

SYNCHRONIZATION OF FRACTIONAL-ORDER DISCRETE-TIME CHAOTIC SYSTEMS BY AN EXACT DELAYED STATE RECONSTRUCTOR: APPLICATION TO SECURE COMMUNICATION

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This paper deals with the synchronization of fractional-order chaotic discrete-time systems. First, some new concepts regarding the output-memory observability of non-linear fractional-order discrete-time systems are developed. A rank criterion for output-memory observability is derived. Second, a dead-beat observer which recovers exactly the true state system from the knowledge of a finite number of delayed inputs and delayed outputs is proposed. The case of the presence of an unknown input is also studied. Third, secure data communication based on a generalized fractional-order Hénon map is proposed. Numerical simulations and application to secure speech communication are presented to show the efficiency of the proposed approach.

Keywords: fractional-order discrete time systems, chaotic map, chaotic synchronization, dead-beat observer, secure data communication.

1. Introduction

Nowadays, it is well recognized that the concept of fractional calculus has become very popular because of its many applications in the modeling of physical phenomena, system theory and controller design (Monje *et al.*, 2010; Sabatier *et al.*, 2008; Magin, 2004). Fractional-order integration and differentiation operators in the field of continuous-time systems have been mostly successfully used. However, since the work of Miller and Ross (1989), we have witnessed the emergence of fractional-order difference operators in the domain of discrete-time systems. This emergence is demonstrated by

the multitude of works published in the last two decades dealing with the definitions of fractional-order differences operators (Atici and Eloe, 2007; 2009; Abdeljawad and Baleanu, 2009; Holm, 2011; Ortigueira, 2000; Chen *et al.*, 2011; Mozyrska and Pawłuszewicz, 2011), as well as Lyapunov stability (Wyrwas *et al.*, 2015; Guermah *et al.*, 2008a), controllability and observability properties (Mozyrska and Bartosiewicz, 2010; Mozyrska and Pawłuszewicz, 2010; Mozyrska *et al.*, 2015; 2013b; Guermah *et al.*, 2008b; Pawłuszewicz and Mozyrska, 2013; Balachandran and Kokila, 2012) and optimal control (Dzieliński, 2016; Bastos *et al.*, 2011b; Bastos *et al.*, 2011a; Trujillo and Ungureanu, 2018).

Various types of fractional difference operators have

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been introduced such as the Riemann–Liouville type difference operator, the Caputo type difference operator and the Grünwald–Letnikov type operator (Trujillo and Ungureanu, 2018). A more general h -difference operator has also been considered for these three types taking into account the presence of the sampling step h , which is useful in engineering applications (Mozyrska et al., 2015; 2013b; 2013a). In the present paper, the Grünwald–Letnikov type difference operator is considered.

Chaotic behavior of fractional-order systems has been extensively investigated in the literature. In the continuous-time domain, several fractional-order chaotic systems were introduced (see, e.g., the book by Petras (2011) and the references therein). Similarly, the study of the chaotic behavior fractional-order discrete-time systems was considered in the literature. Among the most important applications of fractional-order chaotic maps, one finds the fractional-order logistic map (Wu and Baleanu, 2014b; Munkhammar, 2013) and the fractional-order Hénon map (Liu, 2014; Tarasov, 2010).

Due to their complex dynamics, chaotic systems experience widespread success in secure data communication (Feki et al., 2003; Luo and Wang, 2013; Pareek et al., 2006). In cryptosystems, the transmitter (the drive or master system) and the receiver (the response or slave system) are built as chaotic systems. The information to be secured is masked in the chaotic dynamics at the emitter level and then sent to the receiver through the public channel. At the receiver level, the secret information is decrypted. The use of fractional-order chaotic systems in secure data communication devices is very advantageous. Indeed, compared with integer-order chaotic dynamics, fractional-order chaotic dynamics are more complex because fractional-order derivatives have a complex geometric interpretation due to their non-local character and long memory (Podlubny, 2003). This complexity is an advantage in the design of secure communication schemes since it is difficult for an intruder to extract secret information hidden in fractional-order chaotic dynamics. In addition, the security key space is enlarged by fractional-order parameters, which therefore yields a high level of security.

From this perspective, the synchronization of fractional-order chaotic systems is a central problem. Inspired by the work of Pecora and Carrol (1990) and synchronization methods developed in the case of classical integer-order systems, synchronization of fractional-order chaotic systems in the continuous-time domain has received a lot of attention. Various methods such as the projective method, adaptive control, the impulsive method, active control and sliding mode techniques are proposed (Peng et al., 2014; Shao et al., 2016; Xi et al., 2014; Agrawal et al., 2012; Khanzadeh

and Pourgholi, 2016).

On the other hand, the synchronization of fractional-order discrete-time systems is a very recent research topic despite its natural application in secure digitalized data communication such as image encryption (Zhen et al., 2012; Wu et al., 2016). Some works were devoted to this area (e.g., Wu and Baleanu, 2014a; Liao et al., 2013; Wua et al., 2016). As pointed out by Nijmeijer and Mareels (1997), synchronization may be viewed as an observer design problem. Moreover, in the context of secure communication, the reconstruction of the encrypted message is achieved within an unknown input observer (Sharma et al., 2016). The encrypted message is considered an unknown input. The objective of the unknown input observer is to estimate both the state and the unknown input.

In this paper, we investigate the synchronization of fractional-order discrete-time chaotic systems with the presence of unknown inputs. It is well known that observability is a fundamental property for estimating internal states on the basis of available input/output data (Nijmeijer, 1982; Hanba, 1982; Albertini and D’Alessandro, 2002). Also, the observability matching condition is another property when the system is subject to unknown inputs (Barbot et al., 2002). The problem of synchronization with recovery of the input can be seen as a left invertibility problem (Djemai et al., 2009). The observability of fractional-order discrete-time nonlinear systems described with the Grünwald–Letnikov type difference operator was studied by Mozyrska and Bartosiewicz (2010), who stated a rank condition. Similarly, the first aim of the present paper is to derive some new theoretical results on the observability and the observability matching condition of non-linear fractional-order discrete-time systems described by the Grünwald–Letnikov type difference operator. The concepts of output-memory indistinguishability and output-memory observability are introduced. Furthermore, rank criteria for output-memory observability and an output-memory observability matching condition are proposed.

In most works, synchronization is achieved asymptotically. It has been demonstrated in the case of non-linear integer-order discrete-time systems that synchronization can be immediate and accurate thanks to the use of an exact state reconstructor or a dead beat observer (Belmouhoub et al., 2003; Sira-Ramirez and Rouchon, 2001; Sira-Ramirez et al., 2002). In addition, the immediate recovery of encrypted information is a real challenge. The second contribution of this work is designing an exact state reconstructor for non-linear discrete-time fractional-order systems. The true state is recovered from a finite ring of delayed inputs and delayed outputs. The problem of the exact reconstruction of an unknown input is also considered. Exploiting

these results, we propose a new chaotic encryption process to secure digital data communication based on fractional-order discrete-time chaotic systems. In the proposed secure data communication scheme, a fractional-order version of the generalized Hénon map (Richter, 2002) is used.

The outline of the paper is as follows. Section 2 is devoted to some definitions and preliminary results on fractional-order discrete-time system described by the Grünwald–Letnikov type operator. In Section 3, the concepts of output-memory indistinguishability and output-memory observability are introduced. Rank criteria for output-memory observability and the output-memory observability matching condition are established. The proposed exact delayed state reconstructor of a fractional order is developed in Section 4. With this tool, in Section 5, a secure data communication scheme based on the fractional-order generalized Hénon map is studied. Numerical simulations demonstrating the efficiency of the proposed approach are reported in Section 6. In Section 7, some conclusions and future perspectives end the paper.

2. Preliminaries

Let $h \in \mathbb{R}_+$ be the sampling step and $a \in \mathbb{R}$ a fixed real number. Define the time scale as $(h\mathbb{N})_a = \{a, a+h, a+2h, \dots, a+sh, \dots\}$ with $s \in \mathbb{N}$, and denote by \mathcal{F} the set of real-valued functions. Many fractional-order h -difference operator types have been introduced in the literature, namely, the fractional Riemann–Liouville type difference operator, the Caputo type difference operator and the Grünwald–Letnikov type difference operator (Atici and Eloe, 2007; 2009; Abdeljawad and Baleanu, 2009; Holm, 2011; Ortigueira, 2000; Chen *et al.*, 2011; Mozyrska and Pawłuszewicz, 2011; 2012). In this paper, the Grünwald–Letnikov type- h -difference operator, defined below, is used.

