

MIT Open Access Articles

Generalization of the Multiplicative and Additive Compounds of Square Matrices and Contraction Theory in the Hausdorff Dimension

The MIT Faculty has made this article openly available. *Please share* how this access benefits you. Your story matters.

Citation: C. Wu, R. Pines, M. Margaliot and J. -J. Slotine, "Generalization of the Multiplicative and Additive Compounds of Square Matrices and Contraction Theory in the Hausdorff Dimension," in IEEE Transactions on Automatic Control, vol. 67, no. 9, pp. 4629-4644, Sept. 2022, doi: 10.1109/TAC.2022.3162547.

As Published: 10.1109/tac.2022.3162547

Publisher: Institute of Electrical and Electronics Engineers

Persistent URL: https://hdl.handle.net/1721.1/155762

Version: Author's final manuscript: final author's manuscript post peer review, without publisher's formatting or copy editing

Terms of use: Creative Commons Attribution-Noncommercial-ShareAlike



Generalization of the multiplicative and additive compounds of square matrices and contraction in the Hausdorff dimension

Chengshuai Wu, Raz Pines, Michael Margaliot, and Jean-Jacques Slotine

December 29, 2020

Abstract—The k multiplicative and k additive compounds of a matrix play an important role in geometry, multi-linear algebra, the asymptotic analysis of nonlinear dynamical systems, and in bounding the Hausdorff dimension of fractal sets. These compounds are defined for integer values of k. Here, we introduce generalizations called the α multiplicative and α additive compounds of a square matrix, with α real. We study the properties of these new compounds and demonstrate an application in the context of the Douady and Oesterlé Theorem. This leads to a generalization of contracting systems to α contracting systems, with α real. Roughly speaking, the dynamics of such systems contracts any set with Hausdorff dimension larger than α . For $\alpha = 1$ they reduce to standard contracting systems.

Index Terms—Multiplicative compound matrix, additive compound matrix, fractal sets, contraction theory, nonlinear systems.

I. INTRODUCTION

Consider a matrix $A \in \mathbb{C}^{n \times m}$, and fix an integer $k \in \{1, \ldots, \min\{m, n\}\}$. The k multiplicative compound matrix of A, denoted $A^{(k)}$, is the $\binom{n}{k} \times \binom{m}{k}$ matrix that includes all the minors of order k of A organized in a lexicographic order. For example, if n = 3, m = 2, and k = 2, then $A^{(2)} \in \mathbb{R}^{3 \times 1}$ and is given by

$$A^{(2)} = \begin{bmatrix} A(\{1,2\}|\{1,2\}) \\ A(\{1,3\}|\{1,2\}) \\ A(\{2,3\}|\{1,2\}) \end{bmatrix},$$

where $A(\alpha|\beta)$ denotes the minor of A obtained by taking the rows indexed by α and the columns indexed by β . Note that this implies that $A^{(1)} = A$, and if m = n then $A^{(n)} = \det(A)$.

The Cauchy-Binet formula [9] asserts that for any $A \in \mathbb{C}^{n \times m}$, $B \in \mathbb{C}^{m \times r}$, and any $k \in \{1, \dots, \min\{n, m, r\}\}$,

$$(AB)^{(k)} = A^{(k)}B^{(k)}.$$
 (1)

This justifies the term multiplicative compound. For example, if A, B are $n \times n$ then (1) with k = n reduces to the familiar formula $\det(AB) = \det(A) \det(B)$.

The research of MM is supported in part by a research grant from the Israel Science Foundation.

C. Wu and R. Pines are with the School of Electrical Engineering, Tel Aviv University, Tel-Aviv 69978, Israel.

M. Margaliot (Corresponding Author) is with the School of Electrical Engineering, and the Sagol School of Neuroscience, Tel-Aviv University, Tel-Aviv 69978, Israel. E-mail: michaelm@tauex.tau.ac.il

J.-J. Slotine is with the Department of Mechanical Engineering and the Department of Brain and Cognitive Sciences, Massachusetts Institute of Technology, Cambridge, Massachusetts, USA. E-mail: jjs@mit.edu

When n = m, i.e. A is a square matrix, the k additive compound matrix of A is defined by

$$A^{[k]} := \frac{d}{d\varepsilon} (I + \varepsilon A)^{(k)}|_{\varepsilon = 0}.$$
 (2)

This implies that

$$(I + \varepsilon A)^{(k)} = I + \varepsilon A^{[k]} + o(\varepsilon), \tag{3}$$

i.e. $A^{[k]}$ is the first-order term in the Taylor series of $(I + \varepsilon A)^{(k)}$. For example, if $A \in \mathbb{C}^{n \times n}$ is diagonal, denoted $A = \text{diag}(\lambda_1, \ldots, \lambda_n)$, and k = 2 then

$$(I + \varepsilon A)^{(2)}$$

= $(\operatorname{diag}(1 + \varepsilon \lambda_1, \dots, 1 + \varepsilon \lambda_n))^{(2)}$
= $\operatorname{diag}((1 + \varepsilon \lambda_1)(1 + \varepsilon \lambda_2), \dots, (1 + \varepsilon \lambda_{n-1})(1 + \varepsilon \lambda_n))$
= $I + \varepsilon \operatorname{diag}(\lambda_1 + \lambda_2, \dots, \lambda_{n-1} + \lambda_n) + o(\varepsilon),$

so

$$A^{[2]} = \operatorname{diag}(\lambda_1 + \lambda_2, \dots, \lambda_{n-1} + \lambda_n)$$

Note that every eigenvalue of $A^{[2]}$ is the sum of two eigenvalues of A.

It can be shown [32] that (1) and (3) imply that

$$(A+B)^{[k]} = A^{[k]} + B^{[k]}, (4)$$

for any $A, B \in \mathbb{C}^{n \times n}$. This justifies the term additive compound.

Compound matrices have found numerous applications in multi-linear algebra, geometry, and dynamical systems theory. We quickly review some examples.

The exterior product (or wedge product) of vectors generalizes the notions of unsigned area [volume] in dimension 2 [3] to an arbitrary dimension k [52]. For k vectors $u^1, \ldots, u^k \in \mathbb{R}^n$, with $k \leq n$, the wedge product, denoted $u^1 \wedge \cdots \wedge u^k$ or $\wedge_{i=1}^k u^i$, can be defined using the k multiplicative compound as

$$u^i := \begin{bmatrix} u^1 & u^2 & \dots & u^k \end{bmatrix}^{(k)}.$$
 (5)

Note that this has dimensions $\binom{n}{k} \times \binom{k}{k} = \binom{n}{k} \times 1$, i.e. it is a vector in $\mathbb{R}^{\binom{n}{k}}$. The magnitude of $u^1 \wedge \cdots \wedge u^k$, i.e., $|u^1 \wedge \cdots \wedge u^k|$, can be interpreted as the hyper volume of the k-dimensional parallelotope with edges u^1, \ldots, u^k . For

example, if $a, b \in \mathbb{R}^3$ then

$$a \wedge b = \begin{bmatrix} a & b \end{bmatrix}^{(2)} = \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}^{(2)} = \begin{bmatrix} a_1b_2 - b_1a_2 \\ a_1b_3 - b_1a_3 \\ a_2b_3 - b_2a_3 \end{bmatrix}$$

and the entries here are (up to a sign change) the entries of the cross product $a \times b$. Recall that the magnitude of $a \times b$ can be interpreted as the positive area of the parallelogram having a and b as sides.

Recall that $A \in \mathbb{R}^{n \times m}$ is called totally non-negative [totally positive] if all its minors are non-negative [positive]. These matrices have found numerous applications [9], [34]. Clearly, $A \in \mathbb{R}^{n \times m}$ is totally non-negative [totally positive] if and only if the multiplicative compound matrices $A^{(1)}, A^{(2)}, \ldots, A^{(\min\{n,m\})}$ all have non-negative [positive] entries. Thus, every $A^{(i)}$ can be analyzed using the Perron-Frobenius theory of matrices with non-negative entries [15]. This simple fact has important applications in the analysis of totally non-negative matrices (see, e.g., [11], [2]).

We now describe some applications of compound matrices in dynamical systems theory. In this context, the relevant case is square matrices. Suppose that X(t) is the solution of the linear matrix differential equation

$$\frac{d}{dt}X(t) = A(t)X(t), \quad X(t_0) = X_0,$$
 (6)

with $A : \mathbb{R} \to \mathbb{R}^{n \times n}$ continuous. Then

$$\frac{d}{dt}X^{(k)}(t) = A^{[k]}(t)X^{(k)}(t), \quad X^{(k)}(t_0) = (X_0)^{(k)}, \quad (7)$$

for any $k \in \{1, ..., n\}$. In other words, all the k minors of X, stacked in the matrix $X^{(k)}$, also follow a linear dynamics, with the matrix $A^{[k]}$ (see, e.g., [32]). Note that if A is time-invariant and $t_0 = 0$ then the solution of (6) is $X(t) = \exp(At)X_0$ so $(X(t))^{(k)} = (\exp(At))^{(k)} (X_0)^{(k)}$ and combining this with (7) gives

$$(\exp(At))^{(k)} = \exp(A^{[k]}t), \text{ for all } t \in \mathbb{R}.$$
 (8)

Eq. (7) has important applications in the analysis of timevarying nonlinear dynamical systems in the form

$$\dot{x}(t) = f(t, x(t)), \tag{9}$$

where $f : \mathbb{R} \times \mathbb{R}^n \to \mathbb{R}^n$ is C^1 . Indeed, let x(t, a) denote the solution at time $t \ge 0$ of (9) with x(0) = a, and let

$$J_f(t,x) := \frac{\partial}{\partial x} f(t,x) \tag{10}$$

denote the Jacobian of the vector field f. Then the variational equation associated with (9) along the trajectory x(t, a) is

$$\dot{y}(t) = J_f(t, x(t, a))y(t).$$
 (11)

Analysis of this linear time-varying equation plays an important role in the asymptotic analysis of (9). Combining this with (7) has far-reaching applications in the theory of nonlinear dynamical systems [32]. Recent applications include:

• totally positive differential systems [41], [30], that is, systems where the transition matrix corresponding to the variational equation (11) is TP (see also [6]);

- k-cooperative dynamical systems [49], that is, systems where $J_f^{[k]}$ is a Metzler matrix;
- *k*-order contracting systems [53], that is, systems where J^[k]_f is infinitesimally contracting;
 the notion of a 1
- the notion of a discrete-time k-diagonally stable dynamical system, that is, a system of the form x(k +1) = Ax(k), and there exists a positive-definite diagonal matrix D, such that $(A^{(k)})^T D A^{(k)} - D$ is negativedefinite [54].

Since the k multiplicative compound is based on collecting all the $k \times k$ minors of a matrix, it is naturally defined only for integer values of k.

Here, we introduce a generalization of the k multiplicative and k additive compound of a square matrix, called the α multiplicative compound and α additive compound, where $\alpha > 1$ is a *real* number. For $k < \alpha < k + 1$ the α compounds may be interpreted as a weighted interpolation of the k and k+1compounds. When α is an integer this (almost) reduces to the standard k compounds. This generalization is motivated by the Hausdorff dimension of a set and, in particular, the seminal Douady and Oesterlé Theorem [8] that provides an upper bound for the Hausdorff dimension of a set that is negatively invariant under a C^1 mapping. As an application, we show that the α compounds can be used to provide elegant and intuitive expressions for the basic terms that appear in this theorem. Furthermore, this naturally leads to the new notion of α contracting systems, with α real, which generalizes the notion of k-order contracting systems with k an integer [19], [25], [53]. We analyze the properties of α contracting systems and demonstrate their applications. Our results show that if an *n*-dimensional dynamical system contracts *n*-dimensional volumes then there exists a minimal *real* value $\alpha^* \in [1, n]$ such that the system is α contracting for any $\alpha > \alpha^*$. Roughly speaking, an α^* contracting system contracts any set with a Hausdorff dimension larger than α^* . This generates (in a given metric) a continuum of contraction instead of the standard binary view, namely, that a system is either contracting or not contracting.

The remainder of this paper is organized as follows. The next section reviews several known definitions and results that will be used later on. Section III describes our main results. These include the definitions of the α multiplicative compound and α additive compound of a matrix, and analysis of the properties of these compounds. Section IV describes an application of these new notions and introduces α contracting systems. The last section concludes and describes several directions for future research. We focus here on the theory of the generalized compounds and α contracting systems, leaving applications to a sequel paper.

II. PRELIMINARIES

To make this paper more self-contained, we briefly review several topics that are needed to define, analyze and apply the α compounds, and α contracting systems. We begin with reviewing the Hausdorff dimension of a set, following [44], [35], [17].

A. Hausdorff dimension

Let K be a set in \mathbb{R}^n . For $d, \epsilon > 0$, the d-measured volume of ϵ -coverings of K is:

$$\zeta(K, \epsilon, d) := \inf \{ \sum_{i} r_i^d : \text{there exists a countable} \\ \text{cover of } K \text{ by balls with radii } r_i \leq \epsilon \}.$$

Note that the covering may include balls of different sizes, but all are bounded by ϵ . Note also that if K is compact then it would suffice to use finite coverings, since every open cover of K has a finite subcover.

By definition, $\zeta(K, \epsilon, d)$ is non-increasing in ϵ . The Hausdorff d-measure of K is

$$m(K,d) := \lim_{\epsilon \downarrow 0} \zeta(K,\epsilon,d),$$

where the limit may be infinite.

For any s > 0, we have

$$\zeta(K,\epsilon,d+s) \le \epsilon^s \zeta(K,\epsilon,d)$$

implying that if $m(K, d) < \infty$ then m(K, d) = 0. Thus, there is a unique $d^* \in [0, n]$ such that m(K, d) = 0 for all $d > d^*$, and $m(K,d) = \infty$ for all $d < d^*$. The Hausdorff dimension of K is

$$\dim_H K := d^*.$$

Intuitively, if we try to cover a square (which is a 2D set) by 1D balls (i.e. lines) lines then we need an infinite number of lines, but once we try a cover with 2D balls, the number of balls needed is finite. So, d^* is exactly the dimension for which the "volume" of K becomes finite.

