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Guide on Set Invariance for Delay Difference Equations✩

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Abstract

This paper addresses set invariance properties for linear time-delay systems. More precisely, the first goal of the article is to review known necessary and/or sufficient conditions for the existence of invariant sets with respect to dynamical systems described by linear discrete time-delay difference equations (dDDEs). Secondly, we address the construction of invariant sets in the original state space (also called $\mathcal{D}$-invariant sets) by exploiting the forward mappings. The notion of $\mathcal{D}$-invariance is appealing since it provides a region of attraction, which is difficult to obtain for delay systems without taking into account the delayed states in some appropriate extended state space model.

The present paper contains a sufficient condition for the existence of ellipsoidal $\mathcal{D}$-contractive sets for dDDEs, and a necessary and sufficient condition for the existence of $\mathcal{D}$-invariant sets in relation to linear time-varying dDDE stability. Another contribution is the clarification of the relationship between convexity (convex hull operation) and $\mathcal{D}$-invariance of linear dDDEs. In short, it is shown that the convex hull of the union of two or more $\mathcal{D}$-invariant sets is not necessarily $\mathcal{D}$-invariant, while the convex hull of a non-convex $\mathcal{D}$-invariant set is $\mathcal{D}$-invariant.

Keywords: Set invariance, Linear time-delay systems, Discrete time-delay difference equations.

1. Introductory remarks

Positive invariance is an essential concept in control theory, with applications to constrained dynamical systems analysis, uncertainty handling as well as related control design problems [1, 2, 3]. It serves as a basic tool in many topics, such as model predictive control [4, 5, 6], fault tolerant control [7] and reference governor design [8]. Furthermore, there exists a close link between classical stability theory and positive invariant sets. It is worth mentioning that, in Lyapunov theory, invariance is implicitly described by the sub-level sets of a Lyapunov function, which are known to be contractive sets [9].

The response of a dynamical system to external excitation is rarely instantaneous, and time-delay models are well suited for describing dynamics related to propagation phenomena and/or communication flows (see, for example, [10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23]). In closed loop, the dynamics can be represented by delay differential equations (resp. inclusions) or delay difference equations (resp. inclusions) according to the continuous/discrete framework and the presence of disturbances or uncertainties. In the present paper, we consider autonomous dynamics where the delayed arguments are treated as a state dependence and not as a perturbation signal.

From a mathematical point of view, delay difference equations form an important modeling class, since most modern controllers are implemented via computers or dedicated embedded systems. They have been widely studied in the literature (see [24, 25, 26, 27]). Difference and differential equations...
with unbounded random delays have been addressed in [28]. Delay difference inclusions DDIs represent also a rich modeling class including networked control systems and uncertain time-delay systems. The relationship between stability of DDIs and the existence of Lyapunov-Krasovskii and Lyapunov-Razumikhin functions has been studied in detail in [29]. Stabilizing controller construction and stability analysis based on Lyapunov-Krasovskii and Lyapunov-Razumikhin functions for DDIs have been proposed therein.

Positive invariance for dynamical systems described by dDDEs has been addressed in [30]. As hinted before, two main approaches exist in the literature dealing with positive invariant sets for discrete time-delay difference equations. The first approach, referred to as Krasovskii approach, relies on the fact that the discrete-time dDDE allows a finite-dimensional extended state space model (this representing a demarcation with respect to the continuous-time counterpart). This extended state space, whose dimension is finite but strongly related to the delay value, leads to an invariant set characterization with respect to an equivalent linear time-invariant model. This concept is well understood and popular in the literature, but it suffers from an increased numerical complexity when delays are relatively large. Lyapunov-Krasovskii and spectral techniques have been also used in [31] to analyze Lyapunov and asymptotic stability.

The second approach, referred to as Razumikhin approach, has been formulated in the '90s and re-investigated in the last decade, to obtain an invariant set for the dDDE in the original state space, which is independent of the delay value. This concept is also denoted as $\mathcal{D}$-invariance, and is often conservative as long as the existence conditions are restrictive. It is worth mentioning that a relaxation of the Lyapunov-Razumikhin conditions has been proposed by [32]. The proposed conditions, which can be verified by solving an LMI problem for linear dDDEs, prove to be necessary and sufficient for asymptotic stability of dDDEs. Furthermore, the obtained relaxed Lyapunov-Razumikhin functions are useful for constructing invariant sets for dDDEs.