Definition 1. Let $t \in (h\mathbb{N})_a$ and $\alpha \in \mathbb{R}_+$ be a given arbitrary positive real number. The Grünwald–Letnikov type h -difference operator on a function $\nu(t) \in \mathcal{F} : (h\mathbb{N})_a \rightarrow \mathbb{R}$ is defined as

$${}_a\Delta_h^\alpha \nu(t) = \sum_{k=0}^{(t-a)/h} c_k \nu(t-kh), \quad (1)$$

where a denotes the initial time and

$$c_k = \frac{1}{h^\alpha} (-1)^k \binom{\alpha}{k},$$

with

$$\binom{\alpha}{k} = \begin{cases} +1 & \text{for } k = 0, \\ \frac{\alpha(\alpha-1)(\alpha-2)\dots(\alpha-k+1)}{k!} & \text{for } k \geq 1. \end{cases}$$

Remark 1. The Grünwald–Letnikov h -difference operator on a given function $\nu(t)$ for $t \in \mathbb{R}$ is derived from the definition of the Grünwald–Letnikov fractional-order derivative

$${}_{GL}D_t^\alpha \nu(t) = \lim_{h \rightarrow 0} \frac{1}{(h)^\alpha} \sum_{k=0}^{[(t-a)/h]} (-1)^k \binom{\alpha}{k} \nu(t-kh),$$

where $[(t-a)/h]$ represents the integer part of $(t-a)/h$.

With no loss of generality, in what follows, the sampling time h is taken as $h = 1$ and the initial time as $a = 0$. Indeed, in the context of designing a secure data communication scheme, the chaotic system at the emitter level is naturally discrete and is not derived from a sampling process. In this context, the step h is therefore purely contrived. What is important is that the system acting as the emitter exhibits chaotic behavior.

In the works of Bastos *et al.* (2011a) as well as Ferreira and Torres (2011), the fractional difference operator is defined over the time scale $h\mathbb{N}$. This time scale is particularly advantageous in numerical solution of the optimal control problem for fractional-order continuous-time systems, because if h tends to zero, the previous fractional continuous-time results are recovered (Bastos *et al.*, 2011b; Ferreira and Torres, 2011). In addition, as pointed out by Ferreira and Torres (2011), the fractional-order difference operator with a given step h different from 1 can be transformed to an equivalent operator with $h = 1$. For simplicity, we use Δ^α instead of ${}_0\Delta_1^\alpha$, \mathbb{N} instead of $(1\mathbb{N})_0$ and $t \in \mathbb{N}$ as an integer number. If $h = 1$, then $|c_k| \leq \alpha^k/k!$, and thus the sequence (c_k) is absolutely summable. Provided that the function $\nu(t)$ is uniformly bounded, the series (1) converges. It follows that Δ^α is well defined.

Hereafter, the one unit delay operator δ is used in the sense that $\delta\nu(t) = \nu(t-1)$. Its inverse is denoted by δ^{-1} so that $\delta^{-1}\nu(t) = \nu(t+1)$. For any number $s \in \mathbb{N}$, δ^s and δ^{-s} stand for $\delta^s\nu(t) = \nu(t-s)$ and $\delta^{-s}\nu(t) = \nu(t+s)$, respectively. The identity operator is denoted by $\delta^0 = I_d$, i.e., $\delta^0\nu(t) = \nu(t)$. The delayed collection $\{\nu(t-1), \nu(t-2), \dots, \nu(t-s)\}$ is represented by

$$\underline{\delta}^s \nu(t) = \{\delta\nu(t), \delta^2\nu(t), \dots, \delta^s\nu(t)\},$$

and the forward shifts $\{\nu(t), \nu(t+1), \dots, \nu(t+s)\}$ by

$$\underline{\delta}^{-s} \nu(t) = \{\nu(t), \delta^{-1}\nu(t), \dots, \delta^{-s}\nu(t)\}.$$

Definition 2. Let $x(t) \in \mathcal{X} \in \mathbb{R}^n$ be an n -dimensional real-valued vector, where \mathcal{X} denotes a differentiable manifold of dimension n . Denote by x_i , the i -th component of x , $i = 1, 2, \dots, n$, say $x(t) = [x_1(t) \ x_2(t) \ \dots \ x_n(t)]^T$, where $x_i(t) \in \mathcal{F} : \mathbb{N} \rightarrow \mathbb{R}$, $i = 1, 2, \dots, n$, are n real-valued functions. Let

$\alpha_i \in \mathbb{R}_+, i = 1, 2, \dots, n$ be given real positive numbers. The Grünwald–Letnikov type difference operator on the n -dimensional vector $x(t) \in \mathcal{X}$ is defined as

$$\Delta^{[\alpha]}x(t) = [\Delta^{\alpha_1}x_1(t) \quad \Delta^{\alpha_2}x_2(t) \quad \dots \quad \Delta^{\alpha_n}x_n(t)]^T. \quad (2)$$

Similarly, the delay operator on the n -dimensional vector $x(t) \in \mathcal{X}$ is defined as

$$\delta^s x(t) = [\delta^s x_1(t) \quad \delta^s x_2(t) \quad \dots \quad \delta^s x_n(t)]^T. \quad (3)$$

Remark 2. In (2), if there exist a real number $\alpha > 0$ and integers $k_i, i = 1, 2, \dots, n$, such that all orders $\alpha_i, i = 1, 2, \dots, n$, satisfy the relation

$$\alpha_i = k_i \alpha, \quad i = 1, 2, \dots, n,$$

then the fractional-order difference operator $\Delta^{[\alpha]}$ is said to be a commensurate fractional-order difference operator of order α and is denoted simply by Δ^α (Buslowicz, 2008). A particular case of an commensurate fractional-order difference operator is obtained when all orders $\alpha_i, i = 1, 2, \dots, n$, are equal, i.e., $k_i = 1, i = 1, 2, \dots, n$.

Definition 3. Denote by $(\Delta^\alpha)^0$, the identity operator, i.e., $(\Delta^\alpha)^0 x(t) = x(t)$. The ℓ times iterated commensurate fractional-order difference operator of order α on a given n -dimensional real valued vector $x(t)$ for any integer number $\ell \in \mathbb{N}$ is defined as

$$(\Delta^\alpha)^\ell x(t) = \Delta^\alpha (\Delta^\alpha)^{\ell-1} x(t), \quad \ell = 1, 2, \dots \quad (4)$$

or, equivalently,

$$(\Delta^\alpha)^\ell x(t) = \sum_{k=0}^{\ell-1} c_k (\Delta^\alpha)^{\ell-1-k} x(t-k), \quad \ell = 1, 2, \dots, \quad (5)$$

The next lemma provides a useful identity between the Grünwald–Letnikov type difference operator and the delay operator.

Lemma 1. Let $x(t)$ be an n -dimensional vector. For any given $\alpha \in \mathbb{R}_+$ and any given integer $s \in \mathbb{N}$, we have the following commutative properties:

$$\Delta^\alpha (\delta^{-s} x(t)) = \delta^{-s} (\Delta^\alpha x(t)) \quad (6)$$

and, more generally,

$$(\Delta^\alpha)^\ell (\delta^{-s} x(t)) = \delta^{-s} ((\Delta^\alpha)^\ell x(t)), \quad (7)$$

$\ell = 1, 2, \dots$

Proof. From (1), we have

$$\Delta^\alpha x(t) = c_0 x(t) + c_1 x(t-1) + \dots + c_t x(0).$$

It follows that

$$\delta^{-s} \Delta^\alpha x(t) = c_0 x(t+s) + c_1 x(t+s-1) + \dots + c_t x(s). \quad (8)$$

On the other hand, we can write

$$\begin{aligned} \Delta^\alpha \delta^{-s} x(t) &= \Delta^\alpha x(t+s) \\ &= c_0 x(t+s) + c_1 x(t+s-1) + \dots + c_t x(s). \end{aligned} \quad (9)$$

Clearly, (8) and (9) imply

$$\Delta^\alpha (\delta^{-s} x(t)) = \delta^{-s} (\Delta^\alpha x(t)).$$

The relation (7) is obtained by induction. This completes the proof. ■

3. Observability of nonlinear discrete-time fractional-order systems

Observability is a crucial property in systems theory. It informs us about the possibility of constructing a system capable of estimating internal state variables from input-output data. The unavailability of state variables requires the use of reconstruction methods of state variables such as state observers. Observability and observer design of fractional-order systems were investigated by several authors (Kaczorek, 2016; N’Doye et al., 2016). For linear discrete-time systems, the concept of observability was introduced and some observability criteria were proposed (Mozyrska and Bartosiewicz, 2010; Moryrska and Pawłuszkiewicz 2010; Mozyrska et al., 2015; 2013b; Guermah et al., 2008b; Pawłuszewicz and Mozyrska, 2013) and observability with finite initial memory was introduced by Mozyrska and Pawłuszewicz (2010) as well as Mozyrska et al. (2015). However, for fractional-order non-linear discrete-time systems, to the best of our knowledge, only Mozyrska and Bartosiewicz (2010) formally defined the concept of observability.