For smooth shapes, or shapes with a small number of "corners", the Hausdorff dimension is an integer agreeing with the more standard topological dimension. For example, suppose that K is an ℓ -dimensional cube in \mathbb{R}^n . Intuitively speaking, for any $\epsilon > 0$ we require $\theta\left((1/\epsilon)^{\ell}\right)$ balls of radius ϵ to cover K. Hence,

$$\zeta(K,\epsilon,d) \approx (1/\epsilon)^{\ell} \epsilon^{d}.$$
 (12)

As $\epsilon \downarrow 0$, the right-hand side of (12) goes to ∞ if $d < \ell$, and to zero if $d > \ell$. It is not difficult to see that using balls with varying sizes does not change the analysis, so

$$\dim_H K = \ell.$$

However, for fractal sets (e.g. sets that contain strange attractors of chaotic dynamical systems) the Hausdorff dimension is typically not an integer.¹ In this context, the Hausdorff dimension is useful in quantifying sets of Lebesgue measure zero which are nevertheless "substantial". The next example from [51] demonstrates this.

Example 1. The Cantor set $E \subset [0,1]$ is defined inductively as follows. Let $E_0 := [0,1]$. For $j \ge 1$, the set E_j is obtained by removing the open middle third in any interval in E_{j-1} . For example, $E_1 = [0, 1/3] \cup [2/3, 1]$, and $E_2 = [0, 1/9] \cup [2/9, 1/3] \cup [2/3, 7/9] \cup [8/9, 1]$. Cantor's

¹In fact, one possible definition of a fractal set is: a set whose Hausdorff dimension strictly exceeds its topological dimension [40].

set is $E := \bigcap_{i=0}^{\infty} E_j$. Each E_j is the union of 2^j intervals of length 3^{-j} . The topological dimension of E is thus

$$\lim_{j \to \infty} (2/3)^j = 0.$$

We now determine $\dim_H E$. It can be shown [17, Ch. 3] that it is enough to consider the cover of E_j by 2^j intervals, each of length 3^{-j} , so for any $\epsilon > 0$ sufficiently small, $\zeta(E_i, \epsilon, d) =$ $2^{j}3^{-jd}$. Thus, $m(E_{j},d) = \left(\frac{2}{3^{d}}\right)^{j}$ and

$$m(E,d) = \lim_{j \to \infty} \left(\frac{2}{3^d}\right)^j.$$

If $d > \log(2)/\log(3)$ then m(E,d) = 0, and if $d < \infty$ $\log(2)/\log(3)$ then $m(E,d) = \infty$. Thus, $\dim_H E =$ $\log(2)/\log(3) \approx 0.631$. Intuitively, this implies that the Cantor set is less than a line, but more than a discrete set \square of points.

The next result summarizes some useful properties of \dim_H .

Proposition 1 ([17, Chapter 3]). The Hausdorff dimension satisfies the following properties:

- $\dim_H \emptyset = 0;$
- monotonicity: If $A, B \subseteq \mathbb{R}^n$ with $A \subseteq B$ then $\dim_H A \leq$ $\dim_H B$:
- countable subadditivity: If $A_i \subseteq \mathbb{R}^n$, $i = 1, 2, \ldots$, then $\dim_H(\cup_i A_i) \le \sum_i \dim_H(A_i);$
- If $A, B \subseteq \mathbb{R}^n$ are such that

$$\inf\{\operatorname{dist}(x,y): x \in A, y \in B\} > 0$$

then

$$\dim_H(A \cup B) = \dim_H A + \dim_H B.$$

The first three properties imply that \dim_H is an outer measure, and the fourth one that it is a metric outer measure.

B. Explicit formula for $A^{[k]}$

The additive compound, defined in (2), can be given explicitly in terms of the entries a_{ij} of A.

Proposition 2 ([41], [10]). Let $A \in \mathbb{R}^{n \times n}$ and fix $k \in$ $\{1, \ldots, n\}, 1 \leq i_1 < \cdots < i_k \leq n, \text{ and } 1 \leq j_1 < \cdots < j_k \leq n.$ The entry of $A^{[k]}$ corresponding to $(\alpha|\beta) =$ $(i_1, \ldots, i_k | j_1, \ldots, j_k)$ is:

- ∑^k_{ℓ=1} a_{iℓiℓ} if i_ℓ = j_ℓ for all ℓ ∈ {1,...,k};
 (-1)^{ℓ+m}a_{iℓjm} if all the indices in α and β coincide except for a single index $i_{\ell} \neq j_m$; and
- 0, otherwise.

To explain this, consider for example the case k = 2. Let $B := (I + \varepsilon A)^{(2)}$, and let

$$\delta_{pq} := \begin{cases} 1, \text{ if } p = q, \\ 0, \text{ otherwise.} \end{cases}$$

Fix $1 \le i_1 < i_2 \le n$ and $1 \le j_1 < j_2 \le n$. Then

$$B(i_1, i_2|j_1, j_2) = b_{i_1j_1}b_{i_2j_2} - b_{i_1j_2}b_{i_2j_1}$$

= $(\delta_{i_1j_1} + \varepsilon a_{i_1j_1})(\delta_{i_2j_2} + \varepsilon a_{i_2j_2})$
 $- (\delta_{i_1j_2} + \varepsilon a_{i_1j_2})(\delta_{i_2j_1} + \varepsilon a_{i_2j_1})$
= $c + (\delta_{i_1j_1}a_{i_2j_2} + \delta_{i_2j_2}a_{i_1j_1} - \delta_{i_1j_2}a_{i_2j_1})$
 $- \delta_{i_2j_1}a_{i_1j_2})\varepsilon + o(\varepsilon),$

where c is a constant that does not depend on ε . Thus, (2) gives

$$A^{[2]}(i_1, i_2|j_1, j_2) = \delta_{i_1 j_1} a_{i_2 j_2} + \delta_{i_2 j_2} a_{i_1 j_1} - \delta_{i_1 j_2} a_{i_2 j_1} - \delta_{i_2 j_1} a_{i_1 j_2},$$

and it is straightforward to see that this agrees with the expression given in Prop. 2. Note that Prop. 2 implies, in particular, that $A^{[1]} = A$, and $A^{[n]} = \text{trace}(A)$.

C. Real power of a square non-singular matrix

We first recall the definition of the real power of a complex number. Any complex number $a \in \mathbb{C} \setminus \{0\}$ can be written in the polar representation $a = |a| \exp(j\theta(a))$, where $j := \sqrt{-1}$, |a| > 0 is the modulus of a, and $\theta(a) \in (-\pi, \pi]$ is the argument of a. Then for any $\alpha \in \mathbb{R}$,

$$a^{\alpha} := |a|^{\alpha} \exp(j\alpha\theta(a)). \tag{13}$$

For example, for a = -5 we have $(-5)^{\alpha} = 5^{\alpha} \exp(j\alpha\pi)$. Note that although -5 is real, $(-5)^{\alpha}$ is in general a complex (non-real) number.

Recall that any $A \in \mathbb{C}^{n \times n}$ admits a Jordan canonical form [1]: there exist $T, J \in \mathbb{C}^{n \times n}$, with T non-singular, such that

$$A = T^{-1}JT, \quad J = \text{diag}(J_1, J_2, \dots, J_p),$$
 (14)

where every J_i is a Jordan block in the form

$$J_{i} = \begin{bmatrix} \ell_{i} & 1 & & \\ & \ell_{i} & \ddots & \\ & & \ddots & 1 \\ & & & \ell_{i} \end{bmatrix} \in \mathbb{C}^{m_{i} \times m_{i}},$$
(15)

with $\sum_{i=1}^{p} m_i = n$, and every ℓ_i , $i = 1, \ldots, p$, is an eigenvalue of A. The matrix J is unique, up to the ordering of the blocks J_i .

Since the real power of square matrices is a particular class of a matrix function [12], [14], it is defined according to the general definition given in [14, Def. 1.2].

Definition 1 (Real power of a non-singular square matrix). Consider a non-singular matrix $A \in \mathbb{C}^{n \times n}$, given in the Jordan canonical form (14), and let $\alpha \in \mathbb{R}$. Then,

$$A^{\alpha} := T^{-1} J^{\alpha} T, \qquad (16)$$

where

$$J^{\alpha} := \operatorname{diag}(J_1^{\alpha}, \dots, J_n^{\alpha}), \tag{17}$$

with

$$J_{i}^{\alpha} := \begin{bmatrix} \ell_{i}^{\alpha} & \frac{\alpha \ell_{i}^{\alpha-1}}{1!} & \cdots & \frac{\prod_{j=0}^{m_{i}-2} (\alpha-j)\ell_{i}^{\alpha-m_{i}+1}}{(m_{i}-1)!} \\ & \ell_{i}^{\alpha} & \ddots & \vdots \\ & & \ddots & \frac{\alpha \ell_{i}^{\alpha-1}}{1!} \\ & & & \ell_{i}^{\alpha} \end{bmatrix} .$$
(18)

The next result describes some of the properties of A^{α} .

Proposition 3 ([14, Thm. 1.13, Thm 1.15]). Let $A \in \mathbb{C}^{n \times n}$ be non-singular with the Jordan canonical form (14). Fix $\alpha, \beta \in \mathbb{R}$. Then

(a) the eigenvalues of A^{α} are ℓ_i^{α} , $i = 1, \ldots, p$;

(b)
$$(A^T)^{\alpha} = (A^{\alpha})^T;$$

- (c) $(XAX^{-1})^{\alpha} = XA^{\alpha}X^{-1}$, for any non-singular matrix $X \in \mathbb{C}^{n \times n}$;
- (d) $A^{\alpha}A^{\beta} = A^{\alpha+\beta}$.

Note that A^{α} is not necessarily real even if A is real. The next result provides a sufficient condition guaranteeing that A^{α} is real for any $\alpha \in \mathbb{R}$. Let $\mathbb{R}_{\leq 0} := \{x \in \mathbb{R} \mid x \leq 0\}$, and $\mathbb{R}_{>0} := \{x \in \mathbb{R} \mid x > 0\}$. Define the set of matrices $\Omega_n := \{X \in \mathbb{R}^{n \times n} \mid \operatorname{spec}(X) \cap \mathbb{R}_{\leq 0} = \emptyset\}$, where $\operatorname{spec}(X)$ is the set of eigenvalues of X.

Proposition 4. If $A \in \Omega_n$, then $A^{\alpha} \in \mathbb{R}^{n \times n}$ for any $\alpha \in \mathbb{R}$.

Proof: For $a \in \mathbb{C}$, let \overline{a} denote the complex conjugate of a. Note that $A \in \Omega_n$ implies that A is non-singular. Definition 1 guarantees that A^{α} is well-defined. Furthermore, the function $f(a) := a^{\alpha}$ is analytic on $\mathbb{C} \setminus \mathbb{R}_{\leq 0}$, and any connected component of $\mathbb{C} \setminus \mathbb{R}_{\leq 0}$ is closed under conjugation. Note that $f(\mathbb{R} \cap (\mathbb{C} \setminus \mathbb{R}_{\leq 0})) = f(\mathbb{R}_{>0}) \subset \mathbb{R}$. Therefore, the conditions given in [14, Thm. 1.18] are satisfied. This ensures that $f(\overline{A}) = \overline{f(A)}$. Since A is real, i.e., $\overline{A} = A$, this implies that $f(A) = \overline{f(A)}$, that is, f(A) is real.

The next result provides a formula for the derivative of a real power of a parameter-dependent matrix. This result will be used later on to derive a simple expression for the generalized additive compound of a matrix.

Lemma 1. Consider a matrix-valued mapping $A : \mathbb{R} \to \mathbb{C}^{n \times n}$. Assume that $A(\varepsilon)$ is non-singular and has constant (generalized) eigenvectors, that is, it can be written in the Jordan canonical form

$$A(\varepsilon) = T^{-1}J(\varepsilon)T, \ J(\varepsilon) = \operatorname{diag}(J_1(\varepsilon), J_2(\varepsilon), \dots, J_p(\varepsilon)),$$

where $T \in \mathbb{C}^{n \times n}$, and every $J_i : \mathbb{R} \to \mathbb{C}^{m_i \times m_i}$, $\sum_{i=1}^p m_i = n$, is a Jordan block as described in (15), where $\ell_i : \mathbb{R} \to \mathbb{C} \setminus \{0\}$, $i = 1, \ldots, p$, is C^1 . Then for any $\alpha \in \mathbb{R}$,

$$\frac{\mathrm{d}}{\mathrm{d}\varepsilon}A^{\alpha}(\varepsilon) = \alpha A^{\alpha-1}(\varepsilon)\frac{\mathrm{d}}{\mathrm{d}\varepsilon}A(\varepsilon).$$
(19)

Proof: By (18),
$$\frac{\mathrm{d}}{\mathrm{d}\varepsilon} J_i^{\alpha}(\varepsilon) = \alpha J_i^{\alpha-1}(\varepsilon) \frac{\mathrm{d}}{\mathrm{d}\varepsilon} \ell_i(\varepsilon)$$
, so

$$\frac{\mathrm{d}}{\mathrm{d}\varepsilon} J^{\alpha}(\varepsilon) = \alpha J^{\alpha-1}(\varepsilon) \operatorname{diag} \left(\frac{\mathrm{d}}{\mathrm{d}\varepsilon} \ell_1(\varepsilon) I_{m_1}, \dots, \frac{\mathrm{d}}{\mathrm{d}\varepsilon} \ell_p(\varepsilon) I_{m_p} \right)$$

$$= \alpha J^{\alpha-1}(\varepsilon) \frac{\mathrm{d}}{\mathrm{d}\varepsilon} J(\varepsilon),$$

where I_s denotes the $s \times s$ identity matrix. Thus,

$$\begin{aligned} \frac{\mathrm{d}}{\mathrm{d}\varepsilon} A^{\alpha}(\varepsilon) = & T^{-1} \frac{\mathrm{d}}{\mathrm{d}\varepsilon} J^{\alpha}(\varepsilon) T \\ = & \alpha T^{-1} J^{\alpha-1}(\varepsilon) T T^{-1} \frac{\mathrm{d}}{\mathrm{d}\varepsilon} J(\varepsilon) T. \end{aligned}$$

and using Definition 1 yields (19).