It has been recently recognized that $\mathcal{D}$-invariance can be seen as set factorization of an invariant set in the extended state space [33]. It has been established that the extended state space invariance corresponds to a minimal factorization while $\mathcal{D}$-invariance, under the constraints imposed by the dimension of the dDDE, represents the maximal regular ordered factorization. This interesting result opens the way for factorizations which are in between the two representations, by exploiting non-minimal state space equations. In [34], the authors have focused on the maximal factorizations. They have proposed a characterization of the link between the Razumikhin and Krasovskii approaches, by using set factorization. The proposed framework yields a fitting trade-off between the conceptual generality of the extended state space approach and the computational convenience of the $\mathcal{D}$-invariance approach. It has been shown that $\mathcal{D}$-invariance represents a particular realization of a broader family of invariant structures. The relationship between these families of invariant sets has been established via set factorization and conjugacy. In [35], two specific families of controlled $(k, \lambda)$-contractive sets in the augmented state space framework have been characterized and the link between these controlled $(k, \lambda)$-contractive sets and those of the time-delay system has been established in [36].

In [37], a new concept of set invariance with respect to discrete-time linear systems subject to delays has been introduced. A family of sets which represent a sequence of cyclically invariant subsets of the state space was defined and characterized. Basically, the existing algebraic conditions for invariance analysis of linear dynamics have been generalized and conditions for the invariance of a given sequences of sets with respect to linear discrete-time dynamics in the presence of delay have been established. The notion of invariant family of sets has been proposed in [38, 36] to generalize the cyclic invariance concept.

This paper is an extended version of work published in [39], where we addressed the existence of positive invariant sets in the state space of the original dDDE. More precisely, the case of two delays was addressed in the conference paper, while the general case is treated here. $\mathcal{D}$-invariant sets can be seen as invariant sets in both the current and the retarded state space and further related to the stability analysis based on Lyapunov-Razumikhin approach. Sufficient conditions for the existence of a $\mathcal{D}$-invariant set have been first obtained in [40, 41]. Then, a necessary and sufficient characterization for the existence of $\mathcal{D}$-invariant sets has been provided in [42, 43]. Particularly, as far as the construction of $\mathcal{D}$-invariant sets is concerned, we can find a series of results in [44, 45], which will be appropriately recalled in the present.
paper. Recently, [46] has proposed a computationally efficient numerical routine which is necessary to guarantee the existence of \( D \)-invariant sets for the delay difference equations with two delay parameters. This condition covers, for the two delay case, the existing necessary conditions in the literature and proves to reduce considerably the gap with respect to sufficient conditions. In the present work, we provide an interesting example for which the condition in [46] is verified but the existing algorithms fail to construct a \( D \)-invariant set.

As discussed in [47], from the stability point of view a pertinent analysis of \( D \)-invariance can be made in relationship with delay-independent stability. In short, it has been shown that the existence of a diagonal Lyapunov-Krasovskii functional is necessary and sufficient for delay-independent stability. Polyhedral Lyapunov functions have been used for stability and positive invariance analysis of networked control systems in the presence of bounded delays, constant, unknown or time-varying. The problem of finding stability margins has been proved to reduce to a linear programming problem [48].

To summarize, the main objectives of the present paper are resumed as follows: i) an overview of necessary and/or sufficient conditions for the existence of \( D \)-invariant sets for dDDEs with an arbitrary delay value; ii) a sufficient condition for the existence of ellipsoidal \( D \)-invariant sets for dDDEs; iii) the proof of the relationship between time-varying dDDE stability and the existence of \( D \)-invariant sets; iv) the proof of two properties related to convexity and convex operations over \( D \)-invariant sets. Notably, it is established that a dDDE admits a \( D \)-invariant set if and only if it is time-varying delay-independent stable.

