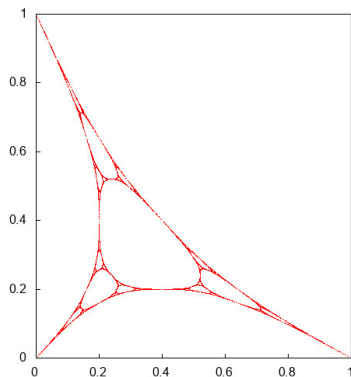
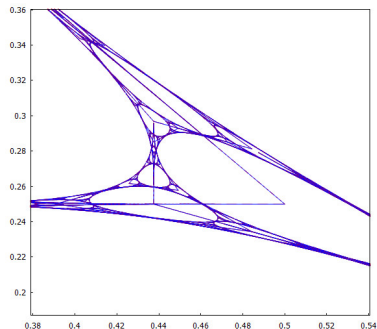
$\mathcal{F} = (\mathbb{R}^3; f_i(u) = A_i \cdot u : i = 1, 2, 3)$  contractive on  $H_c$  but not on  $\mathbb{R}^3$ .

$$A_1 = \begin{pmatrix} 1 & \frac{1}{4} & \frac{1}{4} \\ 0 & \frac{1}{2} & \frac{1}{4} \\ 0 & \frac{1}{4} & \frac{1}{2} \end{pmatrix}, A_2 = \begin{pmatrix} \frac{1}{2} & 0 & \frac{1}{4} \\ \frac{1}{4} & 1 & \frac{1}{4} \\ \frac{1}{4} & 0 & \frac{1}{2} \end{pmatrix}, A_3 = \begin{pmatrix} \frac{1}{2} & \frac{1}{4} & 0 \\ \frac{1}{4} & \frac{1}{2} & 0 \\ \frac{1}{4} & \frac{1}{4} & 1 \end{pmatrix}.$$





# AIMD triangle vs Kigami triangle

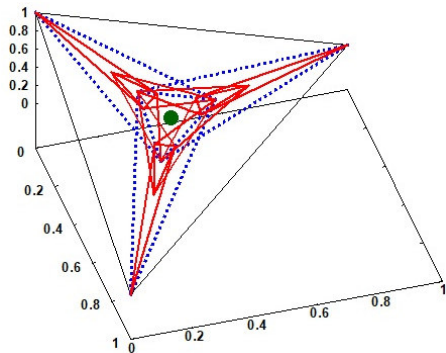
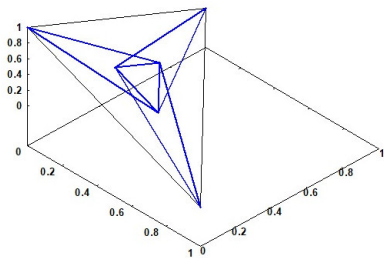


$$\mathcal{F} = (\mathbb{R}^2; f_i : i = 1, 2, 3) \quad f_1 \begin{pmatrix} x \\ y \end{pmatrix} = \frac{1}{5} \cdot \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

$$f_2 \begin{pmatrix} x \\ y \end{pmatrix} = \frac{1}{5} \cdot \left( \begin{pmatrix} 3 & 0 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} 2 \\ 1 \end{pmatrix} \right)$$

$$f_3 \begin{pmatrix} x \\ y \end{pmatrix} = \frac{1}{5} \cdot \left( \begin{pmatrix} 1 & -1 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right)$$

# AIMD triangle: analysis



$$\Delta_c = \text{conv}\{(c, 0, 0), (0, c, 0), (0, 0, c)\},$$

$$\mathcal{L}_{H_c}(\mathbb{A}_*(c)) \leq \mathcal{L}_{H_c}(F^n(\Delta_c)) \leq 3^n \cdot \left(\frac{3}{16}\right)^n \mathcal{L}_{H_c}(\Delta_c) \xrightarrow{n \rightarrow \infty} 0.$$

# Kantrowitz–Neumann criterion (2014) for ergodicity

[2, Theorem 3.3] & improved [2, Proposition 3.2]

Let  $A = (a_{ij})_{i,j \in \{1, \dots, d\}}$  be a column-stochastic  $d \times d$ -matrix. Then the following are equivalent:

(a)  $A$  is ergodic, i.e.,  $\exists!_{u_* \in \Delta_1} \forall_{u \in \Delta_1} A^k u \xrightarrow[k \rightarrow \infty]{} u_*$ ;

(b)  $\exists_{p \geq 1} A^p$  contains a strictly positive row;