D. Kronecker product and Kronecker sum of matrices

The Kronecker product of two matrices $A \in \mathbb{C}^{n \times m}$ and $B \in \mathbb{C}^{p \times q}$ is

$$A \otimes B := \begin{bmatrix} a_{11}B & a_{12}B & \cdots & a_{1m}B \\ a_{21}B & a_{22}B & \cdots & a_{2m}B \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1}B & a_{n2}B & \cdots & a_{nm}B \end{bmatrix}, \quad (20)$$

where a_{ij} denotes the *ij*th entry of A. Hence, $A \otimes B \in \mathbb{C}^{(np) \times (mq)}$.

The Kronecker sum of two square matrices $X \in \mathbb{C}^{n \times n}$ and $Y \in \mathbb{C}^{m \times m}$ is

$$X \oplus Y := X \otimes I_m + I_n \otimes Y, \tag{21}$$

where I_r denotes the $r \times r$ identity matrix.

We list several properties of the Kronecker product and Kronecker sum that will be used later on. For $A \in \mathbb{C}^{n \times m}$, let A^* to denote the conjugate transpose of A.

Lemma 2 (see e.g. [13]). Consider matrices $A, C \in \mathbb{C}^{n \times m}$, $B, D \in \mathbb{C}^{p \times q}$, $F \in \mathbb{C}^{m \times \ell}$, $G \in \mathbb{C}^{q \times r}$, $X \in \mathbb{C}^{n \times n}$, and $Y \in \mathbb{C}^{m \times m}$. Then,

- (a) $(cA) \otimes B = A \otimes (cB) = c(A \otimes B)$ for any $c \in \mathbb{C}$;
- (b) $(A+C)\otimes B = A\otimes B + C\otimes B$;
- (c) $A \otimes (B + D) = A \otimes B + A \otimes D$;
- (d) $(A \otimes B)(F \otimes G) = (AF) \otimes (BG);$
- (e) $(A \otimes B)^* = A^* \otimes B^*;$
- $(f) \ (A \otimes B)^T = A^T \otimes B^T;$
- (g) if X, Y are non-singular then $(X \otimes Y)^{-1} = X^{-1} \otimes Y^{-1}$;
- (h) Let $\lambda_i(X)$, i = 1, ..., n, and $\lambda_j(Y)$, j = 1, ..., m, denote the eigenvalues of X and Y, respectively. Then, $X \otimes Y$ has eigenvalues $\lambda_i(X)\lambda_j(Y)$, i = 1, ..., n, j = 1, ..., m;
- (i) $X \oplus Y$ has eigenvalues $\lambda_i(X) + \lambda_j(Y)$, i = 1, ..., n, j = 1, ..., m;
- (j) $\exp(X) \otimes \exp(Y) = \exp(X \oplus Y).$

Property (a) implies that we can write $cA \otimes B := (cA) \otimes B$ or $A \otimes (cB)$, without any ambiguity.

For a real number $p \ge 1$, let

$$|x|_p := (|x_1|^p + \dots + |x_n|^p)^{\frac{1}{p}}, \qquad (22)$$

denote the L_p vector norm of $x \in \mathbb{C}^n$, and let $||\cdot||_p$ denote the induced matrix norm. Recall [18] that a norm $|\cdot|: \mathbb{C}^n \to \mathbb{R}_+$ is called monotonic if for any $x, y \in \mathbb{C}^n$ such that $|x_i| \leq |y_i|$ for all $i \in \{1, \ldots, n\}$, we have $|x| \leq |y|$. The L_p norm for any $p \geq 1$ is monotonic.

The next result uses ideas from [18] to determine the induced L_p matrix norms of a Kronecker product of two matrices.

Proposition 5. Fix $A \in \mathbb{C}^{\ell \times n}$, $B \in \mathbb{C}^{r \times m}$, and $p \ge 1$. Then,

$$||A \otimes B||_p = ||A||_p ||B||_p.$$
(23)

Proof: It is easy to verify that

$$x \otimes y|_p = |x|_p |y|_p$$
, for all $x \in \mathbb{C}^n$, $y \in \mathbb{C}^m$. (24)

Hence, in the terminology of [18], the L_p norms are cross norms. Let

$$\mathbb{C}^n \otimes \mathbb{C}^m := \{ z = x \otimes y : x \in \mathbb{C}^n, y \in \mathbb{C}^m \}.$$

Note that $\mathbb{C}^n \otimes \mathbb{C}^m \subseteq \mathbb{C}^{nm}$. Then,

$$\begin{split} \|A \otimes B\|_{p} &= \sup_{z \in \mathbb{C}^{n_{m}}} \frac{|(A \otimes B)z|_{p}}{|z|_{p}} \\ &\geq \sup_{z \in \mathbb{C}^{n} \otimes \mathbb{C}^{m}} \frac{|(A \otimes B)z|_{p}}{|z|_{p}} \\ &= \sup_{x \in \mathbb{C}^{n}, y \in \mathbb{C}^{m}} \frac{|(A \otimes B)(x \otimes y)|_{p}}{|x \otimes y|_{p}}, \end{split}$$

and applying Property (d) in Lemma 2 and (24) yields

$$\|A \otimes B\|_{p} \ge \sup_{x \in \mathbb{C}^{n}} \frac{|Ax|_{p}}{|x|_{p}} \sup_{y \in \mathbb{C}^{m}} \frac{|By|_{p}}{|y|_{p}} = \|A\|_{p} \|B\|_{p}.$$
 (25)

Thus, to complete the proof we need to show that $||A \otimes B||_p \le ||A||_p ||B||_p$. Note that $A \otimes B = (A \otimes I_r)(I_n \otimes B)$. Since induced matrix norms are sub-multiplicative,

$$\|A \otimes B\|_p \le \|A \otimes I_r\|_p \|I_n \otimes B\|_p.$$
⁽²⁶⁾

Let $\{e^i\}_{i=1}^n$, $\{f^j\}_{j=1}^m$ denote the canonical basis of \mathbb{C}^n and \mathbb{C}^m , respectively. Any $z \in \mathbb{C}^{nm}$ can be written as

$$z = \sum_{i=1}^{n} e^{i} \otimes y^{i} \text{ and } z = \sum_{j=1}^{m} x^{j} \otimes f^{j}, \qquad (27)$$

for some vectors $y^i \in \mathbb{C}^m$, $x^j \in \mathbb{C}^n$. As in [18], consider the norms

$$|z|_{p,e} := |\sum_{i=1}^{n} |y^{i}|_{p} e^{i}|_{p}, \quad |z|_{p,f} := |\sum_{j=1}^{m} |x^{j}|_{p} f^{j}|_{p}.$$
 (28)

It is easy to show that

$$|z|_{p,e} = |z|_{p,f} = |z|_p, \text{ for any } z \in \mathbb{C}^{nm}.$$
 (29)

Now,

$$|(I_n \otimes B)z|_{p,e} = |(I_n \otimes B)(\sum e^i \otimes y^i)|_{p,e}$$
$$= |\sum e^i \otimes By^i|_{p,e}$$
$$= |\sum |By^i|_p e^i|_p$$
$$\leq ||B||_p |\sum |y^i|_p e^i|_p$$
$$= ||B||_p |z|_{p,e}.$$

Thus, $||I_n \otimes B||_{p,e} \le ||B||_p$. On the other hand, (25) and (29) imply that $||I_n \otimes B||_{p,e} = ||I_n \otimes B||_p \ge ||I_n||_p ||B||_p = ||B||_p$. Hence, $||I_n \otimes B||_p = ||B||_p$. A similar argument using the norm $|\cdot|_{p,f}$ yields $||A \otimes I_r||_p = ||A||_p$. Now (26) yields $||A \otimes B||_p \le ||A||_p ||B||_p$.

E. Matrix measures

Let $|\cdot| : \mathbb{C}^n \to \mathbb{R}_+$ denote a vector norm. The induced matrix norm $||\cdot|| : \mathbb{C}^{n \times n} \to \mathbb{R}_+$ is $||A|| := \max_{|x|=1} |Ax|$, and the induced matrix measure $\mu : \mathbb{C}^{n \times n} \to \mathbb{R}$ is

$$\mu(A) := \lim_{\varepsilon \downarrow 0} \frac{||I + \varepsilon A|| - 1}{\varepsilon}.$$
 (30)

Matrix measures (also called logarithmic norms) play an important role in numerical linear algebra [46] and in systems and control theory [23], [3]. The reason for this is two-fold. First, if there exists a matrix measure such that the matrix A(t) in (6) satisfies

$$\mu(A(t)) \le \eta, \text{ for all } t \ge 0, \tag{31}$$

then

$$||X(t)|| \le \exp(\eta t)||X(0)||$$
, for all $t \ge 0$.

In particular, if $\eta < 0$ then this implies exponential convergence to zero with rate η . Second, let μ_p denote the matrix measure induced by the L_p vector norm, with $p \in \{1, 2, \infty\}$. Then for any $A \in \mathbb{C}^{n \times n}$,

$$\mu_{1}(A) = \max_{j \in \{1,...,n\}} \left(\operatorname{Re}(a_{jj}) + \sum_{\substack{i=1 \ i \neq j}}^{n} |a_{ij}| \right),$$

$$\mu_{2}(A) = \lambda_{\max}(A_{\operatorname{sym}}),$$

$$\mu_{\infty}(A) = \max_{i \in \{1,...,n\}} \left(\operatorname{Re}(a_{ii}) + \sum_{\substack{i=1 \ i \neq j}}^{n} |a_{ij}| \right),$$

(32)

where $A_{\text{sym}} := (A + A^*)/2$, and $\lambda_{\max}(B)$ denotes the largest eigenvalue of a Hermitian matrix B. Using these explicit formulas, it is sometimes possible to easily verify that (31) indeed holds, without computing X(t) itself.

Let $Q^{k,n}$ denote the sequence of k-tuples of distinct numbers from $\{1, \ldots, n\}$, in lexicographic order. For example,

$$Q^{2,3} = \{\{1,2\},\{1,3\},\{2,3\}\}\$$

Note that there are $\binom{n}{k}$ such k-tuples. Let $Q_{\ell}^{k,n}$ denote the ℓ th tuple in $Q^{k,n}$. For example, $Q_2^{2,3} = \{1,3\}$.

For $A \in \mathbb{C}^{n \times n}$ and any $k \in \{1, ..., n\}$, the matrix measures for $A^{[k]}$ are [32]:

$$\mu_{1}(A^{[k]}) = \max_{\{i_{1},...,i_{k}\} \in Q^{k,n}} \left(\sum_{p=1}^{k} \operatorname{Re}(a_{i_{p},i_{p}}) + \sum_{j \notin \{i_{1},...,i_{k}\}} (|a_{j,i_{1}}| + \dots + |a_{j,i_{k}}|) \right),$$

$$\mu_{2}(A^{[k]}) = \sum_{i=1}^{k} \lambda_{i} (A_{\operatorname{sym}}),$$

$$\mu_{\infty}(A^{[k]}) = \max_{\{i_{1},...,i_{k}\} \in Q^{k,n}} \left(\sum_{p=1}^{k} \operatorname{Re}(a_{i_{p},i_{p}}) + \sum_{j \notin \{i_{1},...,i_{k}\}} (|a_{i_{1},j}| + \dots + |a_{i_{k},j}|) \right).$$
(33)

Note that for k = 1, (33) reduces to (32).

Remark 1. For an integer $k \in \{1, \ldots, n\}$, the system (9)

is said to be k-order contracting if there exists a matrix measure μ such that

$$\mu(J_f^{[k]}(t,x)) \le -\eta < 0, \tag{34}$$

for any $t \ge 0$ and any x in the state-space [53] (see also [19], [25]). Roughly speaking, this implies that the dynamics contracts k-dimensional volumes. To explain this, consider the system $\dot{x}(t) = A(t)x(t)$, and suppose that it is norder contracting. Since $A^{[n]}(t) = \text{trace}(A(t))$, this implies that $\text{trace}(A(t)) \le -\eta < 0$ for all $t \ge 0$. Combining this with (7) implies the following. Fix n initial conditions a^i , i = 1, ..., n, and let

$$X(t) := \begin{bmatrix} x(t,a^1) & x(t,a^2) & \dots & x(t,a^n) \end{bmatrix}.$$

Then $|\det(X(t))| \leq \exp(-\eta t)|\det(X(0))|$. Therefore, *n*-dimensional volumes contract at an exponential rate. Following the terminology used in physics, we say that (9) is dissipative if it is *n*-order contracting.

From (32) and (33), it is straightforward to obtain the next result.

Proposition 6 ([53, Corollary 1]). Let $A \in \mathbb{C}^{n \times n}$ and $p \in \{1, 2, \infty\}$. Suppose that there exists an integer $\ell \in \{1, \ldots, n\}$ such that

$$\mu_p(A^{[\ell]}) \le 0,\tag{35}$$

then

$$\mu_p(A^{[\ell]}) \ge \mu_p(A^{[\ell+1]}) \ge \dots \ge \mu_p(A^{[n]}).$$
(36)

In other words, if $\dot{X} = AX$ is ℓ -order contracting w.r.t. $|\cdot|_p$ then it is also $(\ell+1)$ -order contracting, $(\ell+2)$ -order contracting, and so on [53].

The next two sections describe our main results.

III. α compounds

In this section, we define the new notions of the α multiplicative and α additive compound of a square matrix, and analyze the properties of these compounds. In the next section, we show how this leads to the new notion of α contracting systems, with α real.