This paper is structured as follows. Section 2 presents some preliminary mathematical notions and definitions. Basic properties of \( D \)-invariance concept are addressed in Section 3. In the same section, we present necessary and sufficient conditions for the existence of non trivial sets. The relationship between \( D \)-invariance and stability of dDDEs concludes the section. Algorithmic construction based on set iteration using forward mappings, and some illustrative examples are revisited in Section 4. The concepts of cyclic invariance and the invariant families of sets as well as the relationship with the set factorization are presented in Section 5. Finally Section 6 draws some concluding remarks.

2. Prerequisites

2.1. Notations

We denote by \( \mathbb{R} \), \( \mathbb{R}_+ \), \( \mathbb{Z} \) and \( \mathbb{Z}_+ \) sets of real numbers, non-negative reals, integer numbers and non-negative integers, respectively. For every interval \( \Pi \) of \( \mathbb{R} \) we define \( \mathbb{Z}_\Pi := \mathbb{Z} \cap \Pi \). For an arbitrary set \( A \subseteq \mathbb{R}^n \), \( \text{int}(A) \) denotes the interior of \( A \). \( \mathbb{B}_r^n(0) \) denotes the ball of radius \( r \) in Euclidean norm, centered in the origin of \( \mathbb{R}^n \). We denote by \( 1_n \) the vector of dimension \( n \) with all the entries equal to 1. We denote by \( D, \partial D, \text{ext}(D) \) the open unit disc, the unit circle and the exterior of the closed unit disc respectively. For the matrix pair \((A, B)\), the set of generalized eigenvalues and the Kronecker product are denoted by \( \gamma(A, B) \) and \( A \otimes B \), respectively. \( I_n \in \mathbb{R}^{n \times n} \) and \( 0_{n \times m} \in \mathbb{R}^{n \times m} \) denote the identity and the null matrix, respectively. \( \mathcal{X} \oplus \mathcal{Y} \) denotes the Minkowski sum of sets \( \mathcal{X} \) and \( \mathcal{Y} \), it is defined by:

\[
\mathcal{X} \oplus \mathcal{Y} := \{ z | \exists (x, y) \in (\mathcal{X}, \mathcal{Y}) \text{ such that } z = x + y \}\,.
\]

**Definition 1.** A set \( \mathcal{P} \subseteq \mathbb{R}^n \) is bounded if there exists \( r \in \mathbb{R}_+ \) such that \( \mathcal{P} \subset \mathbb{B}_r^n(0) \); closed if \( \forall x \notin \mathcal{P}, \exists \varepsilon \in \mathbb{R}_+ \text{ such that } \mathbb{B}_\varepsilon^n(x) \cap \mathcal{P} = \emptyset \); compact if it is bounded and closed.

**Definition 2.** A set \( \mathcal{P} \subseteq \mathbb{R}^n \) is a (proper) \( C \)-set if it is convex, compact and includes the origin in its strict interior.

We denote by \( \text{Com}(\mathbb{R}^n) \) and \( \text{Com}(\mathbb{R}^n) \) the space of compact subsets and the space of \( C \)-subsets of \( \mathbb{R}^n \) containing the origin, respectively. The spectrum of a matrix \( A \in \mathbb{R}^{n \times n} \) is the set of the eigenvalues of \( A \), denoted by \( \lambda(A) \), while the spectral radius is defined as \( \rho(A) := \max_{\xi \in \lambda(A)} |\xi| \). The spectral norm will be denoted by \( \sigma(A) \) and is defined as \( \sigma(A) := \sqrt{\rho(A^T A)} \).