In the following, this idea is extended, a new concept of *output-memory observability* is introduced, and a new observability criterion for non-linear discrete-time fractional-order systems is proposed.

Consider the following fractional-order nonlinear discrete-time system:

$$\begin{aligned} \Delta^\alpha x(t+1) &= f(x(t), u(t)), \quad x(0) = x_0, \\ y(t) &= h(x(t)), \end{aligned} \quad (10)$$

where $x(t) \in \mathcal{X} \subset \mathbb{R}^n$ is the state vector, x_0 is the initial condition, $u(t) \in \mathcal{U} \subset \mathbb{R}$ stands for the control input and $y(t) \in \mathcal{Y} \subset \mathbb{R}$ signifies the measured output, where \mathcal{X} and \mathcal{Y} are differentiable manifolds of dimension n and 1, respectively, and \mathcal{U} denotes an open control interval in \mathbb{R} . The system (10) is assumed to be of class \mathcal{C}^∞ ,

i.e., the functions $f : \mathcal{X} \times \mathcal{U} \rightarrow \mathcal{X}$ are of class \mathcal{C}^∞ . We assume that the system evolves around an equilibrium point (x_e, u_e, y_e) , $x_e \in \mathcal{X}$ defined as follows.

Definition 4. We say that (x_e, u_e, y_e) is an *equilibrium point* of (10) if

$$\begin{aligned} \Delta^\alpha x_e &= f(x_e, u_e) = x_e, \\ y_e &= h(x_e). \end{aligned} \quad (11)$$

For convenience, we denote by $f_u : \mathcal{X} \rightarrow \mathcal{X}$ the map $f_u(x(t)) = f(x(t), u(t))$ and by $f_{(\Delta^\alpha)^\ell u} : \mathcal{X} \rightarrow \mathcal{X}$ the map $f_{(\Delta^\alpha)^\ell u}(x(t)) = f(x(t), (\Delta^\alpha)^\ell u(t))$, $\ell = 1, 2, \dots$. Moreover, we write f_{u_j} and $f_{(\Delta^\alpha)^\ell u_j}$ for $f(x(t), u(j))$ and $f(x(t), (\Delta^\alpha)^\ell u(j))$, $j = 0, 1, 2, \dots$, respectively. Below, the classical definitions of indistinguishability and observability are revisited in the context of fractional-order non-linear discrete-time systems. More precisely, we introduce the concept of the *output-memory observability* of a state x^0 as the possibility to distinguish the state x^0 from any other state x^j , $j \neq 0$, by applying a finite control sequence $U_k = \{u(0), u(1), \dots, u(k)\}$ and scanning the corresponding memory of the output $Y_{mk} = \{y(0), \Delta^\alpha y(1), \dots, (\Delta^\alpha)^k y(k)\}$.

Definition 5. Two states $x^1, x^2 \in \mathcal{X}$ are said to be *output-memory indistinguishable* if, for each control sequence $U_k = \{u(0), u(1), \dots, u(k)\}$ and for each $k \geq 0$, the produced output memories are identical, i.e.,

$$\begin{aligned} &h \circ f_{(\Delta^\alpha)^k u_k} \circ f_{(\Delta^\alpha)^{k-1} u_{k-1}} \circ \dots \circ f_{\Delta^\alpha u_1} \circ f_{u_0}(x^1) \\ &= h \circ f_{(\Delta^\alpha)^k u_k} \circ f_{(\Delta^\alpha)^{k-1} u_{k-1}} \circ \dots \circ f_{\Delta^\alpha u_1} \circ f_{u_0}(x^2), \end{aligned} \quad (12)$$

where “ \circ ” is the composition operator.

Denote by xIx^0 the set of all states x indistinguishable from x^0 . The concept of *output-memory observability* is defined below.

Definition 6. A state x^0 is said to be *output-memory observable* if $xIx^0 = \{x^0\}$, $\forall x \in \mathcal{X}$.

Remark 3. As for integer-order systems, local weak and strong observability properties can be defined. We say that a state x^0 is locally weakly *output-memory observable* if there exists a neighbourhood $\mathcal{W}_{x^0} \subset \mathcal{X}$ of x^0 such that, for any state $x \in \mathcal{W}_{x^0}$, we have $xIx^0 = \{x^0\}$, $\forall x \in \mathcal{W}_{x^0}$. Similarly, a state x^0 is locally strongly *output-memory observable* if there exists a neighbourhood $\mathcal{W}_{x^0} \subset \mathcal{X}$ of x^0 such that, for each $x^1, x^2 \in \mathcal{W}_{x^0}$, x^1Ix^2 implies that $x^1 = x^2$.

Definition 7. The fractional-order non-linear discrete-time system (10) is *output-memory observable* if each state $x \in \mathcal{X}$ is *output-memory observable*.

For integer-order nonlinear discrete-time systems, the notion of observability is often related to the injectivity of an observation map and the full rank of its Jacobian (Nijmeijer, 1982; Hanba, 1982; Albertini and D’Alessandro, 2002). Mimicking these results, the *output-memory observability* criterion of the fractional-order non-linear discrete-time system (10) is proposed below. Define the observability matrix as follows:

$$\mathcal{O}_{n-1}(x) = \begin{pmatrix} dh(x(t)) \\ dh \circ f_u(x(t)) \\ dh \circ f_{\Delta^\alpha u} \circ f_u(x(t)) \\ \vdots \\ dh \circ f_{(\Delta^\alpha)^{(n-2)}u} \circ \dots \circ f_u(x(t)) \end{pmatrix}, \quad (13)$$

where $d(\dots)$ stands for the standard differential operator with respect to x , i.e., $d\mu(x) = [\frac{\partial \mu(x)}{\partial x_1} \dots \frac{\partial \mu(x)}{\partial x_n}]$ for any real-valued function $\mu(x)$. As in the integer-order case, the dimension $\mathcal{O}_{n-1}(x)$ may be non-constant in the domain considered and may decrease (Albertini and D’Alessandro, 1996; Jakubczyk and Sontag, 1990). A strong result of *output-memory observability* can be derived under the following assumption.

Assumption 1. The observability matrix $\mathcal{O}_{n-1}(x)$ is a constant dimension in a neighbourhood of $\mathcal{W}_{x_e} \in \mathcal{X}$ of an equilibrium point x_e .

Theorem 1. Let Assumption 1 be satisfied. The fractional-order non-linear discrete-time system (10) is locally *output-memory observable* in a neighbourhood $\mathcal{W}_{x_e} \in \mathcal{X}$ of an equilibrium point x_e if and only if the following rank condition holds:

$$\text{rank}(\mathcal{O}_{n-1}(x)) = n \quad (14)$$

for all $x \in \mathcal{W}_{x_e}$.