A. α multiplicative compound

Consider a non-integer real number $\alpha \in (1, n) \setminus \mathbb{Z}$. From here on we decompose α as

$$\alpha = k + s, \ k \in \{1, 2, \dots, n - 1\}, \ s \in (0, 1).$$

Definition 2. Let $A \in \mathbb{C}^{n \times n}$ be non-singular. The α multiplicative compound matrix of A is

$$A^{(\alpha)} := (A^{(k)})^{1-s} \otimes (A^{(k+1)})^s.$$
(37)

Note that $A^{(\alpha)} \in \mathbb{C}^{r \times r}$, where $r := \binom{n}{k}\binom{n}{k+1}$, and that $A^{(\alpha)}$ may be complex (non-real) even if A is real. Since A is non-singular, $A^{(\ell)}$ is non-singular for all $\ell \in \{1, \ldots, n\}$, so $(A^{(k)})^{1-s}$ and $(A^{(k+1)})^s$ in (37) are well-defined.

For example, for $\alpha = 2.5$, we have k = 2 and s = 1/2, so $A^{(2.5)} = (A^{(2)})^{1/2} \otimes (A^{(3)})^{1/2}$, which can be interpreted

as a "multiplicative interpolation", with equal weights, between $A^{(2)}$ and $A^{(3)}$.

Example 2. Suppose that $D = \text{diag}(d_1, \ldots, d_4)$ is nonsingular. Fix $\alpha \in (2, 3)$, so that k = 2 and $s = \alpha - 2 \in (0, 1)$. Then

$$D^{(\alpha)} = (D^{(2)})^{1-s} \otimes (D^{(3)})^s$$

= diag($(d_1d_2)^{1-s}, (d_1d_3)^{1-s}, \dots, (d_3d_4)^{1-s}$))
 \otimes diag($(d_1d_2d_3)^s, (d_1d_2d_4)^s, (d_1d_3d_4)^s, (d_2d_3d_4)^s$)
= diag($d_1d_2d_3^s, d_1d_2d_4^s, \dots, d_2^sd_3d_4$),

so, any eigenvalue of $D^{(\alpha)}$ is a "multiplicative interpolation" between eigenvalues of $D^{(2)}$ and $D^{(3)}$.

Remark 2. If α is allowed to be an integer, say, $\alpha = k$ then s = 0 and (37) becomes

$$A^{(\alpha)} = (A^{(k)})^1 \otimes (A^{(k+1)})^0 = A^{(k)} \otimes I_r$$

where $r := \binom{n}{k+1}$. This is not equal to $A^{(k)}$ (but, ignoring multiplicity, it has the same eigenvalues as $A^{(k)}$). Therefore, Definition 2 only considers the case where α is not an integer. For the integer case, we will just use the standard definition for the k multiplicative compounds.

Remark 3. Suppose that $A \in \mathbb{C}^{n \times n}$ is non-singular. Using Property (b) in Prop. 3, Property (f) in Lemma 2, and the fact that $(X^{(\ell)})^T = (X^T)^{(\ell)}$ for any $X \in \mathbb{C}^{n \times n}$ and $\ell \in \{1, \ldots, n\}$ yields

$$(A^{(\alpha)})^T = \left((A^{(k)})^{1-s} \otimes (A^{(k+1)})^s \right)^T$$

= $((A^{(k)})^{1-s})^T \otimes ((A^{(k+1)})^s)^T$
= $((A^T)^{(k)})^{1-s} \otimes ((A^T)^{(k+1)})^s$
= $(A^T)^{(\alpha)},$

In particular, if $A = A^T$ then $A^{(\alpha)} = (A^{(\alpha)})^T$.

An alternative possible definition of the α multiplicative compound matrix is

$$A_{\rm alt}^{(\alpha)} := (A^{1-s})^{(k)} \otimes (A^s)^{(k+1)}.$$
 (38)

The next result shows that (37) and (38) are equivalent. This is useful because as we will see below some results are easier to derive using the definition in (37) and others using the alternative definition (38).

Theorem 1. Consider a non-singular matrix $A \in \mathbb{C}^{n \times n}$ and fix $\alpha \in (1, n) \setminus \mathbb{Z}$. Then

$$A^{(\alpha)} = A^{(\alpha)}_{alt}.$$
(39)

Proof: Fix $k \in \{1, ..., n\}$ and $s \in (0, 1)$. It is enough to show that

$$(A^{(k)})^s = (A^s)^{(k)}.$$
(40)

We first consider the case when A is diagonalizable, that is, there exist a non-singular $T \in \mathbb{C}^{n \times n}$, and a diagonal matrix $D \in \mathbb{C}^{n \times n}$ such that $A = T^{-1}DT$, which is also the Jordan canonical form of A. Then $(A^{(k)})^s =$ $((T^{-1})^{(k)}D^{(k)}T^{(k)})^s$. Using the fact that $(T^{-1})^{(k)} =$ $(T^{(k)})^{-1}$ and Property (c) in Prop. 3 gives

$$(A^{(k)})^s = (T^{(k)})^{-1} (D^{(k)})^s T^{(k)}$$

Since $D^{(k)}$ is also diagonal, $(D^{(k)})^s = (D^s)^{(k)}$. Hence,

$$(A^{(k)})^s = (T^{-1})^{(k)} (D^s)^{(k)} T^{(k)}$$
$$= (T^{-1} D^s T)^{(k)}$$
$$= (A^s)^{(k)}.$$

We conclude that (40) holds when A is diagonalizable. The proof in the general case follows from the fact that diagonalizable matrices are dense in $\mathbb{C}^{n \times n}$ (see, e.g., [20, Corollary 7.3.3]).

The next example demonstrates Theorem 1.

Example 3. Consider
$$A = \begin{bmatrix} a & 1 & 0 \\ 0 & a & 0 \\ 0 & 0 & b \end{bmatrix}$$
, with $a, b \neq 0$.
Note that A is not diagonalizable. A straightforward computation yields $A^{(2)} = \begin{bmatrix} a^2 & 0 & 0 \\ 0 & ab & b \\ 0 & 0 & ab \end{bmatrix}$, and
the Jordan decomposition of this matrix is $A^{(2)} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & ab & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a^2 & 0 & 0 \\ 0 & ab & 1 \\ 0 & 0 & ab \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1/b & 0 \\ 0 & 0 & 1 \end{bmatrix}$. Definition I gives
 $(A^{(2)})^s = \begin{bmatrix} 1 & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a^{2s} & 0 & 0 \\ 0 & a^s b^s & s(ab)^{s-1} \\ 0 & 0 & a^s b^s \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1/b & 0 \\ 0 & 0 & 1 \end{bmatrix}$
 $= \begin{bmatrix} a^{2s} & 0 & 0 \\ 0 & a^s b^s & sa^{s-1}b^s \\ 0 & 0 & a^s b^s \end{bmatrix}$.

On the other-hand,

$$(A^{s})^{(2)} = \begin{bmatrix} a^{s} & sa^{s-1} & 0\\ 0 & a^{s} & 0\\ 0 & 0 & b^{s} \end{bmatrix}^{(2)} = \begin{bmatrix} a^{2s} & 0 & 0\\ 0 & a^{s}b^{s} & sa^{s-1}b^{s}\\ 0 & 0 & a^{s}b^{s} \end{bmatrix},$$

so $(A^{(2)})^s = (A^s)^{(2)}$. Since $A^{(1)} = A$ and $A^{(3)} = \det(A)$, this implies that $A^{(\alpha)} = A^{(\alpha)}_{alt}$ for any $\alpha \in (1,3) \setminus \mathbb{Z}$.

The following discussion shows that, unlike the standard multiplicative compound matrix, the formula $(AB)^{(\alpha)} = A^{(\alpha)}B^{(\alpha)}$ does not hold in general.

Remark 4. Eq. (37) yields

$$(AB)^{(\alpha)} = ((AB)^{(k)})^{1-s} \otimes ((AB)^{(k+1)})^s = (A^{(k)}B^{(k)})^{1-s} \otimes (A^{(k+1)}B^{(k+1)})^s.$$
(41)

On the other-hand, using Property (d) in Lemma 2 gives

$$A^{(\alpha)}B^{(\alpha)} = ((A^{(k)})^{1-s} \otimes (A^{(k+1)})^s)((B^{(k)})^{1-s} \otimes (B^{(k+1)})^s) = ((A^{(k)})^{1-s}(B^{(k)})^{1-s}) \otimes ((A^{(k+1)})^s(B^{(k+1)})^s).$$
(42)

This shows that $(AB)^{(\alpha)} \neq A^{(\alpha)}B^{(\alpha)}$ in general. However it may hold in some special cases. Suppose for example that both A and B are diagonalizable, and that A commutes with B. Then, the same holds for $A^{(\ell)}, B^{(\ell)}, \ell \in \{1, ..., n\}$. This

implies that $A^{(\ell)}$ and $B^{(\ell)}$ are simultaneously diagonalizable. Then it is easy to show that for any $s \in (0,1)$, we have $(A^{(\ell)}B^{(\ell)})^s = (A^{(\ell)})^s (B^{(\ell)})^s$, and this implies that in this special case $(AB)^{(\alpha)} = A^{(\alpha)}B^{(\alpha)}$.

B. Spectral properties of $A^{(\alpha)}$

Let $\lambda_i(A) \in \mathbb{C}, \sigma_i(A) \in \mathbb{R}_+$, $i = 1, \dots, n$, denote the eigenvalues and singular values of $A \in \mathbb{R}^{n \times n}$, respectively, ordered such that

$$\operatorname{Re}(\lambda_1(A)) \ge \operatorname{Re}(\lambda_2(A)) \ge \dots \ge \operatorname{Re}(\lambda_n(A)),$$
 (43)

and

$$\sigma_1(A) \ge \sigma_2(A) \ge \dots \ge \sigma_n(A) \ge 0. \tag{44}$$

Lemma 3. Fix a non-singular matrix $A \in \mathbb{R}^{n \times n}$ and $\alpha \in (1, n) \setminus \mathbb{Z}$. Write $\alpha = k + s$, with k an integer and $s \in (0, 1)$.

(i) The eigenvalues of $A^{(\alpha)}$ are

$$\prod_{i \in Q_{\ell}^{k,n}} (\lambda_i(A))^{1-s} \prod_{i \in Q_j^{k+1,n}} (\lambda_i(A))^s$$

for
$$\ell \in \{1, \dots, \binom{n}{k}\}$$
, $j \in \{1, \dots, \binom{n}{k+1}\}$.
(ii) The eigenvalues of $(A^T A)^{(\alpha)}$ are

$$\prod_{i \in Q_{\ell}^{k,n}} (\sigma_i(A))^{2(1-s)} \prod_{i \in Q_j^{k+1,n}} (\sigma_i(A))^{2s}$$

for $\ell \in \{1, \dots, \binom{n}{k}\}, \ j \in \{1, \dots, \binom{n}{k+1}\}.$

Proof: It is well-known [32] that $\eta \in \mathbb{C}$ is an eigenvalue of $A^{(k)}$ iff it is the product of k eigenvalues of A, that is, there exists $\ell \in \{1, \ldots, \binom{n}{k}\}$ such that $\eta = \prod_{i \in Q_{\ell}^{k,n}} \lambda_i(A)$. By Definition 1,

$$\eta^{1-s} = \prod_{i \in Q_{\ell}^{k,n}} (\lambda_i(A))^{1-s}$$

is an eigenvalue of $(A^{(k)})^{1-s}$. Similarly, every eigenvalue of $(A^{(k+1})^s$ has the form

$$\prod_{k \in Q_i^{k+1,n}} (\lambda_i(A))^s$$

for some $j \in \{1, \dots, \binom{n}{k+1}\}$. Using Property (h) in Lemma 2 proves Property (i).

To prove Property (ii), note that $(A^T A)^{(\alpha)} = ((A^T A)^{(k)})^{1-s} \otimes ((A^T A)^{(k+1)})^s$. The eigenvalues of $A^T A$ are $\sigma_i^2(A)$, $i = 1, \ldots, n$, and using Property (i) yields Property (ii).

Remark 5. Fix a real $\alpha \geq 1$, and define $\omega_{\alpha} : \mathbb{R}^{n \times n} \to \mathbb{R}_+$ by

$$\omega_{\alpha}(A) := \sigma_1(A) \cdots \sigma_k(A) \left(\sigma_{k+1}(A)\right)^s$$

This function plays a crucial role in the Douady and Oesterlé Theorem [8], see Section IV below. Combining Property (ii) with the ordering of the eigenvalues and singular values implies that

$$\lambda_1((A^T A)^{(\alpha)}) = (\sigma_1(A) \cdots \sigma_k(A))^{2(1-s)}$$
$$(\sigma_1(A) \cdots \sigma_{k+1}(A))^{2s}$$
$$= (\omega_\alpha(A))^2.$$
(45)

Since $A^T A$ is symmetric, it follows from Remark 3 that so is $(A^T A)^{(\alpha)}$. Hence, $\lambda_1((A^T A)^{(\alpha)}) = \sigma_1((A^T A)^{(\alpha)}) =$ $||(A^T A)^{(\alpha)}||_2$, so we conclude that

$$||(A^T A)^{(\alpha)}||_2 = (\omega_\alpha(A))^2.$$
(46)

Thus, the α multiplicative compound provides a matrix norm expression for ω_{α} . This was our original motivation for introducing the α multiplicative compound.

Note also that in general

$$(\sigma_1(A^{(\alpha)}))^2 = \lambda_1((A^{(\alpha)})^T A^{(\alpha)})$$

$$\neq \lambda_1((A^T A)^{(\alpha)}).$$

We now turn to defining a generalization of the k additive compound.

C. α additive compound

The definition of the α additive compound matrix follows (2).