2.2. System Dynamics

In the sequel, we will consider discrete time-delay difference equations of the form:

\[
x(k + 1) = \sum_{i=0}^{d} A_i x(k-i) \tag{1}
\]

where \( x(k) \in \mathbb{R}^n \) is the state vector at the time \( k \in \mathbb{Z}_+ \), \( d \in \mathbb{Z}_+ \) is the maximal fixed time-delay, the matrices \( A_i \in \mathbb{R}^{n \times n} \) for \( i \in \mathbb{Z}_{[0,d]} \) and the initial conditions are considered to be given by \( x(-i) = x_{-i} \in \mathbb{R}^n \), for \( i \in \mathbb{Z}_{[0,d]} \).
The following statements hold:

\[ x(k + 1) = \sum_{i=0}^{m} A_i x(k - d_i) \quad (2) \]

is asymptotically stable if \( \forall \epsilon > 0, \exists \delta > 0 \) such that whenever \( \sup_j |x(-j)| \leq \delta, j = \{1, \cdots, m\} \), \( |x(k)| < \epsilon, \forall k \in \mathbb{Z}_+ \) and \( x(k) \to 0 \) when \( k \to \infty \).

**Definition 4.** The dDDE (2) is delay-independently stable if its null solution is stable \( \forall d = [d_0 \cdots d_m] \in (\mathbb{Z}_+)^{m+1} \).

**Definition 5.** The dDDE with time-varying (positive) delay values:

\[ x(k + 1) = \sum_{i=0}^{m} A_i x(k - d_i(k)) \quad (3) \]

is delay-independently stable if its null solution is stable \( \forall d = [d_0(k) \cdots d_m(k)] \in (\mathbb{Z}_+)^{m+1} \).

It is clear that an extended state space representation can be constructed for any given delay realization. For instance, by setting \( \xi(k) = [x(k)^T \cdots x(k - d)^T]^T \), equation (1) can be rewritten as:

\[ \xi(k + 1) = A \xi(k) \]

\[ \xi(k + 1) = A \xi(k) = \begin{bmatrix} A_0 & \cdots & A_{d-1} & A_d \\ I & \cdots & 0 & 0 \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \cdots & I & 0 \end{bmatrix} \xi(k), \quad (4) \]

This class is relevant for modeling several propagation and transmission phenomena. One example is represented by networked control systems (see [49]) where the feedback mechanism is affected by communication delays. These delays are known to degrade the performances and eventually affect stability [50]. We report next, without proof, some well-known results related to asymptotic stability of systems (1) and (4) (see e.g [51]).

**Lemma 1.** The following statements hold:

- System (4) is asymptotically stable if and only if:
  \[ \rho(A) < 1. \]

**Theorem 2.** The following statements are equivalent:

- The delay difference equation (1) is asymptotically stable.
- The system (4) is asymptotically stable.

### 3. D-INVARIANCE PROPERTIES

Let us first consider the generic (nonlinear) discrete-time dynamical system:

\[ x(k + 1) = f(x(k)) \quad (7) \]

where \( x(k) \in \mathbb{R}^n \) is the state vector at time \( k \in \mathbb{Z}_+ \) and the function \( f : \mathbb{R}^n \to \mathbb{R}^n \) is continuous.

**Definition 6.** The set \( \mathcal{P} \subset \mathbb{R}^n \) is said positively invariant for the system (7) if for all \( x(k) \in \mathcal{P} \), \( x(k + 1) \in \mathcal{P} \) for \( k \in \mathbb{Z}_+ \). Alternatively, the set \( \mathcal{P} \subset \mathbb{R}^n \) is positively invariant for (7) if \( f(\mathcal{P}) \subseteq \mathcal{P} \).

**Definition 7.** Given a scalar \( \epsilon \in \mathbb{R}_{(0,1)} \), a set \( \mathcal{P} \subset \mathbb{R}^n \) containing the origin is called \( \epsilon \)-contractive with respect to system (7) if for any \( x(k) \in \mathcal{P} \), \( x(k + 1) \in \epsilon \mathcal{P} \) for \( k \in \mathbb{Z}_+ \).

One can notice from Definitions 6 and 7 that positive invariance is a limit case of \( \epsilon \)-contractivness (it would amount to choosing \( \epsilon = 1 \) in Definition 7). In the sequel, we will come back to these notions and detail analogies and particularities of time-delay systems. The \( \mathcal{D} \)-invariance concept, recalled below, will be widely used throughout this paper for the set-characterization of dDDEs. The notations by [44, 45] will be mainly used in this endeavor.