Proof. The proof consists in constructing an appropriate observability map $\Theta_{n-1}(x)$ from the input and output data. From system equations and Lemma 1, the following relation is obtained:

$$(\Delta^\alpha)^\ell \delta^{-\ell} y(t) = h \circ f_{(\Delta^\alpha)^{(\ell-1)}u} \circ \dots \circ f_u(x(t)),$$

$\ell = 1, 2, \dots$. Then the observability map is constructed as

$$\Theta_{n-1}(x) = \begin{pmatrix} y(t) \\ \Delta^\alpha \delta^{-1} y(t) \\ (\Delta^\alpha)^2 \delta^{-2} y(t) \\ \vdots \\ (\Delta^\alpha)^{n-1} \delta^{-(n-1)} y(t) \end{pmatrix}$$

$$= \begin{pmatrix} h(x(t)) \\ h \circ f_u(x(t)) \\ h \circ f_{\Delta^\alpha u} \circ f_u(x(t)) \\ \vdots \\ h \circ f_{(\Delta^\alpha)^{(n-2)u}} \circ \dots \circ f_u(x(t)) \end{pmatrix}. \quad (15)$$

The observability matrix is given by the co-distribution of the observability map, i.e., $\mathcal{O}_{n-1} = d\Theta_{n-1}$. By Assumption 1, the observability matrix $\mathcal{O}_{n-1}(x)$ is a constant dimension in a neighbourhood of $\mathcal{W}_{x_e} \in \mathcal{X}$, i.e., $\dim \mathcal{O}_{n-1}(x) = n$ for a all $x \in \mathcal{W}_{x_e}$. Then, with this assumption, if the observability matrix $\mathcal{O}_{n-1}(x)$ is of full rank, Eqn. (15) possesses a unique solution $x(t)$. This completes the proof. ■

Remark 4. Note that the observability map $\Theta_{n-1}(x)$ and the observability matrix $\mathcal{O}_{n-1}(x)$ depend on the input u . The fractional-order nonlinear discrete-time system (10) is uniformly *output-memory observable* in a neighbourhood $\mathcal{W}_{x_e} \in \mathcal{X}$ if the rank condition (14) is satisfied for all $u \in \mathcal{U} \subset \mathbb{R}$.

Another property of interest is the possibility of recovering not only the state but also the input assumed to be unknown from the observation of the output. This property is connected to the left-invertibility problem (Djemai et al., 2009). The left-invertibility property is guaranteed by the so-called observability matching condition (Barbot et al., 2002). For this problem, we consider the class of fractional-order discrete-time input affine non-linear system described by

$$\begin{aligned} \Delta^\alpha x(t+1) &= f(x(t)) + g(x(t))u(t), \quad x(0) = x_0, \\ y(t) &= h(x(t)) \end{aligned} \quad (16)$$

We assume that the input $u(t) \in \mathbb{R}$ is unknown. Before addressing the observability matching condition, the relative degree of the fractional-order non-linear discrete-time system (16) should be defined.

Definition 8. The *relative degree* of the fractional-order nonlinear discrete-time system (16) with respect to its input $u(t)$ is the smallest required number r such that the output quantity $(\Delta^\alpha)^r \delta^{-r} y(t)$ depends explicitly on $u(t)$.

Definition 9. We say that the system (16) fulfils the *output-memory observability matching condition* if $y(t), \Delta^\alpha \delta^{-1} y(t), \dots, (\Delta^\alpha)^{(n-1)} \delta^{-(n-1)} y(t)$ does not depend on the unknown input while, in contrast, $(\Delta^\alpha)^{(n)} \delta^{-(n)} y(t)$ depends on the unknown input $u(t)$. Equivalently, the system (16) satisfies the *output-memory observability matching condition* if its relative degree is equal to n .

Accordingly, we have the following proposition.

Proposition 1. *The system (16) satisfies the locally output-memory observability matching condition for $x \in$*

$\mathcal{W}_{x_e} \subset \mathcal{X}$ if

$$\mathcal{O}_{n-1}(x)g(x) = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ \pi(x) \end{pmatrix},$$

$$\forall x \in \mathcal{W}_{x_e} \subset \mathcal{X}, \quad (17)$$

where $\pi(x)$ is any non-vanishing function of $x \in \mathcal{W}_{x_e} \subset \mathcal{X}$.

This *output-memory observability matching condition* allows recovering the unknown input $u(t)$ from a finite sequence of past values of the output $y(t)$.

4. Exact state reconstructor

In this section, we propose an exact delay reconstructor which guarantees an exact recovery of the true state provided that a finite string of past inputs applied and the produced outputs is available. In the first part of this section, we assume that the input is known and only the reconstruction of the state matters. The exact delayed reconstruction of the true state is announced in the following proposition.

Proposition 2. *Assume that the system (10) is output-memory observable in a domain $\mathcal{W}_{x_e} \in \mathcal{X}$ around an equilibrium point (x_e, u_e, y_e) . Then the fractional-order non-linear discrete-time system (10) is reconstructible in \mathcal{W} , i.e., the state $x(t)$ at time t is exactly expressed in terms of the input and output data collections $\underline{\delta}^{n-1} u(t)$ and $\underline{\delta}^{n-1} y(t)$, respectively, provided that these collections are completely known.*

Proof. From Theorem 1, there exists a mapping Ψ such that the state $x(t)$ at each instant t can be expressed by the increments in the outputs and the inputs as

$$x(t) = \Psi(\underline{\delta}^{-(n-1)} y(t), \underline{\delta}^{-(n-2)} u(t)). \quad (18)$$

It follows that

$$\delta^{n-1} x(t) = \Psi(y(t), \underline{\delta}^{(n-1)} y(t), \underline{\delta}^{(n-1)} u(t)), \quad (19)$$

which means that $\delta^{n-1} x(t)$ can be expressed by the past output and input data. On the other hand, from the system equation (10), expanding $\Delta^\alpha x(t+1)$ yields

$$\begin{aligned} x(t+1) &= f(x(t), u(t)) - m_x(t) \\ &= F(x(t), u(t)), \end{aligned} \quad (20)$$

where $m_x(t)$ is the accumulated memory of the state given by

$$m_x(t) = \sum_{j=0}^t (-1)^{j+1} c_{j+1} x(t-j). \quad (21)$$

Iterating (20), we obtain

$$x(t) = F^{(n-1)}(\delta^{n-1}x(t), \underline{\delta}^{(n-1)}u(t)), \quad (22)$$

where $F^{(n-1)}$ obeys the recursion

$$\begin{aligned} F^{(k)}(\cdot) &= F(F^{(k-1)}(\cdot)), \quad \forall k = 2, 3, \dots, \\ F^{(1)}(\cdot) &= F(\cdot) \end{aligned}$$

Substituting (19) on the right-hand side of (22) leads to the following relation:

$$x(t) = \phi(y(t), \underline{\delta}^{(n-1)}y(t), \underline{\delta}^{(n-1)}u(t)). \quad (23)$$

Clearly, the true state $x(t)$ can be exactly reconstructed from the delayed input and output data. This completes the proof. ■

In the second part of this section, we discuss the problem of reconstructing both the system state and input. For this problem, we consider the observable form of (16),

$$\begin{aligned} \Delta^\alpha x(t+1) &= Ax(t) + \Gamma(x(t)) + g(x)u(t), \\ y(t) &= Cx(t). \end{aligned} \quad (24)$$

The system matrices (A, C) take the following Brunowski form:

$$\begin{aligned} A &= \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix}, \\ C &= [1 \ 0 \ \dots \ 0]. \end{aligned} \quad (25)$$

The non-linear vector function $\Gamma(x(t))$ possesses the following triangular structure:

$$\Gamma(x) = \begin{pmatrix} \Gamma_1(x_1) \\ \Gamma_2(x_1, x_2) \\ \Gamma_3(x_1, x_2, x_3) \\ \vdots \\ \Gamma_n(x_1, x_2, x_3, \dots, x_n) \end{pmatrix}, \quad (26)$$

and the vector field $g(x)$ is given by

$$g(x) = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ b_n(x) \end{pmatrix}, \quad (27)$$

with $b_n(x) \neq 0 \forall x(t) \in \mathcal{W}_{x_e} \subset \mathcal{X}$. This means that the unknown input affects only the dynamics of the last variable $x_n(t)$.