(c)  $\exists_{p \geq 1} B := A^p$  is scrambling, i.e.,

$$\forall_{k, l \in \{1, \dots, d\}} \exists_{j \in \{1, \dots, d\}} b_{jk}, b_{jl} > 0;$$

(d)  $\exists_{p \geq 1} B := A^p$  is an Edelstein contraction on  $\Delta_1$  w.r.t.  $\|\cdot\|_1$ , i.e.,  $\|Bu - Bv\|_1 < \|u - v\|_1$  for probability vectors  $u \neq v \in \Delta_1$ ;

(e)  $A$  is an eventual contraction on  $(H_1, \|\cdot\|_1) \supseteq \Delta_1$ , i.e.,  $\exists_{p \geq 1} A^p$  is a Banach contraction on  $H_1$ ;

(f)  $\exists_{p \geq 1} (A^p)^T \cdot A^p \gg 0$ .

Remark: If the above has simple proof, then the classical criterion for ergodicity of Markov matrices easily follows from metric fixed point theory.

# Wielandt-type matrix

$$W_d := \begin{pmatrix} 0 & \alpha & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ & \dots & & \dots & \\ 0 & 0 & 0 & \dots & 1 \\ 1 & 1 - \alpha & 0 & \dots & 0 \end{pmatrix}, \alpha \in (0, 1).$$

$W_{10}^p$	(b)= has positive row	$\gg 0$	(d)= is contractive
$p$	73	82	41

- [K & N & Ransford 2011]  $A$  satisfies (b)  $\Rightarrow A^p$  contains positive row for  $p = d^2 - 3d + 3$ ; optimal when  $A = W_d$ ;
- [Horn & Johnson "Matrix Analysis"]  $W_d^p \gg 0 \Leftrightarrow p \geq d^2 - 2d + 2$ ;
- $A$  satisfies (d)  $\Rightarrow A^p$  contractive for  $p = d^2 - 3d + 3$ ;  
UNKNOWN OPTIMAL  $p$ ;
- $d \leq 10 \Rightarrow p = \lfloor \frac{1}{2} \cdot (d^2 - 2d + 2) \rfloor$  is optimal for contractivity of  $W_d^p$ .

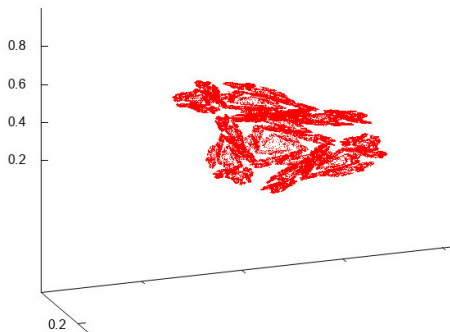
# Eventually contractive IFS. Example 1

$\mathcal{F} = (\mathbb{R}^3; f_i(u) = A_i \cdot u : i = 1, 2), A_i = W_3(\alpha = i/3),$

$$A_1 = \begin{pmatrix} 0 & \frac{1}{3} & 0 \\ 0 & 0 & 1 \\ 1 & \frac{2}{3} & 0 \end{pmatrix}, \quad A_2 = \begin{pmatrix} 0 & \frac{2}{3} & 0 \\ 0 & 0 & 1 \\ 1 & \frac{1}{3} & 0 \end{pmatrix}.$$

$\mathcal{F}^2 = (\mathbb{R}^3; f_i \circ f_j : i, j \in \{1, 2\})$  is  $\ell^1$ -contractive on  $H_c$  (by the KN positivity condition:  $(A_i A_j)^T \cdot A_i A_j \gg 0$ )

The attractor of  $\mathcal{F}^2|_{H_c}$  and  $\mathcal{F}|_{H_c}$



## Eventually contractive IFS. Example 2



$$\mathcal{F} = (\mathbb{R}^3; f_i(u) = A_i \cdot u : i = 1, 2),$$

$$A_1 = \begin{pmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \end{pmatrix}, \quad A_2 = A_1^T = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ 1 & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \end{pmatrix}.$$

$A_i$  are eventually contractive on  $H_c$  (KN criterion:  $(A_i^2)^T \cdot A_i^2 \gg 0$ ), but  $A_2 \cdot A_1$  is not.

$\mathcal{F}$  has no (local) attractor on  $H_c$ . (Careful verification by hand. Hint: some lines always stick out of the candidate for a local attractor).

## THANK YOU

-  M. Corless, C. King, R. Shorten, F. Wirth, *AIMD Dynamics and Distributed Resource Allocation*, SIAM, Philadelphia, 2016.
-  R. Kantrowitz, M.M. Neumann, *A fixed point approach to the steady state for stochastic matrices*, Rocky Mt. J. Math. **44** (2014), no. 4, 1243-1250.