**Definition 8.** A set \( \mathcal{P} \subset \mathbb{R}^n \) is called \( \mathcal{D} \)-invariant for the system (1) with initial conditions \( x_{-i} \in \mathcal{P} \) for all \( i \in \mathbb{Z}_{[0,d]} \) if the state trajectory satisfies \( x(k) \in \mathcal{P}, \forall k \in \mathbb{Z}_+ \).

**Lemma 3.** [52] The following statements are equivalent:

1. \( \mathcal{P} \subset \mathbb{R}^n \) is \( \mathcal{D} \)-invariant for system (1).
2. \( \mathcal{D} \mathcal{P} \subset \mathcal{P} \)
Several properties fix a set of basic relations between $\mathcal{D}$-invariant sets.

**Proposition 1.** The following properties hold:

1. If $\mathcal{P} \subset \mathbb{R}^n$ is $\mathcal{D}$-invariant then $\alpha \mathcal{P}$ is $\mathcal{D}$-invariant for any $\alpha \in \mathbb{R}_+$.
2. Let $\mathcal{P}_1, \mathcal{P}_2 \subset \mathbb{R}^n$ be two $\mathcal{D}$-invariant sets for (1). Then $\mathcal{P}_1 \cap \mathcal{P}_2$ is a $\mathcal{D}$-invariant set for the same dynamical system.
3. Let $\mathcal{P}_1, \mathcal{P}_2 \subset \mathbb{R}^n$ be two $\mathcal{D}$-invariant sets for (1). The Minkowski sum $\mathcal{P}_1 \oplus \mathcal{P}_2$ is a $\mathcal{D}$-invariant set for the same dynamical system.
4. If the set $\mathcal{P} \subset \mathbb{R}^n$ is $\mathcal{D}$-invariant for the system:
   \[ x(k + 1) = \sum_{i=0}^{d} A_i x(k - i) \]  
   then $\mathcal{P}$ is $\mathcal{D}$-invariant for
   \[ x(k + 1) = \sum_{i=0}^{d} A_i x(k - \tau_i) \]  
   for any $\tau_i \in \mathbb{Z}_+$.
5. If the compact set containing the origin $\mathcal{P}$ is $\mathcal{D}$-invariant, then its convex hull $\text{Conv}(\mathcal{P})$ is $\mathcal{D}$-invariant.
6. If $\mathcal{P}_1, \mathcal{P}_2 \subset \mathbb{R}^n$ are two $\mathcal{D}$-invariant sets for (1), their union $\mathcal{P}_1 \cup \mathcal{P}_2$ is not necessarily $\mathcal{D}$-invariant.
7. The convex hull of the union of $\mathcal{D}$-invariant sets is not necessarily $\mathcal{D}$-invariant.

**Proof.** Properties (1), (2) and (4) were proved in [52]. The proof of properties (3) and (6) is straightforward. For the proof of property (5), one can exploit the relationship: $A_1 \text{Conv}(\mathcal{P}) \oplus A_2 \text{Conv}(\mathcal{P}) = \text{Conv}(A_1 \mathcal{P} \oplus A_2 \mathcal{P}) = \text{Conv}(A_1 \mathcal{P} \oplus A_2 \mathcal{P})$. The first equality is a direct application of the convex hull definition and Minkowski sum properties.

For the second equality, let $\mathcal{P}_1, \mathcal{P}_2 \subset \mathbb{R}^n$, and let $x \in \text{Conv}(\mathcal{P}_1 \oplus \mathcal{P}_2)$, then $x = \sum \lambda_i (x_i + y_i)$ with $x_i \in \mathcal{P}_1$ and $y_i \in \mathcal{P}_2$, $\lambda_i \geq 0$ and $\sum \lambda_i = 1$, then $x = \sum \lambda_i x_i + \sum \lambda_i y_i \subset \text{Conv}(\mathcal{P}_1) \oplus \text{Conv}(\mathcal{P}_2)$. Suppose now that $x \in \text{Conv}(\mathcal{P}_1) \oplus \text{Conv}(\mathcal{P}_2)$ then $x = \sum x_i + \sum \beta_j y_j$, with $\sum \lambda_i = \sum \beta_j = 1$, and $\lambda_i, \beta_j \geq 0$, $x_i \in \mathcal{P}_1, y_j \in \mathcal{P}_2$. Since $\sum \lambda_i \beta_j = \sum \lambda_i \beta_j = 1$ we can write $x = \sum x_i \lambda_i \beta_j (x_i + y_j)$, then $x \in \text{Conv}(\mathcal{P}_1 \oplus \mathcal{P}_2)$. Note that