In general, a given system is not in this particular form. However, if this system satisfies the *output-memory*

observability condition, this form can be obtained after a suitable change of coordinates (Barbot *et al.*, 2002; Djemai *et al.*, 2009). With this form, the observability map $\Theta_{n-1}(x)$ does not depend on the unknown input u . This implies that the state $x(t)$ in (23) does not depend on the input, i.e.,

$$x(t) = \phi(y(t), \underline{\delta}^{(n-1)}y(t)). \quad (28)$$

The state $x(t)$ is reconstructed uniquely from the output data. The relative degree is equal to n . This means that the input $u(t)$ appears explicitly in $(\Delta^\alpha)^{(n)}\delta^{-(n)}y(t)$, i.e.,

$$(\Delta^\alpha)^{(n)}\delta^{-(n)}y(t) = \tilde{\Gamma}(x(t)) + b_n(x(t))u(t), \quad (29)$$

where $\tilde{\Gamma}(x(t))$ is the compound function of $\Gamma_i(x(t))$, $i = 1, 2, \dots, n$. The relation (28) means that the input depends on the advances of the output. It follows that, if $b(x) \neq 0$, for all $x(t) \in \mathcal{W}_{x_e} \subset \mathcal{X}$, the unknown input $u(t)$ can be exactly recovered after n backward shifts from the reconstructed state $x(t)$ and the past output data.

5. Exact synchronization of the fractional-order chaotic generalized Hénon map for secure communication

The Hénon map (Hénon, 1976) has been widely studied to illustrate the chaotic behaviors of integer-order discrete-time non-linear systems. The map depends on parameters two a and b ($a \neq 0, b \neq 0$), and is described by the two following difference equations:

$$\begin{aligned} x_1(t+1) &= a - x_1^2(t) - bx_2(t), \\ x_2(t+1) &= x_1(t). \end{aligned} \quad (30)$$

A generalized Hénon map was proposed by Richter (2002). It is described by the following difference equations:

$$\begin{aligned} x_1(t+1) &= a - x_{n-1}^2(t) - bx_n(t), \\ x_i(t+1) &= x_{i-1}(t), \quad i = 2, 3, \dots, n. \end{aligned} \quad (31)$$

The equivalent commensurate fractional-order generalized Hénon map is written as

$$\begin{aligned} \Delta^\alpha x_1(t+1) &= a - x_1(t) - x_{n-1}^2(t) - bx_n(t), \\ \Delta^\alpha x_i(t+1) &= x_{i-1}(t) - x_i(t), \quad i = 2, 3, \dots, n. \end{aligned} \quad (32)$$

In what follows, a three-dimensional fractional-order generalized Hénon map is considered to build a secure communication scheme. The transmitter is given by the driven (master) system

$$\begin{aligned} \Delta^\alpha x_1(t+1) &= a - x_1(t) - x_2^2(t) - bx_3(t), \\ \Delta^\alpha x_2(t+1) &= x_1(t) - x_2(t), \\ \Delta^\alpha x_3(t+1) &= x_2(t) - x_3(t) + w(t), \end{aligned} \quad (33)$$

where $w(t)$ represents a confidential message to be securely transmitted to the receiver. This message is modulated into chaotic system dynamics.

5.1. Chaotic behaviour of the fractional-order Hénon map. Chaotic behaviour of discrete-time fractional-order systems has been investigated recently by many authors. Generally, the main tools used to confirm the strange chaotic attractor are the phase portrait, the bifurcation diagram, and the computation of the Lyapunov exponents. It must be emphasized that there are major differences between integer-order chaotic maps and the corresponding fractionalized maps.

Despite the difficulties brought about by the complex aspect of fractional-order systems, these tools have been successfully exploited. For instance, the chaotic behaviour of the fractional-order Hénon map has been proven by using the bifurcation diagram and phase portraits (Liu, 2014; Tarasov, 2010). Similarly in the works of Wu and Baleanu (2014b) as well as Munkhammar (2013), the bifurcation diagram and the phase portrait of the fractional-order logistic map were used to detect the chaotic behaviour. Wu and Baleanu (2015) investigated the Lyapunov exponent analysis for the fractional-order logistic map and the fractional-order delayed logistic map.

As is mentioned by Wu and Baleanu (2015), the computation of the Lyapunov exponents for fractional-order maps is difficult because the Jacobian matrix cannot be directly obtained. Indeed, the current state of the fractional-order system depends on all past states as given by Eqn. (20). Because of this accumulated memory effect, the Jacobian matrix algorithm (Wolf et al., 1985; Eckmann and Ruelle, 1985) is no longer possible. A tangent map based on the Volterra solution was proposed by Wu and Baleanu (2015). Even in this representation, the memory, effect is still present.

In order to make the calculations more accessible, we reduce the length of memory which is called the *short memory principle* in the literature (Podlubny, 1998). It is obvious that the system with the reduced memory is a new one, whose behaviour is different from the original system. The short memory system without the message is given in the following explicit form:

$$\begin{aligned}
 x_1(t+1) &= a + (\alpha - 1)x_1(t) - x_2^2(t) - bx_3(t) \\
 &\quad - \sum_{k=1}^L c_{k+1}x_1(t-k), \\
 x_2(t+1) &= x_1(t) + (\alpha - 1)x_2(t) \\
 &\quad - \sum_{k=1}^L c_{k+1}x_2(t-k),
 \end{aligned}$$

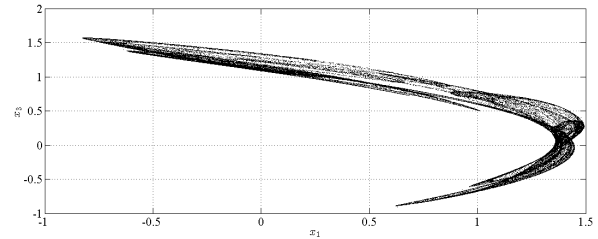


Fig. 1. Strange attractor for the fractional-order generalized discrete Hénon map in the (x_1, x_3) -plane.

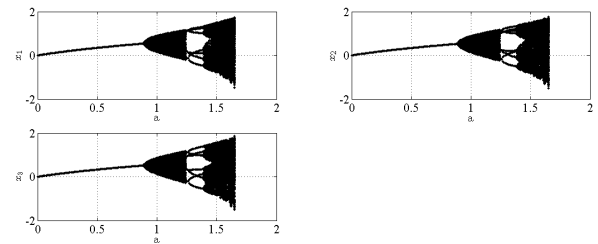


Fig. 2. Bifurcation diagrams versus the parameter a of the fractional-order generalized discrete Hénon map for $\alpha = 0.9, b = 0.1$ and $L = 5$.

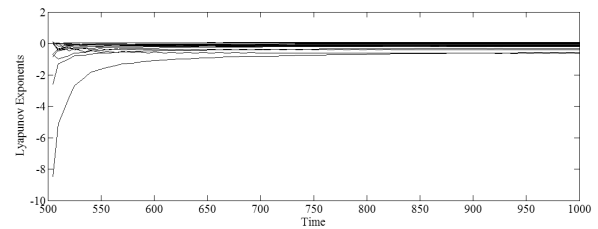


Fig. 3. Lyapunov exponents of the fractional-order generalized discrete Hénon map for $\alpha = 0.9, a = 1.5, b = 0.1$ and $L = 5$.

$$\begin{aligned}
 x_3(t+1) &= x_2(t) + (\alpha - 1)x_3(t) \\
 &\quad - \sum_{k=1}^L c_{k+1}x_3(t-k),
 \end{aligned} \tag{34}$$

where L denotes the memory length.

The use of this new system is highly interesting in the design of secure communication schemes. In fact, as we will see in the next section, the length L of the memory is an additional parameter of the security key. The chaotic behaviour of the new system depends on the parameters but also the length of the memory. Note that the stability and trajectory boundedness depend on both L and α , as reported by Dzieliński and Sierociuk (2008) or Edelman (2018).

In what follows, the chaotic behaviour of the fractional-order map (34) is analyzed. The phase portrait, the bifurcation diagram and the Lyapunov exponents are successively used. The phase portrait is plotted in Fig. 1. The simulations performed for the values of the system parameters taken as $\alpha = 0.9, a = 1.5, b = 0.1$ and

$L = 5$ indicate that the fractional-order system presents chaotic behaviour. Indeed, Fig. 1 plots the strange chaotic attractor of the three-dimensional fractional-order generalized Hénon map on the (x_1, x_3) plane.