Definition 3. Let $A \in \mathbb{R}^{n \times n}$ and $\alpha \in (1, n) \setminus \mathbb{Z}$. The α additive compound matrix of A is

$$A^{[\alpha]} := \frac{d}{d\varepsilon} (I + \varepsilon A)^{(\alpha)}|_{\varepsilon=0}.$$
(47)

Note that for any $\varepsilon > 0$ sufficiently small and any $k \in \{1, 2, \ldots, n\}$, $(I + \varepsilon A)^{(k)}$ is non-singular and $(I + \varepsilon A)^{(k)} \in \Omega_n$. Hence, Proposition 4 and Definition 2 guarantee that $A^{[\alpha]}$ is well-defined and is a real matrix. Note also that (47) implies that

$$(I + \varepsilon A)^{(\alpha)} = I + \varepsilon A^{[\alpha]} + o(\varepsilon).$$
(48)

Example 4. Suppose that $D = \text{diag}(d_1, \ldots, d_4)$ is nonsingular. Fix $\alpha \in (2, 3)$, so that k = 2 and $s = \alpha - 2 \in (0, 1)$. Let $h_i(\varepsilon) := 1 + \varepsilon d_i$. By Example 2,

$$(I + \varepsilon D)^{(\alpha)} = \operatorname{diag}(h_1(\varepsilon)h_2(\varepsilon)h_3^s(\varepsilon), h_1(\varepsilon)h_2(\varepsilon)h_4^s(\varepsilon), \\ \dots, h_2^s(\varepsilon)h_3(\varepsilon)h_4(\varepsilon)) \\ = I + \varepsilon \operatorname{diag}(d_1 + d_2 + sd_3, d_1 + d_2 + sd_4, \\ \dots, sd_2 + d_3 + d_4) + o(\varepsilon),$$

and (47) yields

$$D^{[\alpha]} = \operatorname{diag}(d_1 + d_2 + sd_3, d_1 + d_2 + sd_4, \dots, sd_2 + d_3 + d_4).$$

Note that every eigenvalue of $D^{[\alpha]}$ is an "additive interpolation" of three eigenvalues of D.

The next result provides an expression for $A^{[\alpha]}$ in terms of $A^{[k]}$ and $A^{[k+1]}$.

Theorem 2. Fix
$$A \in \mathbb{R}^{n \times n}$$
 and $\alpha \in (1, n) \setminus \mathbb{Z}$. Then

$$A^{[\alpha]} = ((1 - s)A^{[k]}) \oplus (sA^{[k+1]}).$$
(49)

Note that this also shows that $A^{[\alpha]}$ is real, as $A^{[\ell]}$ is real for any $\ell \in \{1, \ldots, n\}$.

Proof: Consider the case where A is diagonalizable, that is, $A = T^{-1}DT$, where $T \in \mathbb{C}^{n \times n}$ is non-singular, and $D \in \mathbb{C}^{n \times n}$ is a diagonal matrix. Let $B(\varepsilon) := I + \varepsilon A$. Fix $k \in \{1, \ldots, n\}$. Then

$$(B(\varepsilon))^{(k)} = (T^{-1}(I + \varepsilon D)T)^{(k)}$$

= $(T^{-1})^{(k)}(I + \varepsilon D)^{(k)}T^{(k)}$.

Since D is diagonal, so is $(I + \varepsilon D)^{(k)}$. Therefore, $(B(\varepsilon))^{(k)}$ satisfies the conditions in Lemma 1. We use Lemma 1 to determine the derivative of $(B(\varepsilon))^{(\alpha)}$ with respect to ε . To simplify the notation, we write B for $B(\varepsilon)$. Then

$$\begin{split} \frac{d}{d\varepsilon}B^{(\alpha)} &= \frac{d}{d\varepsilon} \left((B^{(k)})^{1-s} \otimes (B^{(k+1)})^s \right) \\ &= \left(\frac{d}{d\varepsilon} (B^{(k)})^{1-s} \right) \otimes (B^{(k+1)})^s \\ &+ (B^{(k)})^{1-s} \otimes \left(\frac{d}{d\varepsilon} (B^{(k+1)})^s \right) \\ &= \left((1-s)(B^{(k)})^{-s} \frac{d}{d\varepsilon} B^{(k)} \right) \otimes (B^{(k+1)})^s \\ &+ (B^{(k)})^{1-s} \otimes \left(s(B^{(k+1)})^{s-1} \frac{d}{d\varepsilon} B^{(k+1)} \right). \end{split}$$

Setting $\varepsilon = 0$ and using the fact that $B(\varepsilon)|_{\varepsilon=0} = I$ and (2) yields

$$\frac{d}{d\varepsilon} (I + \varepsilon A)^{(\alpha)}|_{\varepsilon=0} = (1 - s)A^{[k]} \otimes I_{r_1} + I_{r_2} \otimes (sA^{[k+1]}) = ((1 - s)A^{[k]}) \oplus (sA^{[k+1]}),$$

where $r_1 := \binom{n}{k+1}$, and $r_2 := \binom{n}{k}$. This completes the proof when A is diagonalizable, and the general case follows from using a similar argument as in the proof of Theorem 1.

Remark 6. Suppose that $A \in \mathbb{R}^{n \times n}$. Then it is easy to see that Theorem 2 implies that $(A^{[\alpha]})^T = (A^T)^{[\alpha]}$. In particular, if A is symmetric then so is $A^{[\alpha]}$.

The next result shows that the α additive compound satisfies the same additivity property as the k additive compound.

Theorem 3. Let
$$A, B \in \mathbb{R}^{n \times n}$$
 and $\alpha \in (1, n) \setminus \mathbb{Z}$. Then,
 $(A + B)^{[\alpha]} = A^{[\alpha]} + B^{[\alpha]}.$

Proof: Using (49) gives

$$(A+B)^{[\alpha]} = ((1-s)(A+B)^{[k]}) \oplus (s(A+B)^{[k+1]})$$

= (1-s)(A+B)^{[k]} \otimes I_{r_1} + sI_{r_2} \otimes (A+B)^{[k+1]}

Applying (4) and Properties (a)-(c) in Lemma 2 yields

$$(A+B)^{[\alpha]} = (1-s)A^{[k]} \otimes I_{r_1} + sI_{r_2} \otimes A^{[k+1]} + (1-s)B^{[k]} \otimes I_{r_1} + sI_{r_2} \otimes B^{[k+1]} = ((1-s)A^{[k]}) \oplus (sA^{[k+1]}) + ((1-s)B^{[k]}) \oplus (sB^{[k+1]}) = A^{[\alpha]} + B^{[\alpha]}.$$

and this completes the proof.

The next result shows that (8) also holds for the α compounds.

Theorem 4. Let $A \in \mathbb{R}^{n \times n}$ and $\alpha \in (1, n) \setminus \mathbb{Z}$. Then,

$$\exp(A^{[\alpha]}t) = (\exp(At))^{(\alpha)}, \text{ for any } t \in \mathbb{R}.$$
 (50)

Proof: Using the alternative definition (38) gives

$$\begin{aligned} (\exp(At))_{\text{alt}}^{(\alpha)} &= ((\exp(At))^{1-s})^{(k)} \otimes ((\exp(At))^{s})^{(k+1)} \\ &= (\exp(At(1-s)))^{(k)} \otimes (\exp(Ats))^{(k+1)}, \end{aligned}$$

and applying (8) and Properties (a) and (j) in Lemma 2, gives

$$(\exp(At))_{\text{alt}}^{(\alpha)} = \exp(A^{[k]}t(1-s)) \otimes \exp(A^{[k+1]}ts) = \exp\left((A^{[k]}t(1-s)) \oplus (A^{[k+1]}ts)\right) = \exp\left(((A^{[k]}(1-s)) \oplus (A^{[k+1]}s))t\right)$$
(51)
$$= \exp(A^{[\alpha]}t).$$

Applying Theorem 1 completes the proof.

Combining Lemma 3, Theorem 4, and the fact that the eigenvalues of $\exp(At)$ are $\exp(\lambda_i(A)t)$, $i \in \{1, \ldots, n\}$, yields the following result.

Corollary 1. Let $A \in \mathbb{R}^{n \times n}$ and $\alpha \in (1,n) \setminus \mathbb{Z}$. The eigenvalues of $A^{[\alpha]}$ are

$$(1-s)\sum_{i\in Q_{\ell}^{k,n}}\lambda_i(A) + s\sum_{i\in Q_j^{k+1,n}}\lambda_i(A)$$

for $\ell \in \{1, \dots, \binom{n}{k}\}, j \in \{1, \dots, \binom{n}{k+1}\}.$

This implies in particular that $\lambda_1(A^{[\alpha]})$, that is, the eigenvalue of $A^{[\alpha]}$ with the largest real part is

$$\lambda_1(A^{[\alpha]}) = (1-s) \sum_{i=1}^k \lambda_i(A) + s \sum_{i=1}^{k+1} \lambda_i(A) = \sum_{i=1}^k \lambda_i(A) + s \lambda_{k+1}(A).$$
(52)

Remark 7. By the spectral properties of the standard additive compounds [32], $\sum_{i \in Q_{\ell}^{k,n}} \lambda_i(A)$ and $\sum_{i \in Q_{j}^{k+1,n}} \lambda_i(A)$ are the eigenvalues of $A^{[k]}$ and $A^{[k+1]}$, respectively. Now Property (i) in Lemma 2 implies that $(1 - s) \sum_{i \in Q_{\ell}^{k,n}} \lambda_i(A) + s \sum_{i \in Q_{j}^{k+1,n}} \lambda_i(A)$ is an eigenvalue of $A^{[\alpha]}$. This provides another proof for Corollary 1.

The next result follows from Thm. 4.

Corollary 2. Suppose that X(t) is the solution of the linear time-invariant matrix differential equation

$$\frac{d}{dt}X(t) = AX(t), \ X(0) = I,$$
(53)

with $A \in \mathbb{R}^{n \times n}$. Fix $\alpha \in (1, n) \setminus \mathbb{Z}$. Then

$$\frac{d}{dt}X^{(\alpha)}(t) = A^{[\alpha]}X^{(\alpha)}(t), \quad X^{(\alpha)}(0) = I.$$
(54)

Proof: The solution of (53) is $X(t) = \exp(At)$. By (50), $X^{(\alpha)}(t) = \exp(A^{[\alpha]}t)$, and this yields (54).

It is important to note that Eq. (54) holds only when A is a constant matrix. The next example demonstrates this.

Example 5. Consider the linear time-varying matrix differential equation

$$\begin{split} \dot{X}(t) &= A(t)X(t), \quad X(0) = I, \\ \text{with } A(t) &= \begin{bmatrix} 1 & 0 \\ 1 & t \end{bmatrix}. \text{ In this case, } X(t) = \begin{bmatrix} a(t) & 0 \\ b(t) & c(t) \end{bmatrix} \text{ with} \\ a(t) &:= \exp(t), \ b(t) := \int_0^t \exp\left(\sigma + \frac{t^2 - \sigma^2}{2}\right) \mathrm{d}\sigma, \\ c(t) &:= \exp(t^2/2). \end{split}$$

For any $t \in (0, \infty) \setminus \{2\}$, the Jordan decomposition of X is

$$X = \begin{bmatrix} \frac{a-c}{b} & 0\\ 1 & 1 \end{bmatrix} \begin{bmatrix} a & 0\\ 0 & c \end{bmatrix} \begin{bmatrix} \frac{b}{a-c} & 0\\ -\frac{b}{a-c} & 1 \end{bmatrix},$$

so

$$X^{s} = \begin{bmatrix} \frac{a-c}{b} & 0\\ 1 & 1 \end{bmatrix} \begin{bmatrix} a^{s} & 0\\ 0 & c^{s} \end{bmatrix} \begin{bmatrix} \frac{b}{a-c} & 0\\ -\frac{b}{a-c} & 1 \end{bmatrix} = \begin{bmatrix} a^{s} & 0\\ -\frac{a^{s}-c^{s}}{a-c}b & c^{s} \end{bmatrix}.$$

Let $\alpha = 1 + s$, with $s \in (0,1)$. Then Definition 2 gives
 $X^{(\alpha)} = (ac)^{s} X^{1-s}$, and (49) gives $A^{[\alpha]} = \begin{bmatrix} 1+st & 0\\ 1-s & s+t \end{bmatrix}.$
In particular, substituting $s = 0.5$ and $t = 1$ yields

$$\frac{\mathrm{d}}{\mathrm{d}t}X^{(1.5)}(1) = \begin{bmatrix} 5.23551 & 0\\ 4.48053 & 4.07742 \end{bmatrix}$$
$$A^{[1.5]}(1)X^{(1.5)}(1) = \begin{bmatrix} 5.23551 & 0\\ 4.26352 & 4.07742 \end{bmatrix},$$

so $\frac{\mathrm{d}}{\mathrm{d}t}X^{(\alpha)}(t) \neq A^{[\alpha]}(t)X^{(\alpha)}(t).$

It is well-known [32] that for any $A, T \in \mathbb{C}^{n \times n}$, with T non-singular, and any integer $\ell \in \{1, \ldots, n\}$, we have

$$(TAT^{-1})^{[\ell]} = T^{(\ell)}A^{[\ell]}(T^{-1})^{(\ell)} = T^{(\ell)}A^{[\ell]}(T^{(\ell)})^{-1}.$$
 (55)

The next result shows the α additive compounds under a coordinate transformation.

Theorem 5. Let $A, T \in \mathbb{C}^{n \times n}$, with T non-singular, and pick $\alpha \in (1, n) \setminus \mathbb{Z}$. Then

$$(TAT^{-1})^{[\alpha]} = (T^{(k)} \otimes T^{(k+1)}) A^{[\alpha]} (T^{(k)} \otimes T^{(k+1)})^{-1}.$$
 (56)

Proof: Let $B := TAT^{-1}$. Using (49) and (55) yields

$$B^{[\alpha]} = ((1-s)(TAT^{-1})^{[k]}) \oplus (s(TAT^{-1})^{[k+1]})$$

= (1-s)(T^(k)A^[k](T^{-1})^{(k)}) \otimes (T^{(k+1)}I_{r_1}(T^{(k+1)})^{-1})
+ s(T^(k)I_{r_2}(T^{(k)})^{-1}) \otimes (T^{(k+1)}A^{[k+1]}(T^{-1})^{(k+1)}).

Using Properties (d) and (g) in Lemma 2 gives

$$B^{[\alpha]} = (1-s)(T^{(k)} \otimes T^{(k+1)})(A^{[k]} \otimes I_{r_1}) ((T^{(k)})^{-1} \otimes (T^{(k+1)})^{-1}) + s(T^{(k)} \otimes T^{(k+1)}) (I_{r_2} \otimes A^{[k+1]}) ((T^{(k)})^{-1} \otimes (T^{(k+1)})^{-1}) = (T^{(k)} \otimes T^{(k+1)}) ((1-s)(A^{[k]} \otimes I_{r_1}) + sI_{r_2} \otimes A^{[k+1]}) (T^{(k)} \otimes T^{(k+1)})^{-1} = (T^{(k)} \otimes T^{(k+1)})A^{[\alpha]}(T^{(k)} \otimes T^{(k+1)})^{-1},$$

and this completes the proof.