\[ A_1 \mathcal{P} \oplus A_2 \mathcal{P} \subset \mathcal{P} \implies \text{Conv}(A_1 \mathcal{P} \oplus A_2 \mathcal{P}) \subset \text{Conv}(\mathcal{P}) \]

to conclude that:

\[ A_1 \text{Conv}(\mathcal{P}) \oplus A_2 \text{Conv}(\mathcal{P}) \subset \text{Conv}(\mathcal{P}) \]

In order to check the property (7), consider the system:

\[ x(k+1) = \begin{bmatrix} 0.2 & 0.01 \\ 0 & 0.7 \end{bmatrix} x(k) + \begin{bmatrix} 0.6 & 0 \\ 0.005 & 0.25 \end{bmatrix} x(k-1), \]

then the set

\[ \mathcal{P}_1 = \left\{ x \in \mathbb{R}^2 \middle| \begin{bmatrix} -0.1 \\ -1 \end{bmatrix} \leq x \leq \begin{bmatrix} 0.1 \\ 1 \end{bmatrix} \right\} \]

is $\mathcal{D}$-invariant as well as

\[ \mathcal{P}_2 = \left\{ x \in \mathbb{R}^2 \middle| \begin{bmatrix} -1 \\ -0.1 \end{bmatrix} \leq x \leq \begin{bmatrix} 1 \\ 0.1 \end{bmatrix} \right\} \]

However, the set obtained as convex hull of the union $\mathcal{P}_1 \cup \mathcal{P}_2$, denoted $\mathcal{P} = \text{Conv}(\mathcal{P}_1, \mathcal{P}_2)$, is not $\mathcal{D}$-invariant.

**Remark 1.** Property (7) of Proposition 1 raises a warning on the convex hull operation applied to the union of two or more $\mathcal{D}$-invariant sets, which is not a closed operation over the class of $\mathcal{D}$-invariant sets. However, property (5) of Proposition 1 points out that for one $\mathcal{D}$-invariant operand, the convex hull operation preserves $\mathcal{D}$-invariance. It becomes clear that under the (unfortunately uncheckable) assumption that a $\mathcal{D}$-invariant set exists, an efficient (convexity based) construction will be able to characterize it.

**Remark 2.** The property (4) of Proposition 1 holds also for the limit case $\tau_i = \infty$. As a consequence, if $\mathcal{P} \subset \mathbb{R}^n$ is a $\mathcal{D}$-invariant set containing the origin, then $\mathcal{P}$ is positively invariant with respect to the time invariant linear dynamics:

\[ x(k + 1) = A_0 x(k), \]

\[ x(k + 1) = A_d x(k). \]

Equivalently, $A_0 \mathcal{P} \subset \mathcal{P}, \cdots , A_d \mathcal{P} \subset \mathcal{P}$. The same result holds for a dDDE represented by a partial sum of (1). Note that the second property of Proposition 1 can be generalized. The intersection of a finite or infinite family of $\mathcal{D}$-invariant sets is $\mathcal{D}$-invariant.
The goal of the next subsections is to collect necessary and/or sufficient conditions for the existence of a $D$-invariant set for dDDEs. The existence of a non-degenerate and bounded $D$-invariant set\footnote{It is easy to observe that sets like $\{0\}$ or $\mathbb{R}^n$ are $D$-invariant but they do not satisfy the non-degenerate or boundedness conditions.} is related to the stability of the discrete-time dynamical system (1) affected by delay. It is obvious that asymptotic stability is only a necessary condition for the existence of a $D$-contractive set and stricter conditions have to be imposed for guaranteeing this existence. In the following we enumerate a series of necessary and/or sufficient conditions available in the literature, to the best of our knowledge; whenever possible, we will link the conditions to classical numerical routines for the eigenvalue problems.