Bifurcation diagrams versus the parameter a are plotted in Figs. 2(a)–(c) for $\alpha = 0.9$, $b = 0.1$ and $L = 5$. From these figures, we deduce that the chaos phenomenon appears for $a = 1.5$. Other simulations not reported here show that the chaotic behaviour still appears for higher values of L , while other parameters remain unchanged.

Finally, the Lyapunov exponents method is investigated. To apply the classical Jacobian matrix algorithm (Wu and Baleanu, 2015), the short-memory system (34) for $L = 5$ is rewritten in augmented form as follows. Define new state variables $z_i^j(t) = x_i(t - j)$, $i = 1, 2, 3$ and $j = 1, 2, 3, 4, 5$. Then the system (34) becomes

$$\begin{aligned} z_1^1(t+1) &= a + (\alpha - 1)z_1^1(t) - (z_2^1)^2(t) - bz_3^1(t) \\ &\quad - \sum_{j=1}^5 c_{j+1}z_1^j(t), \\ z_2^1(t+1) &= z_1^1(t) + (\alpha - 1)z_2^1(t) \\ &\quad - \sum_{j=1}^5 c_{j+1}z_2^j(t), \\ z_3^1(t+1) &= z_2^1(t) + (\alpha - 1)z_3^1(t) \\ &\quad - \sum_{j=1}^5 c_{j+1}z_3^j(t), \\ z_i^j(t+1) &= z_i^{j-1}(t), \quad i = 1, 2, 3, \quad j = 2, 3, 4, 5. \end{aligned}$$

Figure 3 plots the Lyapunov exponents for $\alpha = 0.9$, $a = 1.5$, $b = 0.1$ and $L = 5$. The eighteen Lyapunov exponents are 0.079277, 0.035635, -0.0045199 , -0.045184 , -0.062916 , -0.089242 , -0.093286 , -0.10464 , -0.13501 , -0.16417 , -0.17261 , -0.1804 , -0.18501 , -0.19651 , -0.32537 , -0.39704 , -0.572424 , -0.6341 . The system possesses two positive Lyapunov exponents. Then, it is hyperchaotic. This confirms the chaotic behaviour.

5.2. Observability and observer design. We use single channel communication. The only information to be transmitted to the receiver is taken as $y(t) = x_2(t)$. In order to develop the observer, the *output-memory observability* property and the *output-memory observability matching condition* should be satisfied. The system (33) with the output $y(t) = x_2(t)$ is written as

$$\begin{aligned} \Delta^\alpha x(t+1) &= f(x(t)) + g(x(t))w(t), \\ y(t) &= h(x), \end{aligned}$$

where $x(t) = [x_1(t) \ x_2(t) \ x_3(t)]^T$, $h(x) = x_2(t)$ and

$$\begin{aligned} f(x) &= \begin{pmatrix} a - x_1(t) - x_2^2(t) - bx_3(t) \\ x_1(t) - x_2(t) \\ x_2(t) - x_3(t) \end{pmatrix}, \\ g(x) &= \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}. \end{aligned}$$

From the relations

$$\begin{aligned} y(t) &= h(x(t)) = x_2(t), \\ \Delta^\alpha \delta^{-1}y(t) &= h \circ f(x(t)) = \Delta^\alpha x_2(t+1) \\ &= -x_2(t) + x_1(t), \\ \Delta^{\alpha(2)} \delta^{-2}y(t) &= h \circ f^{(2)}(x(t)) \\ &= -\Delta^\alpha x_2(t+1) + \Delta^\alpha x_1(t+1) \\ &= -2x_1(t) + a + x_2(t) - x_2^2(t) - bx_3(t), \end{aligned}$$

the observability matrix of (33) is deduced as

$$\begin{aligned} \mathcal{O}_2(x) &= \begin{pmatrix} dh(x(t)) \\ dh \circ f(x(t)) \\ dh \circ f^{(2)}(x(t)) \end{pmatrix} \\ &= \begin{bmatrix} 0 & 1 & 0 \\ 1 & -1 & 0 \\ -2 & 1 - 2x_2 & -b \end{bmatrix}. \end{aligned} \quad (35)$$

Since $\det(\mathcal{O}_2(x)) = b \neq 0$, the system (33) is *output-memory observable* for all $x(t) \in \mathbb{R}^3$. The *output-memory observability matching condition* can also be checked. Indeed, we have

$$\mathcal{O}_2(x)g(x) = \begin{bmatrix} 0 \\ 0 \\ -b \end{bmatrix}.$$

Since $b \neq 0$, the *output-memory observability matching condition* is satisfied for all $x(t) \in \mathbb{R}^3$. As mentioned by Nijmeijer and Mareels (1997), synchronization of a chaotic system can be formulated as an observer design problem. The receiver is described by the following response (slave) system:

$$\begin{aligned} \Delta^\alpha \hat{x}_1(t+1) &= a - \hat{x}_1(t) - y^2(t) - b\hat{x}_3(t), \\ \Delta^\alpha \hat{x}_2(t+1) &= \hat{x}_1(t) - y(t), \\ \Delta^\alpha \hat{x}_3(t+1) &= y - \hat{x}_3(t), \end{aligned} \quad (36)$$

where \hat{x}_1 , \hat{x}_2 and \hat{x}_3 are the estimates of the driven state variables x_1 , x_2 and x_3 , respectively.

Without any additional control signal injection, except the transmitted signal $y(t)$, we show that this observer permits us to exactly reconstruct the true variables $x_i(t)$, $i = 1; 2, 3$, and to find the original

message $w(t)$. Subtracting (36) from (33) yields the dynamic error given by

$$\begin{aligned} \Delta^\alpha e_1(t+1) &= -e_1(t) - be_3(t), \\ \Delta^\alpha e_2(t+1) &= e_1(t), \\ \Delta^\alpha e_3(t+1) &= -e_3(t) + w(t), \end{aligned} \quad (37)$$

where $e_i(t) = x_i(t) - \hat{x}_i(t)$, $i = 1, 2, 3$, are the estimation errors. Since $y(t) = x_2(t)$ is known from the receiver, $e_2(t)$ is known at each instant t . We can then consider $e_2(t)$ to be an output of the estimation errors system (37), i.e.,

$$e_y(t) = e_2(t) = x_2(t) - \hat{x}_2(t) = y(t) - \hat{y}(t). \quad (38)$$

The objective is to find the secret message $w(t)$. By applying the exact delayed reconstructor developed in the previous section for (37), the errors $e_1(t)$ and $e_3(t)$ are determined from $e_3(t)$, and the message $w(t)$ is also found. The error system (37) is written as

$$e_1(t+1) = (\alpha - 1)e_1(t) - be_3(t) - m_{e_1}(t-1), \quad (39)$$

$$e_y(t+1) = e_1(t) - m_{e_y}(t), \quad (40)$$

$$e_3(t+1) = (\alpha - 1)e_3(t) - m_{e_3}(t-1) + w(t), \quad (41)$$

with

$$m_{e_i}(t-1) = \sum_{k=1}^t c_{k+1}e_i(t-k), \quad i = 1, 3,$$

$$m_{e_y}(t) = \sum_{k=0}^t c_{k+1}e_y(t-k).$$

From (40), we have

$$e_1(t) = e_y(t+1) + m_{e_y}(t). \quad (42)$$

Since $e_y(t+1)$ is not available, we must delay (42) by one time step; then

$$e_1(t-1) = e_y(t) + m_{e_y}(t-1). \quad (43)$$

This means that $e_1(t)$ can be known after one backward shift. This allows us to reconstruct the state variable $x_1(t)$ after one backward shift. Applying two backward shifts for (39), we get

$$\begin{aligned} e_1(t-1) &= (\alpha - 1)e_1(t-2) - be_3(t-2) \\ &\quad - m_{e_1}(t-3). \end{aligned} \quad (44)$$

Since $e_1(t-1)$, $e_1(t-2)$ and, obviously, $m_{e_1}(t-3)$ are known, $e_3(t-2)$ can be obtained. This allows us to reconstruct $x_3(t)$ after two backward shifts. Now, applying three backward shifts to the last equation (41), we get

$$e_3(t-2) = (\alpha - 1)e_3(t-3) - m_{e_3}(t-4) + w(t-3). \quad (45)$$

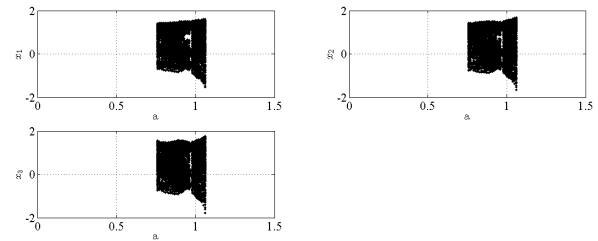


Fig. 4. Bifurcation diagrams versus the order α of the fractional-order generalized discrete Hénon map for $a = 1.5$, $b = 0.1$ and $L = 5$.