Remark 8. For a non-singular $T \in \mathbb{R}^{n \times n}$, note that $T^{(k+\frac{1}{2})} = (T^{(k)})^{\frac{1}{2}} \otimes (T^{(k+1)})^{\frac{1}{2}}$, and thus

$$T^{(k+\frac{1}{2})})^{2} = \left((T^{(k)})^{\frac{1}{2}} \otimes (T^{(k+1)})^{\frac{1}{2}} \right) \left((T^{(k)})^{\frac{1}{2}} \otimes (T^{(k+1)})^{\frac{1}{2}} \right)$$
$$= \left((T^{(k)})^{\frac{1}{2}} (T^{(k)})^{\frac{1}{2}} \right) \otimes \left((T^{(k+1)})^{\frac{1}{2}} (T^{(k+1)})^{\frac{1}{2}} \right)$$
$$= T^{(k)} \otimes T^{(k+1)}.$$

Therefore, (56) can be rewritten in the more compact form

$$(TAT^{-1})^{[\alpha]} = (T^{(k+\frac{1}{2})})^2 A^{[\alpha]} (T^{(k+\frac{1}{2})})^{-2}.$$

The next subsection analyzes the matrix measure of the α additive compound. This will play an important role in the analysis of α contracting systems.

D. Matrix measures of the α additive compound

It is well-known [5] that if $|\cdot|$ is monotonic then the induced matrix norm satisfies $||D|| = \max(|d_1|, \ldots, |d_n|)$, for any diagonal matrix $D = \operatorname{diag}(d_1, \ldots, d_n) \in \mathbb{C}^{n \times n}$. This implies that the induced matrix measure satisfies

$$\mu(D) = \lim_{\varepsilon \downarrow 0} \varepsilon^{-1}(||I + \varepsilon D|| - 1)$$

=
$$\lim_{\varepsilon \downarrow 0} \varepsilon^{-1}(\max_{i} \{|1 + \varepsilon d_{i}|\} - 1)$$

=
$$\max_{i} \{\operatorname{Re}(d_{i})\}.$$
 (57)

Therefore, for any $\ell \in \{1, \ldots, n\}$, we have

$$\mu(D^{[\ell]}) = \max_{\{i_1, \dots, i_\ell\} \in Q^{\ell, n}} \left(\sum_{p=1}^{\ell} \operatorname{Re}(d_{i_p}) \right).$$
(58)

Eqs. (49) and (58) imply that

$$\mu(D^{[\alpha]}) = (1-s)\mu(D^{[k]}) + s\mu(D^{[k+1]}).$$
(59)

Our next goal is to show that this holds for general matrices. Towards this end, we first provide a useful expression for the matrix measure of a Kronecker sum of matrices.

Theorem 6. Let μ denote a matrix measure associated with a induced matrix norm $\|\cdot\|$ such that

$$||A \otimes B|| = ||A|| ||B|| \tag{60}$$

for any matrices A, B. Then

$$\mu(X \oplus Y) = \mu(X) + \mu(Y), \tag{61}$$

for any $X \in \mathbb{R}^{n \times n}$ and $Y \in \mathbb{R}^{m \times m}$.

Proof: Fix $\varepsilon > 0$. Properties (a) and (j) in Lemma 2 yield

$$||\exp(\varepsilon(X \oplus Y))|| = ||\exp(\varepsilon X \oplus \varepsilon Y)||$$
$$= ||\exp(\varepsilon X) \otimes \exp(\varepsilon Y)||$$

By (60), $||\exp(\varepsilon(X\oplus Y))||=||\exp(\varepsilon X)||||\exp(\varepsilon Y)||,$ and thus

$$\begin{split} \frac{d}{d\varepsilon} || \exp(\varepsilon(X \oplus Y)) || &= \left(\frac{d}{d\varepsilon} || \exp(\varepsilon X) || \right) || \exp(\varepsilon Y) || \\ &+ || \exp(\varepsilon X) || \left(\frac{d}{d\varepsilon} || \exp(\varepsilon Y) || \right). \end{split}$$

It follows from (30) that

$$\mu(A) = \left(\frac{d}{d\varepsilon} ||\exp(\varepsilon A)||\right)|_{\varepsilon=0}$$

for any $A \in \mathbb{R}^{n \times n}$. Thus,

$$\mu(X \oplus Y) = \left(\frac{d}{d\varepsilon} || \exp(\varepsilon(X \oplus Y)) ||\right) |_{\varepsilon=0}$$
$$= \left(\frac{d}{d\varepsilon} || \exp(\varepsilon X) ||\right) |_{\varepsilon=0} + \left(\frac{d}{d\varepsilon} || \exp(\varepsilon Y) ||\right) |_{\varepsilon=0}$$
$$= \mu(X) + \mu(Y),$$

and this completes the proof.

We can now provide a useful expression for the matrix measure of $A^{[\alpha]}$.

Corollary 3. Let μ_p denote a matrix measure induced by some L_p norm with $p \ge 1$. For any $A \in \mathbb{R}^{n \times n}$ and any $\alpha \in (1, n) \setminus \mathbb{Z}$, we have

$$\mu_p(A^{[\alpha]}) = (1-s)\mu_p(A^{[k]}) + s\mu_p(A^{[k+1]}).$$
(62)

Proof: From Prop. 5, (60) holds for all L_p norms. Since $A^{[\alpha]} = ((1-s)A^{[k]}) \oplus (sA^{[k+1]})$, Theorem 6 yields

$$\mu_p(A^{[\alpha]}) = \mu_p\left((1-s)A^{[k]}\right) + \mu_p\left(sA^{[k+1]}\right) = (1-s)\mu_p\left(A^{[k]}\right) + s\mu_p\left(A^{[k+1]}\right),$$

where the last equality follows from the homogeneity of the matrix measure, and the fact that $s \in (0, 1)$.

The next example demonstrates Corollary 3 in the case n = 2.

Example 6. Let $A \in \mathbb{R}^{2 \times 2}$. Fix $\alpha \in (1, 2)$. Then $\alpha = k + s$, with k = 1 and $s \in (0, 1)$, so

$$A^{[\alpha]} = ((1-s)A^{[1]}) \oplus (sA^{[2]})$$

= $((1-s)A) \oplus (s \operatorname{trace}(A))$
= $((1-s)A) \otimes I_1 + I_2 \otimes (s \operatorname{trace}(A))$
= $(1-s)A + s \operatorname{trace}(A)I_2.$

Recall that for any matrix measure μ and any $c \in \mathbb{R}$, $\mu(A + cI) = \mu(A) + c$ (see e.g. [7]). Thus,

$$\mu(A^{[\alpha]}) = \mu((1-s)A) + s \operatorname{trace}(A)$$

= (1-s)\mu(A^{[1]}) + s\mu(A^{[2]}).

Note that for this particular example, (62) holds for all matrix measures. $\hfill \Box$

In the remainder of this paper we always assume that μ is induced from some L_p norm, with $p \ge 1$.

The next section describes an application of the α compounds in the context of the Douady and Oesterlé Theorem [8]. For a modern treatment of this theorem and its numerous extensions and applications, see the recent monograph [17]. Some connections between contracting systems and the Douady and Oesterlé Theorem have already appeared in the note [25].

IV. An application: α contracting systems

In this section, $\alpha \in [1, n)$, and the special case where α is an integer is also allowed. The Hausdorff dimension of a set $K \subset \mathbb{R}^n$ is denoted by $\dim_H K$. Let $D \subseteq \mathbb{R}^n$ be an open set, and let $q: D \to \mathbb{R}^n$ be a C^1 mapping. Let

$$J_g(x) := \frac{\partial}{\partial x} g(x).$$

A set $K \subseteq D$ is said to be *negatively invariant under* g if $K \subseteq g(K)$. Intuitively speaking, g "increases" K. The next result is the Douady and Oesterlé theorem [8]. We state in the form given in [44].

Theorem 7. Suppose that $K \subset D$ is compact and negatively invariant under g. Fix $\alpha \in [1, n)$, and write $\alpha = k + s$, with k an integer and $s \in [0, 1)$. Let

$$\omega(K,\alpha,g) := \max_{x \in K} \left(\sigma_1(J_g(x)) \cdots \sigma_k(J_g(x)) (\sigma_{k+1}(J_g(x)))^s \right)$$
(63)

If
$$\omega(K, \alpha, g) < 1$$
 then $\dim_H K < \alpha$

Intuitively speaking, (63) implies that g is a "contraction in dimension α ", uniformly in K. If g "increases" K then necessarily $\dim_H K < \alpha$.

The next simple example demonstrates Thm. 7.

Example 7. Let $g : \mathbb{R}^3 \to \mathbb{R}^3$ be the linear mapping given by g(x) = diag(1, 1/2, 1/4)x. Then the singular values of the Jacobian of g are 1, 1/2, 1/4, and (63) holds for any $\alpha > 1$. Thus, Thm. 7 implies that for any compact set $K \subset \mathbb{R}^3$ such that $K \subseteq g(K)$, we have $\dim_H K \leq 1$. For example, the set $K := [0, 1] \times \{0\} \times \{0\}$ satisfies $K \subseteq g(K)$ and $\dim_H K =$ 1.

Using the α multiplicative compound we can express condition (63) in a more elegant form. Indeed, it follows from (46) that

$$(\omega(K,\alpha,g))^2 = \max_{x \in K} || \left(J_g^T(x) J_g(x) \right)^{(\alpha)} ||_2,$$

so the condition in Thm. 7 becomes

$$\max_{x \in K} || (J_g^T(x) J_g(x))^{(\alpha)} ||_2 < 1.$$

This provides a more intuitive description for "contraction in dimension α " of a mapping g.

Thm. 7 has been used to upper bound the Hausdorff dimension of invariant sets (and, in particular, attractors) of dynamical systems. Our results allow to restate and generazlize these results in a more intuitive fashion using the α additive compound.

Consider the time-varying dynamical system:

$$\dot{x} = f(t, x),\tag{64}$$

where f is C^1 . Let $x(t, t_0, x_0)$ denote the solution of (64) at time t with $x(t_0) = x_0$. We assume from here on that $t_0 =$ 0, and let $x(t, x_0) := x(t, 0, x_0)$. We also assume that there exists an invariant set $D \subseteq \mathbb{R}^n$, that is, for any $x_0 \in D$ we have $x(t, x_0) \in D$ for all $t \ge 0$. Let $J_f(t, x) := \frac{\partial}{\partial x} f(t, x)$, and consider the matrix differential equation

$$X(t) = J_f(x(t, x_0))X(t), \quad X(0) = X_0.$$

From now on, we always consider the L_p norms with $p \ge 1$ and the associated matrix measure μ . We begin with an auxiliary result.

Proposition 7. Let $K \subset \mathbb{R}^n$ be a compact invariant set of (64). Fix $\alpha \in [1, n)$ and let $\alpha = k + s$, with k integer and $s \in [0, 1)$. For an induced matrix measure μ and $t \ge 0$, let

$$\gamma_{J_f}(t) := \max_{x_0 \in K} \int_0^t \mu(J_f^{[\alpha]}(x(\tau, x_0))) \,\mathrm{d}\tau.$$

Then

$$\begin{aligned} ||X^{(k)}(t)||^{1-s} ||X^{(k+1)}(t)||^s \\ &\leq \exp(\gamma(t)) ||X^{(k)}_0||^{1-s} ||X^{(k+1)}_0||^s, \text{ for any } x_0 \in K \end{aligned}$$

Proof: Pick $\ell \in \{1, ..., n\}$. Since $\frac{d}{dt}X^{(\ell)} = J^{[\ell]}X^{(\ell)}$, $||X^{(\ell)}(t)|| \leq \exp(\int_0^t \mu(J^{[\ell]}(\tau)) \, d\tau)||X_0^{(\ell)}||$. Applying this bound to $||X^{(k)}(t)||^{1-s}||X^{(k+1)}(t)||^s$, and using (62) completes the proof. ■

We say that a constant set $K \subseteq D$ is a strongly invariant set of (64) if

$$K = x(t, K) \text{ for all } t \ge 0.$$
(65)

For example, an equilibrium or a limit cycle are strongly invariant sets.

We can now bound the Hausdorff dimension of strongly invariant sets of (64), thus extending a result in [44]. For generality, contraction theory typically uses contraction metrics [23] and associated scaled norms. Consider a C^1 scaling matrix $\Theta: K \to \mathbb{R}^{n \times n}$ satisfying

$$\det(\Theta(z)) \neq 0 \text{ for all } z \in K.$$
(66)

Let $\Theta_f(z)$ denote the matrix obtained by replacing every entry $\theta_{ij}(z)$ in $\Theta(z)$ by the value $(\frac{\partial \theta_{ij}(z)}{\partial z})^T f(z)$, and define the so-called generalized Jacobian [23] as

$$\bar{J} := \Theta_f \Theta^{-1} + \Theta J_f \Theta^{-1}.$$

Note that if $\Theta(z) = I$ for all z then $\overline{J} = J_f$. The next result bounds the Hausdorff dimension of a strongly invariant set using the generalized Jacobian \overline{J} .

Theorem 8. Let $K \subset \mathbb{R}^n$ be a compact and strongly invariant set of (64). Fix $\alpha \in [1, n)$ and let $\alpha = k + s$, with k integer and $s \in [0, 1)$. Assume there exist an induced matrix measure $\mu : \mathbb{R}^{n \times n} \to \mathbb{R}$ and $\tau > 0$ such that

$$\gamma_{\bar{J}}(\tau) < 0. \tag{67}$$

Then $\dim_H K < \alpha$.