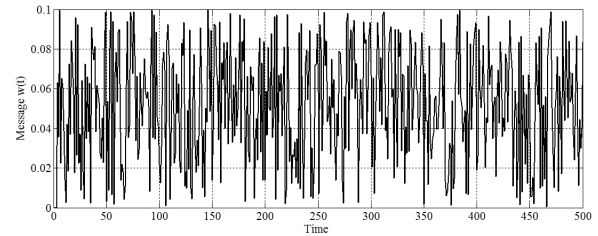


Fig. 5. Message $w(t) = 0.1\text{rand}(500, 1)$.

Again here, since $e_3(t-2)$, $e_3(t-3)$ and $m_{e_1}(t-4)$ are known, it follows that $w(t-3)$ can be recovered. Accordingly, one has to wait three steps to receive the transmitted message. Of course, the first three bits of the message are lost. In practice, in order to avoid any loss of information, meaningless information which is three samples long is added at the beginning of the message.

Remark 5. In the study of *output-memory observability* and *deadbeat observer design*, the results are established for any value of the parameter α . On the other hand, the behaviour of the fractional-order Hénon system depends, of course, on this parameter. Indeed, for some values of α the system can converge towards a fixed point, while for other values of α it can become unstable. The most important in the application of the chaotic system in the design of secure communication schemes, which is our goal, is that the system exhibits chaotic behaviour whose strange attractor can be reached for initial conditions judiciously chosen in the basin of attraction.

To illustrate this, we give, in Fig. 4, the bifurcation diagram versus the parameter α for $L = 5$, $a = 1.5$, $b = 0.1$. As we can see, the system presents chaotic behaviour for α between 0.76 and 1.1. However, for $\alpha < 0.76$ and $\alpha > 1.1$, the system becomes unstable. In the case of fractional linear discrete systems, a study of the stability as a function of the memory length L and of the order α has been proposed by Dzieliński and Sierociuk (2008). In the case of non-linear systems, although stability has been the subject of some studies (e.g., Wyrwas et al., 2015), the investigation of stability with respect to the parameters α and L remains, to our knowledge, an open problem.

6. Numerical simulations

The efficiency of the proposed technique of encryption is illustrated by the following simulation results.

6.1. Example 1: Signal reconstruction. In this first example, the message $w(t)$ is generated randomly and is represented in Fig. 5. The amplitude of the signal $w(t)$ is chosen sufficiently small in order to maintain the chaotic behaviour of the fractional-order generalized Hénon map as depicted in Fig. 6, where the chaotic attractor is plotted in the (x_1, x_3) plane. As pointed out in the previous section, the secret message $w(t)$ is injected in the dynamics of the third variable $x_3(t)$, while the transmitted signal is the second variable $x_2(t) = y(t)$. The signal $y(t)$ transmitted via the public channel is reported in Fig. 7, which illustrates the unintelligibility character. The effectiveness of the synchronization between the transmitter and the receiver achieved by the proposed dead-beat observer is illustrated in Figs. 8 and 9 where the true variables x_1 and x_3 and the reconstructed variables \hat{x}_1 and \hat{x}_3 are plotted. It is shown that the first variable x_1 is reconstructed exactly after one-unit delay and the third variable x_3 is reconstructed after two units of delay. Then, the synchronization is exactly successful after two-unit delay. The decrypted message $\hat{w}(t)$ is plotted in Fig. 10. It is clear that the message is recovered exactly after three units of delay corresponding to the dimension of the chaotic system. It is important to note that the message was reconstructed accurately in finite time after only three units of delay. This is an advantage over other methods, where the message is estimated asymptotically and with an estimation error converging to zero as t goes to infinity. This property is interesting in the design of secure speech communication as illustrated in the second example below.

6.2. Example 2: Speech encryption and decryption. The proposed deadbeat observer can be really used for secure speech communication scheme design as basically depicted in Fig. 11. The confidential digitalized speech signal is masked by inclusion in the fractional-order chaotic map. The encrypted drive signal is sent to the receiver throughout the public channel. The proposed deadbeat observer decrypts and recovers the original speech signal.

The following test is performed using MATLAB. The speech signal illustrated in Fig. 12 is sampled at a frequency of 44.1 kHz. The encrypted drive signal is illustrated in Fig. 13. The decrypted signal given in Fig. 14 is very similar to the original speech. The proposed speech encryption scheme permits us to recover the original signal very quickly with good performance.

The robustness of a transmission scheme against attacks by intruders depends on the space of the

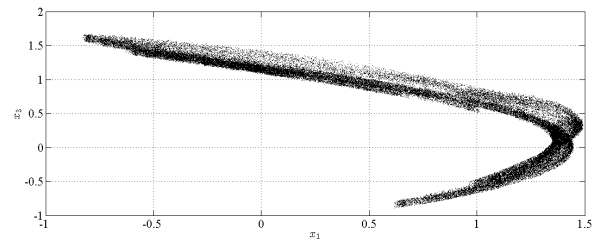


Fig. 6. Strange attractor for the fractional-order generalized discrete Hénon map in the (x_1, x_3) -plane in the presence of the input $w(t) = 0.1\text{rand}(500, 1)$.

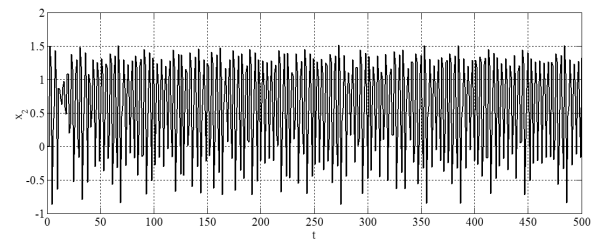


Fig. 7. Signal transmitted x_2 via the public channel.

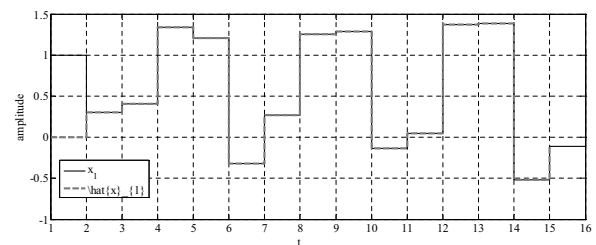


Fig. 8. True state x_1 (solid line) and reconstructed signal \hat{x}_1 (dashed line).

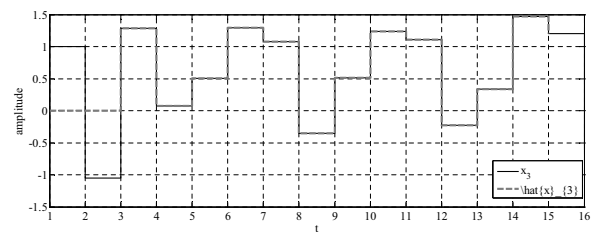


Fig. 9. True state x_3 (in continuous line) and reconstructed signal \hat{x}_3 (in dashed line)

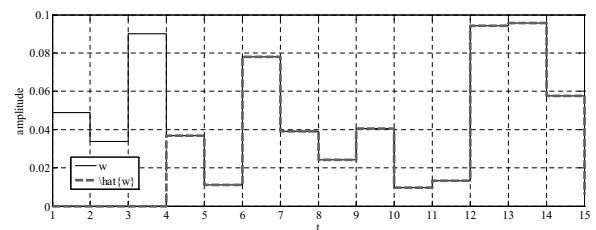


Fig. 10. Original message $w(t)$ (solid line) and reconstructed message $\hat{w}(t)$ (dashed line).

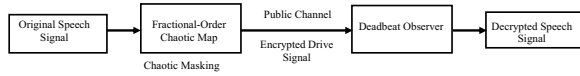


Fig. 11. Block diagram of the speech encryption scheme based on the fractional-order chaotic map.

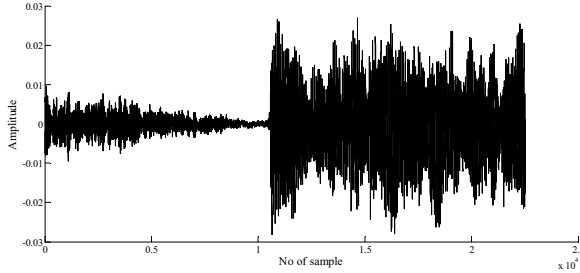


Fig. 12. Histogram of the original speech signal.

security key, which is evaluated by the number of secret parameters constituting this key. The greater the number of these parameters, the higher the transmission scheme. In an integer-order chaotic system based transmission scheme, the parameters of the security key are the parameters of the chaotic system and the initial conditions of the state variables. An intruder who attempts to retrieve secret information must determine these parameters by any means.

The use of fractional order chaotic systems improves the level of security through the fact that the orders of the fractional difference operators are considered additional parameters of the security key. In the case of secure transmission based on integer-order chaotic systems, methods of breaking the key have been proposed, such as as the geometric one (Orue *et al.*, 2008). However, to the best of our knowledge, no method of identifying the orders of fractional difference operators and the length of the memory has been implemented.

In our case, additional parameters constitute the space of the key in addition to the other parameters already used in the integer-order case. These parameters are the fractional orders of the fractional difference operators and the length of the memory. The key space space consists of eight numbers ($\alpha_1 = 0.9, \alpha_2 = 0.9, \alpha_3 = 0.9, L_1 = 5, L_2 = 5, L_3 = 5, a = 1.5, b = 0.1$), where α_i and L_i are the order of the fractional difference operator and the length of the memory on the state variable $x_i, i = 1, 2, 3$, respectively.

To be able to recover secret information, it is necessary to know the exact values of these parameters. Thus, breaking the security key is a complex, if not impossible, task. To demonstrate this, we study the sensitivity of the proposed secure speech communication scheme to the key parameters. Suppose that an intruder gets in some way approximate values of the parameters of the key. Denote by $(\tilde{\alpha}_1, \tilde{\alpha}_2, \tilde{\alpha}_3, \tilde{L}_1, \tilde{L}_2, \tilde{L}_3, \tilde{a}, \tilde{b})$ these

approximate values.

In Figs. 16(a), 16(b), 17(a), 17(b), the following intruder's parameters: $(\tilde{\alpha}_1 = 0.901, \tilde{\alpha}_2 = 0.9, \tilde{\alpha}_3 = 0.9, \tilde{L}_1 = 5, \tilde{L}_2 = 5, \tilde{L}_3 = 5, \tilde{a} = 1.5, \tilde{b} = 0.1), (\tilde{\alpha}_1 = 0.899, \tilde{\alpha}_2 = 0.9, \tilde{\alpha}_3 = 0.9, \tilde{L}_1 = 5, \tilde{L}_2 = 5, \tilde{L}_3 = 5, \tilde{a} = 1.5, \tilde{b} = 0.1), (\tilde{\alpha}_1 = 0.9, \tilde{\alpha}_2 = 0.9, \tilde{\alpha}_3 = 0.9, \tilde{L}_1 = 4, \tilde{L}_2 = 5, \tilde{L}_3 = 5, \tilde{a} = 1.5, \tilde{b} = 0.1)$, and $(\tilde{\alpha}_1 = 0.9, \tilde{\alpha}_2 = 0.9, \tilde{\alpha}_3 = 0.9, \tilde{L}_1 = 6, \tilde{L}_2 = 5, \tilde{L}_3 = 5, \tilde{a} = 1.5, \tilde{b} = 0.1)$ are considered, respectively. As depicted in these figures, for a slight mismatch with the real values of the key parameters, the original speech message is not recovered. Then the proposed fractional-order chaotic system based communication scheme guarantees high security.

To analyze robustness with respect to the noise of the transmission channel, we assumed that the emitted output sent to the receiver is corrupted by white Gaussian noise and is written as

$$y_n(t) = x_2(t) + b(t).$$

where $b(t)$ denotes the white Gaussian noise. Figure 17 shows that, for a signal-to-noise ratio of 40 dB, the message could not be reconstructed. Indeed, the observer used is very sensitive to noise because the reconstruction of the states and the secret message is done directly from the noisy output without any filtering process. This is a major drawback of the deadbeat observer. A design of a fractional-order deadbeat Kalman filter would be more appropriate. This is an interesting perspective to consider in future work.

7. Conclusion

In this paper, first, some theoretical results on *output-memory observability* and the *output-memory observability matching condition* for nonlinear fractional-order discrete-time systems were introduced. Second, an exact state reconstructor which can be considered a dead beat observer for this class of systems was proposed. Third, an application to the design of a secure speech communication scheme based on a fractional-order chaotic map was developed. The benefit of the new secure data transmission scheme comes from the fact that the secret message is reconstructed accurately and in finite time after a certain number of units of delay equal to the dimension of the chosen chaotic system. As an example, an application to speech encryption was given. However, to evaluate the security performance of the proposed speech encryption system, important features with standard quantitative measure such as statistical analysis, sensitivity analysis, key space analysis, robustness to channel noise and speed performance should be investigated. Future works in these directions are in progress.

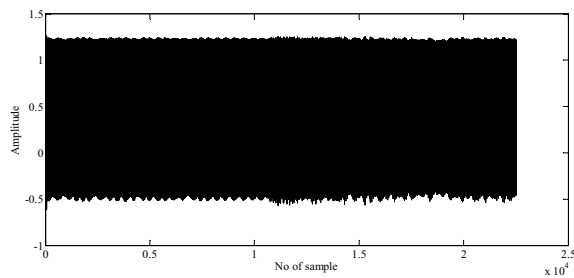


Fig. 13. Histogram of the drive signal.

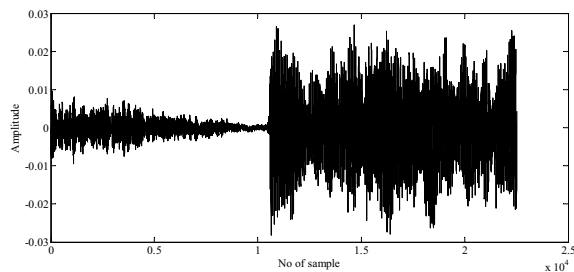


Fig. 14. Histogram of the decrypted speech signal.

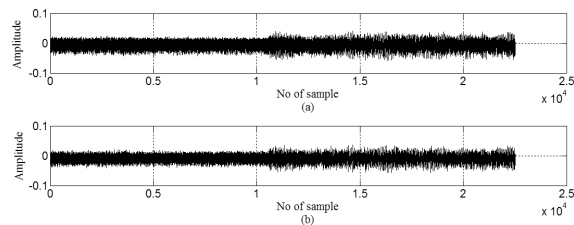


Fig. 15. Key sensitivity analysis: histogram of the decrypted speech signal.

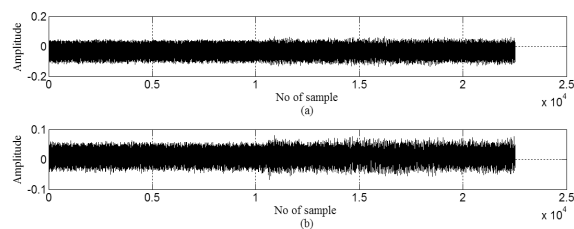


Fig. 16. Key sensitivity analysis: histogram of the decrypted speech signal.

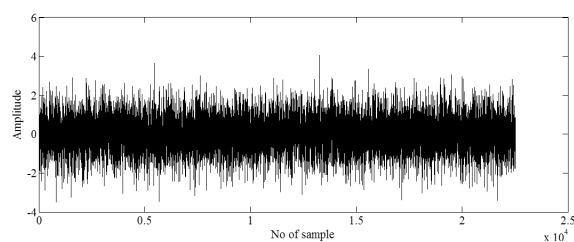


Fig. 17. Histogram of the decrypted speech signal in the presence of a noise signal in the channel.

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