 $\begin{array}{l} \textit{Proof: Define }g:\mathbb{R}_+\times K\to K \text{ by }g(t,x_0):=x(t,x_0).\\ \text{Then }J_g(t,x_0):=\frac{\partial}{\partial x_0}x(t,x_0). \text{ Let} \end{array}$

$$Y(t, x_0) := \Theta(x(t, x_0)) \frac{\partial}{\partial x_0} g(t, x_0).$$
(68)

To simplify the notation, we sometimes write $\Theta(x)$ or $\Theta(t)$

for $\Theta(x(t, x_0))$. By (68),

$$\dot{Y} \ = \ \dot{\Theta} \frac{\partial}{\partial x_0} g + \Theta \frac{\partial}{\partial x_0} \dot{x} \ = \ (\Theta_f J_g + \Theta J_f \Theta^{-1}) Y.$$

Thus, $Y(t, x_0)$ is the solution at time t of the matrix differential equation $\dot{Y} = \bar{J}Y$, initialized with $Y(0) = \Theta(x_0)$. Let $c(x_0) := ||\Theta^{(k)}(x_0)||^{1-s} ||\Theta^{(k+1)}(x_0)||^s$. Prop. 7 and (67) imply that

$$||Y^{(k)}(\tau)||^{1-s}||Y^{(k+1)}(\tau)||^{s} \le c(x_{0})\exp(\gamma_{\bar{J}}(\tau)),$$

for any $x_0 \in K$. Hence, for any integer $\ell \geq 1$,

$$\begin{aligned} ||\Theta^{(k)}(\ell\tau)J_{g}^{(k)}(\ell\tau)||^{1-s}||\Theta^{(k+1)}(\ell\tau)J_{g}^{(k+1)}(\ell\tau)||^{s} \\ &\leq c(x_{0})\exp(\gamma_{\bar{J}}(\ell\tau)). \end{aligned}$$
(69)

Recall that if $|| \cdot || : \mathbb{C}^{n \times n} \to \mathbb{R}_+$ is an induced matrix norm and $P \in \mathbb{C}^{n \times n}$ is non-singular, then the *P*-weighted induced matrix norm is $||M||_P := ||PMP^{-1}||$. Eq. (69) yields

$$\begin{aligned} ||J_g^{(k)}(\ell\tau)||_{\Theta^{(k)}(\ell\tau)}^{1-s}||J_g^{(k+1)}(\ell\tau)||_{\Theta^{(k+1)}(\ell\tau)}^s \\ &\leq c(x_0)\exp(\gamma_{\bar{J}}(\ell\tau))||(\Theta^{(k)}(\ell\tau))^{-1}||^{1-s}||(\Theta^{(k+1)}(\ell\tau))^{-1}||^s. \end{aligned}$$

Since K is compact, we can make the right-hand side of this equation arbitrarily small by taking ℓ large enough. Using the equivalence of norms implies that there exists an integer $\bar{\ell}$ such that $||J_g^{(k)}(\bar{\ell}\tau)||_2^{1-s}||J_g^{(k+1)}(\bar{\ell}\tau)||_2^s < 1$. Let $\sigma_i, i = 1, ..., n$, denote the singular values of $J_g(\bar{\ell}\tau)$. Then we conclude that

$$\sigma_1 \dots \sigma_k \sigma_{k+1}^s < 1.$$

Since $g(\bar{\ell}\tau, K) = K$, Thm. 7 implies that $\dim_H K < \alpha$. Of course, a sufficient condition for (67) to hold is that $\mu(\bar{J}^{[\alpha]}(x)) < 0$ for all $x \in K$.

From now on we consider for simplicity the non-scaled case, i.e. $\bar{J} = J_f$. Thm. 8 naturally leads to the following new definition.

Definition 4. Let μ be a matrix measure induced by an L_p norm with $p \ge 1$. Suppose that the trajectories of (64) evolve on a state space D. Pick a real $\alpha \ge 1$. System (64) is called α contracting w.r.t. the norm $|\cdot|_p$ if

$$\mu(J_f^{[\alpha]}(t,x)) \le -\eta < 0, \text{ for all } t \ge 0, \ x \in D.$$
 (70)

Remark 9. An important property of contracting systems is that various compositions of contracting systems yield a contracting system [23], [42], [37], [25]. The subadditivity of the matrix measure and Thm. 3 suggest that this remains valid for interconnections of α contracting systems. As a simple example, consider the interconnected system

$$\dot{x}(t) = c_1(t)f(t,x) + c_2(t)g(t,x), \tag{71}$$

with $c_i(t) \ge 0$ for any $t \ge 0$. The Jacobian of this system is $c_1J_f + c_2J_g$, and

$$\mu((c_1 J_f + c_2 J_g)^{[\alpha]}) = \mu(c_1 J_f^{[\alpha]} + c_2 J_g^{[\alpha]})$$

$$\leq c_1 \mu(J_f^{[\alpha]}) + c_2 \mu(J_g^{[\alpha]})$$

Thus, it is straightforward to provide sufficient conditions for α contraction of (71) in terms of the sub-systems $\dot{x}(t) = f(t, x)$ and $\dot{x}(t) = g(t, x)$.

Corollary 4. Suppose that (64) is α contracting. Then any compact and strongly invariant set has Hausdorff dimension smaller than α .

Example 8. Consider the system $\dot{x}(t) = A(t)x(t)$, with Example 8. Consider the spin $A(t) = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & -t \end{bmatrix}$. Take $\alpha = 2 + s$ with $s \in (0, 1)$. By Prop. 2, $A^{[2]} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & -t & -1 \\ 0 & 1 & -t \end{bmatrix}$ and $A^{[3]} = -t$. Hence, the

$$A^{[\alpha]} = ((1-s)A^{[2]}) \oplus (sA^{[3]}) = \begin{bmatrix} -st & 0 & 0\\ 0 & -t & s-1\\ 0 & 1-s & -t \end{bmatrix}.$$

Note that $\mu_2(A^{[2]}) = 0$, and $\mu_2(A^{[\alpha]}) = -st < 0$ for any t > 0. That is, this system is (2 + s) contracting with $s \in (0,1)$. Thm. 8 thus guarantees that any compact and strongly invariant set K satisfies $\dim_H K < 2 + s$. Since $s \in (0,1)$ can be arbitrarily small,

$$\dim_H K \le 2. \tag{72}$$

For example, the set $K := \{x \in \mathbb{R}^3 : x_1^2 + x_2^2 \le c, x_3 = 0\}$ with any $c \ge 0$ is compact, strongly invariant, and satisfies (72).

The next results shows that if the system is α contracting w.r.t. $|\cdot|_p$, for some $p \in \{1, 2, \infty\}$, then it is also $\bar{\alpha}$ contracting w.r.t. the same norm for any $\bar{\alpha} \geq \alpha$.

Theorem 9. Consider the system (64). Suppose that condition (70) holds for some μ_p with $p = \{1, 2, \infty\}$, and some $\alpha \in [1, n)$. Then (64) is β contracting for any $\beta \in (\alpha, n]$.

Proof: Consider first the case that α is an integer, that is, $\alpha = k \in \{1, ..., n - 1\}$. Then (70) becomes

$$\mu_p(J_f^{[\kappa]}(t,x)) \le -\eta < 0, \text{ for all } t \ge 0, x \in D.$$

Fix arbitrary $x \in D$ and $t \ge 0$. To simplify the notation, we write J_f for $J_f(t,x)$. Prop. 6 ensures that $\mu_p(J_f^{[k+1]}) \leq$ $\mu_p(J_f^{[k]})$. Fix $\varepsilon \in (0, 1)$. By Corollary 3,

$$\begin{split} \mu_p(J_f^{[\alpha+\varepsilon]}) = & \mu_p(J_f^{[k+\varepsilon]}) = (1-\varepsilon)\mu_p(J_f^{[k]}) + \varepsilon\mu_p(J_f^{[k+1]}) \\ = & \mu_p(J_f^{[k]}) - \varepsilon \left(\mu_p(J_f^{[k]}) - \mu_p(J_f^{[k+1]})\right) \\ \leq & \mu_p(J_f^{[k]}), \end{split}$$

so and the system is $\alpha + \varepsilon$ contracting.

Now suppose that α is not an integer, i.e. $\alpha = k + s$, with k an integer and $s \in (0, 1)$. Then condition (70) becomes

$$(1-s)\mu_p(J_f^{[k]}) + s\mu_p(J_f^{[k+1]}) \le -\eta < 0.$$

We claim that

$$\mu_p(J_f^{[k]}) \ge \mu_p(J_f^{[k+1]}). \tag{73}$$

Indeed, if $\mu_p(J_f^{[k]}) \ge 0$, then $\mu_p(J_f^{[k+1]}) \le -\eta/s < 0$, so (73)

holds, and if $\mu_p(J_f^{[k]})<$ 0, then (73) follows from Prop. 6. Hence, for any $\varepsilon \in (0, 1-s)$,

$$\mu_p(J_f^{[\alpha+\varepsilon]}) = (1-s-\varepsilon)\mu_p(J_f^{[k]}) + (s+\varepsilon)\mu_p(J_f^{[k+1]})$$
$$= \mu_p(J_f^{[\alpha]}) - \varepsilon \left(\mu_p(J_f^{[k]}) - \mu_p(J_f^{[k+1]})\right)$$
$$\leq \mu_p(J_f^{[\alpha]}),$$

and this completes the proof.

Theorem 9 implies the following result.

Corollary 5. Consider the dynamical system (64). Suppose that condition (70) holds for some μ_p with $p = \{1, 2, \infty\}$, and some $\alpha \in [1, n]$. Then there exists a minimal real value $\alpha^* \in$ $[1, \alpha]$ such that (64) is β contracting for any $\beta > \alpha^*$.

In other words, contraction is not a binary property, but rather the system is located on a continuous axis of contraction level. It is important to note that the value α^* depends on the norm that induces the matrix measure. This is also true of standard contraction, where the analysis of contraction critically depends on using the "right" norm.

Several recent papers considered systems that are, in some sense, on "the verge of contraction" [23], [26], [45], [16], [29], [31]. Such systems are referred to as semi-contracting [23], [50], or sometimes weakly-contracting [16] (note that this terminology is used instead for k-order contraction in [25]). Since 1-order contraction corresponds to contracting systems, we can expect semi-contracting systems to be α contracting for some $\alpha > 1$. This is indeed the case. We demonstrate this for the important example of studying synchronization using contraction theory [48], [33].

Example 9. Consider the LTI system

$$\dot{x} = -Lx,\tag{74}$$

where L is the Laplacian of a (directed or undirected) weighted graph with a globally reachable vertex. We claim that (74) is not 1-order contracting w.r.t. any norm. Yet, for any $\varepsilon \in (0,1)$ there exists a vector norm $|\cdot|$ such that (74) is $1 + \varepsilon$ contracting w.r.t. to $|\cdot|$.

Indeed, for any $c \in \mathbb{R}$ we have that $c1_n$ is an equilibrium of (74), so the system cannot be 1-order contracting w.r.t. any norm. On the other-hand, the eigenvalues $\lambda_i(A)$, ordered as in (43), satisfy $\lambda_1 = 0$ and $\operatorname{Re}(\lambda_2) < 0$. Fix $\varepsilon \in (0, 1)$. *By* (52),

$$\operatorname{Re}(\lambda_1(A^{[1+\varepsilon]})) = \operatorname{Re}(\lambda_1(A) + \varepsilon \lambda_2(A)) < 0,$$

so $A^{[1+\varepsilon]}$ is Hurwitz, and it is well-known [3], [22] that this implies that there exists a matrix measure μ such that $\mu(A^{[1+\varepsilon]}) < 0$. Combining this with Corollary 4 implies that any compact and strongly invariant set K of the dynamics satisfies $\dim_H K \leq 1$. This agrees with the fact that the \square dynamics converges to "lines".

The next example demonstrates an application of our theoretical results to the control of a chaotic system.

Example 10. A popular example for a chaotic system, introduced by Thomas [47] (see also the recent review [4]), is

$$\dot{x}_1 = \sin(x_2) - bx_1,
\dot{x}_2 = \sin(x_3) - bx_2,
\dot{x}_3 = \sin(x_1) - bx_3,$$
(75)

where b > 0 is the dissipation constant. Note that the convex set $D := \{x \in \mathbb{R}^3 : b|x|_{\infty} \leq 1\}$ is an invariant set of the dynamics.

This system undergoes a series of bifurcations as b decreases. For b > 1 the origin is the single stable equilibrium. When b = 1 it undergoes a pitchfork bifurcation, splitting into two attractive fixed points. As b is decreased further to $b \approx 0.32899$ these undergo a Hopf bifurcation, creating a stable limit cycle. The limit cycle undergoes a period doubling cascade and becomes chaotic at $b \approx 0.208186$.

Fig. 1 depicts the solution of the system emanating from $\begin{bmatrix} -1 & 1 \end{bmatrix}^T$ for

$$b = 0.193186$$
 (76)

Note the symmetric strange attractor.

Let f denote the vector field in (75). The Jacobian is

$$J_f(x) = \begin{bmatrix} -b & \cos(x_2) & 0\\ 0 & -b & \cos(x_3)\\ \cos(x_1) & 0 & -b \end{bmatrix},$$

and thus

$$J_f^{[2]}(x) = \begin{bmatrix} -2b & \cos(x_3) & 0\\ 0 & -2b & \cos(x_2)\\ -\cos(x_1) & 0 & -2b \end{bmatrix},$$

and $J_f^{[3]} = \text{trace}(J_f(x)) = -3b$. This implies that the system is 3 contracting (that is, dissipative), w.r.t. any norm, for any b > 0. Let $\alpha = 2 + s$, with $s \in (0, 1)$. Then

$$\begin{aligned} J_f^{[\alpha]}(x) &= (1-s)J_f^{[2]}(x) \oplus sJ_f^{[3]}(x) \\ &= \begin{bmatrix} -(2+s)b & (1-s)\cos(x_3) & 0 \\ 0 & -(2+s)b & (1-s)\cos(x_2) \\ -(1-s)\cos(x_1) & 0 & -(2+s)b \end{bmatrix}. \end{aligned}$$

This implies that

$$\mu_1(J_f^{[\alpha]}(x)) \le 1 - 2b - s(b+1), \text{ for all } x \in D.$$

We conclude that for any $b \in (0, 1/2)$ the system is 2 + s contracting for any $s > \frac{1-2b}{1+b}$.

We now demonstrate how our results can be applied to design a partial-state controller for the system guaranteeing that the closed-loop system has a "well-ordered" behaviour. Suppose that the closed-loop system is:

$$\dot{x} = f(x) + g(x),$$

where g is the controller. Let $\alpha = 2 + s$, with $s \in (0, 1)$. The Jacobian of the closed-loop system is $J_{cl} := J_f + J_g$, so

$$\begin{split} \mu_1(J_{cl}^{[\alpha]}) &= \mu_1(J_f^{[\alpha]} + J_g^{[\alpha]}) \le \mu_1(J_f^{[\alpha]}) + \mu_1(J_g^{[\alpha]}) \\ &\le 1 - 2b - s(b+1) + \mu_1(J_g^{[\alpha]}). \end{split}$$



Fig. 1. A trajectory of (75) emanating from $x(0) = \begin{bmatrix} -1 & 1 & 1 \end{bmatrix}^T$ for the dissipation parameter in (76).

This implies that the closed-loop system is α contracting if

$$\mu_1(J_q^{[\alpha]}(x)) < s(b+1) + 2b - 1 \text{ for all } x \in D.$$
(77)

Consider, for example, the controller

$$g(x_1, x_2) = \text{diag}(c, c, 0)x$$
, with $c < 0$.

 $J_g^{[\alpha]} = c \operatorname{diag}(2, 1+s, 1+s)$

Then

and for any c < 0 condition (77) becomes

$$(1+s)c < s(b+1) + 2b - 1.$$
(78)

This provides a simple recipe for determining the gain c so that the closed-loop system is 2 + s contracting. For example, when $s \rightarrow 0$, Eq. (78) yields

$$c < 2b - 1$$

and this guarantees that the closed-loop system is 2-order contracting. Recall that in a 2-order contracting system every nonempty omega limit set is a single equilibrium, thus ruling out chaotic attractors and even non-trivial limit cycles [21]. Fig. 2 depicts the behaviour of the closed-loop system with b as in (76) and c = 2b - 1.1. The closed-loop system is thus 2-order contracting, and as expected every solution converges to an equilibrium.

V. CONCLUSION

The k multiplicative and k additive compounds of a matrix play an important role in geometry, multi-linear algebra, dynamical systems, and more. These compounds are based on $k \times k$ minors and are thus defined for integer values of k only. The k compounds were recently used to study an extension of contracting systems to k-order contracting systems [53].

Here, we generalised k compounds to α compounds, with α real, and analyzed the properties of these compounds. As an application, we showed that these compounds provide more direct and intuitive interpretation of important functions,



Fig. 2. Several trajectories of the closed-loop system for the dissipation parameter in (76). The circles denote the initial conditions of the trajectories.

e.g. $\omega(K, \alpha, g)$ appearing in the seminal work of Douady and Oesterlé. We also introduced the new notion of α contracting systems, with α real, generalizing the notion of korder contracting systems with k an integer, recently analyzed in [53]. Thus, rather than a binary choice – contracting or not contracting in a given metric – one can place any system on a continuous axis of contraction levels.

Due to space limitations, we focused here on theoretical results, but we believe that many applications are possible. First, there exist nonlinear systems where the "level of contraction" naturally changes in a continuous way, for example, systems that involve a continuous-time dynamics and discretetime switching (see, e.g. [24]). Second, contraction theory (i.e. the theory of 1-order contracting systems [23]) has found many applications in control synthesis (see e.g., [39], [43], [27], [38], [28], [36], [55]). An interesting research direction is to apply the generalization described here to control synthesis in such contexts. Finally, our results could be used to define generalized notions of convexity in optimization and machine learning. Just as Riemannian convexity of a scalar function with respect to a metric is equivalent to contraction in that metric of natural gradient descent [50], notions of α Riemannian convexity could similarly be defined through equivalent α contracting autonomous dynamical systems.

REFERENCES

- [1] N. I. Achieser, *Theory of Approximation (translated by C.J. Hyman)*. Mineola, New York: Dover Publications, Inc., 1992.
- [2] R. Alseidi, M. Margaliot, and J. Garloff, "On the spectral properties of nonsingular matrices that are strictly sign-regular for some order with applications to totally positive discrete-time systems," *J. Math. Anal. Appl.*, vol. 474, pp. 524–543, 2019.
- [3] Z. Aminzare and E. D. Sontag, "Contraction methods for nonlinear systems: A brief introduction and some open problems," in *Proc. 53rd IEEE Conf. on Decision and Control*, Los Angeles, CA, 2014, pp. 3835– 3847.
- [4] V. Basios, C. G. Antonopoulos, and A. Latifi, "Labyrinth chaos: Revisiting the elegant, chaotic, and hyperchaotic walks," *Chaos*, vol. 30, no. 11, p. 113129, 2020.
- [5] F. L. Bauer, J. Stoer, and C. Witzgall, "Absolute and monotonic norms," *Numer. Math.*, vol. 3, pp. 257–264, 1961.

- [6] T. Ben-Avraham, G. Sharon, Y. Zarai, and M. Margaliot, "Dynamical systems with a cyclic sign variation diminishing property," *IEEE Trans. Automat. Control*, vol. 65, pp. 941–954, 2020.
- [7] C. A. Desoer and M. Vidyasagar, Feedback Synthesis: Input-Output Properties. Philadelphia: SIAM, 2009.
- [8] A. Douady and J. Oesterle, "Dimension de Hausdorff des attracteurs," C. R. Acad. Sc. Paris, vol. 290, pp. 1135–1138, 1980.
- [9] S. M. Fallat and C. R. Johnson, *Totally Nonnegative Matrices*. Princeton, NJ: Princeton University Press, 2011.
- [10] M. Fiedler, Special Matrices and Their Applications in Numerical Mathematics, 2nd ed. Mineola, NY: Dover Publications, 2008.
- [11] F. R. Gantmacher and M. G. Krein, Oscillation Matrices and Kernels and Small Vibrations of Mechanical Systems. Providence, RI: American Mathematical Society, 2002, translation based on the 1941 Russian original.
- [12] F. R. Gantmacher, *The Theory of Matrices (translated by K. A. Hirsch)*. New York: Chelsea Publishing Company, 1966.
- [13] A. Graham, Kronecker Products and Matrix Calculus with Applications. Courier Dover Publications, 2018.
- [14] N. J. Higham, Functions of Matrices: Theory and Computation. Philadelphia: SIAM, 2008.
- [15] R. A. Horn and C. R. Johnson, *Matrix Analysis*, 2nd ed. Cambridge University Press, 2013.
- [16] S. Jafarpour, P. Cisneros-Velarde, and F. Bullo, "Weak and semicontraction theory with application to network systems," 2020. [Online]. Available: https://arxiv.org/abs/2005.09774
- [17] N. Kuznetsov and V. Reitmann, Attractor Dimension Estimates for Dynamical Systems: Theory and Computation. Dedicated to Gennady Leonov. Cham, Switzerland: Springer, 2021.
- [18] P. Lancaster and H. K. Farahat, "Norms on direct sums and tensor products," *Mathematics of Computation*, vol. 26, no. 118, pp. 401–414, 1972.
- [19] G. Leonov, I. M. Burkin, and A. I. Shepeljavyi, Frequency Methods in Oscillation Theory. Springer, 1996.
- [20] D. W. Lewis, Matrix Theory. World scientific, 1991.
- [21] M. Y. Li and J. S. Muldowney, "On R. A. Smith's autonomous convergence theorem," *Rocky Mountain J. Math.*, vol. 25, no. 1, pp. 365–378, 1995.
- [22] M. Y. Li and L. Wang, "A criterion for stability of matrices," J. Math. Anal. Appl., vol. 225, pp. 249–264, 1998.
- [23] W. Lohmiller and J.-J. E. Slotine, "On contraction analysis for non-linear systems," *Automatica*, vol. 34, pp. 683–696, 1998.
- [24] —, "Nonlinear process control using contraction theory," *AIChE Journal*, vol. 46, no. 3, pp. 588–596, 2000.
- [25] I. R. Manchester and J.-J. E. Slotine, "Combination properties of weakly contracting systems," 2014. [Online]. Available: https: //arxiv.org/abs/1408.5174
- [26] —, "Transverse contraction criteria for existence, stability, and robustness of a limit cycle," *Systems and Control Letters*, vol. 63, 2014.
- [27] I. R. Manchester, J. Z. Tang, and J.-J. E. Slotine, "Unifying robot trajectory tracking with control contraction metrics," in *Robotics Research: Volume 2*, A. Bicchi and W. Burgard, Eds. Springer International Publishing, 2018, pp. 403–418.
- [28] M. Margaliot, E. D. Sontag, and T. Tuller, "Entrainment to periodic initiation and transition rates in a computational model for gene translation," *PLoS ONE*, vol. 9, no. 5, p. e96039, 2014.
- [29] —, "Contraction after small transients," Automatica, vol. 67, pp. 178– 184, 2016.
- [30] M. Margaliot and E. D. Sontag, "Revisiting totally positive differential systems: A tutorial and new results," *Automatica*, vol. 101, pp. 1–14, 2019.
- [31] M. Margaliot, T. Tuller, and E. D. Sontag, "Checkable conditions for contraction after small transients in time and amplitude," in *Feedback Stabilization of Controlled Dynamical Systems: In Honor of Laurent Praly*, N. Petit, Ed. Cham, Switzerland: Springer International Publishing, 2017, pp. 279–305.
- [32] J. S. Muldowney, "Compound matrices and ordinary differential equations," *The Rocky Mountain J. Math.*, vol. 20, no. 4, pp. 857–872, 1990.
- [33] Q. C. Pham and J.-J. Slotine, "Stable concurrent synchronization in dynamic system networks," *Neural Networks*, vol. 20, no. 1, pp. 62– 77, 2007.
- [34] A. Pinkus, *Totally Positive Matrices*. Cambridge, UK: Cambridge University Press, 2010.
- [35] A. Y. Pogromsky and H. Nijmeijer, "On estimates of the Hausdorff dimension of invariant compact sets," *Nonlinearity*, vol. 13, no. 3, pp. 927–945, 2000.

- [36] G. Russo, M. di Bernardo, and E. D. Sontag, "Global entrainment of transcriptional systems to periodic inputs," *PLOS Computational Biology*, vol. 6, p. e1000739, 2010.
- [37] —, "A contraction approach to the hierarchical analysis and design of networked systems," *IEEE Trans. Automat. Control*, vol. 58, no. 5, pp. 1328–1331, 2013.
- [38] G. Russo and M. di Bernardo, "Solving the rendezvous problem for multi-agent systems using contraction theory," in *Proc. of the 48th IEEE Conference on Decision and Control held jointly with 28th Chinese Control Conference*. IEEE, 2009, pp. 5821–5826.
- [39] R. G. Sanfelice and L. Praly, "Convergence of nonlinear observers on Rⁿ with a Riemannian metric (part I)," *IEEE Trans. Automat. Control*, vol. 57, no. 7, pp. 1709–1722, 2011.
- [40] D. Schleicher, "Hausdorff dimension, its properties, and its surprises," *The American Mathematical Monthly*, vol. 114, no. 6, pp. 509–528, 2007.
- [41] B. Schwarz, "Totally positive differential systems," Pacific J. Math., vol. 32, no. 1, pp. 203–229, 1970.
- [42] J.-J. E. Slotine, "Modular stability tools for distributed computation and control," Int. J. Adaptive Control and Signal Proc., vol. 17, no. 6, 2003.
- [43] J.-J. E. Slotine and W. Wang, "A study of synchronization and group cooperation using partial contraction theory," in *Cooperative Control*, ser. Lecture Notes in Control and Information Science, V. Kumar, N. Leonard, and A. S. Morse, Eds. Berlin, Heidelberg: Springer, 2005, vol. 309, pp. 207–228.
- [44] R. A. Smith, "Some applications of Hausdorff dimension inequalities for ordinary differential equations," *Proc. Royal Society of Edinburgh: Section A Mathematics*, vol. 104, no. 3-4, pp. 235–259, 1986.
- [45] E. D. Sontag, M. Margaliot, and T. Tuller, "On three generalizations of contraction," in *Proc. 53rd IEEE Conf. on Decision and Control.* IEEE, 2014, pp. 1539–1544.
- [46] T. Strom, "On logarithmic norms," SIAM J. Numerical Analysis, vol. 12, no. 5, pp. 741–753, 1975.
- [47] R. Thomas, "Deterministic chaos seen in terms of feedback circuits: Analysis, synthesis, "labyrinth chaos"," *Int. J. Bifurc. Chaos.*, vol. 9, no. 10, pp. 1889–1905, 1999.
- [48] W. Wang and J.-J. Slotine, "On partial contraction analysis for coupled nonlinear oscillators," *Biological Cybernetics*, vol. 92, 2004.
- [49] E. Weiss and M. Margaliot, "A generalization of linear positive systems with applications to nonlinear systems: Invariant sets and the Poincaré-Bendixson property," *Automatica*, vol. 123, p. 109358, 2021.
- [50] P. Wensing and J.-J. Slotine, "Beyond convexity-contraction and global convergence of gradient descent," *PLoS One*, vol. 15, no. 8, pp. 1–29, 2020.
- [51] J. K. Williams, "A simple example of little big set," *The American Mathematical Monthly*, vol. 100, no. 2, pp. 172–174, 1993.
- [52] S. Winitzki, Linear Algebra via Exterior Products. lulu.com, 2010.
- [53] C. Wu, I. Kanevskiy, and M. Margaliot, "k-order contraction: theory and applications," 2020, submitted. [Online]. Available: https://arxiv.org/abs/2008.10321
- [54] C. Wu and M. Margaliot, "Diagonal stability of discrete-time k-positive linear systems with applications to nonlinear systems," 2020, submitted.
- [55] C. Wu, A. van der Schaft, and J. Chen, "Robust trajectory tracking for incrementally passive nonlinear systems," *Automatica*, vol. 107, pp. 595–599, 2019.