### 3.1. Necessary conditions for $D$-invariance

#### 3.1.1. Basic algebraic conditions

**Proposition 2.** [44] Considering the system (1), the existence of a $D$-contractive set $P$ implies that:

1. The spectral radii of the matrices $A_i$ are sub-unitary:
   \[ \rho(A_i) \leq 1, \forall i \in \mathbb{Z}_{[0,d]} \]

2. The spectral radius of the matrix $\left( \sum_{i=0}^{d} A_i \right)$ is sub-unitary:
   \[ \rho \left( \sum_{i=0}^{d} A_i \right) \leq 1. \]

3. The spectral radius of the extended state-space matrix is sub-unitary:
   \[ \rho(A_\xi) \leq 1. \]

Proposition 2 in conjunction with property (4) of Proposition 1 gives a measure of the complexity of establishing necessary and sufficient conditions. Practically, the difficulty is related to the need of testing the spectral radius of the extended state-space matrix for all possible delay realizations.

#### 3.1.2. Alternative algebraic conditions

Alternative necessary conditions were proposed in [53] in terms of asymptotic stability of dDDEs, for the existence of a $D$-contractive set. The main idea is to cover the possible sign combinations for the tuple $A_i$, $i \in \mathbb{Z}_{[0,d]}$: a straightforward task for any value of the delay parameter. In order to simplify the notation, let us introduce the set $S = \{-1,0,1\}$ and $\Delta = [\delta(0), \cdots, \delta(d)]$.

**Proposition 3.** [53] System (1) admits a $D$-contractive set only if:

\[ \rho \left( \sum_{i=0}^{d} \delta(i) A_i \right) \leq 1, \forall \Delta \in S^{d+1}. \] (12)

If a given dDDE does not satisfy the above condition, then it does not admit a $D$-contractive set. [53] shows that the condition derived in Proposition 3 is not sufficient for the existence of a $D$-contractive set, numerical examples being available in this sense.

#### 3.1.3. Specific algebraic conditions for 2 delay dDDEs

For dDDEs with two delay parameters, in order to decrease the conservativeness of the time-domain methods, [46] has used the frequency-domain framework. The $D$-invariance concept was studied, along with its relation to robust asymptotic stability, considered as a strong stability of dDDEs. This notion defines stability with respect to all delay realizations. Due to the incompleteness of the discrete time, the characterization of robust asymptotic stability is not simple. Thus using a more general class of difference equation (precisely the ones that are specified in the continuous-time domain) proved to be useful. In the sequel the concept of strong stability is denoted by delay-independent stability\footnote{also known as stability in the delays.} and it represents the continuous-time counterpart to robust asymptotic stability.

Recently, [46] has provided a computationally efficient numerical condition which is necessary to guarantee the existence of Lyapunov-Razumikhin contractive sets. This test is sufficient for the robust asymptotic stability with respect to the delay parameter and can be employed in the $D$-invariance context. The main result can be summarized in the next theorem.

**Theorem 4.** [46] Assume that $\rho(A_0 + A_1) \leq 1$ and that $d_0 \in \mathbb{R}_+$ and $d_1 \in \mathbb{R}_+$. Then, the system

\[ x(k) = \sum_{i=0}^{1} A_i x(k - d_i) \] (13)
admits a $\mathcal{D}$-contractive set only if $\gamma(U, V) \cap \partial \mathcal{D} = \emptyset$, where

$$U = \begin{pmatrix} 0_{n_2 \times n_2} & I_{n_1} \\ -B_0 & -B_1 \end{pmatrix}, \quad V = \begin{pmatrix} I_{n_1} & 0_{n_2 \times n_2} \\ 0_{n_2 \times n_2} & B_2 \end{pmatrix}$$

$$B_0 = A_0 \otimes A_1^T, B_1 = A_0 \otimes A_0^T + A_1 \otimes A_1^T - I_{n_2},$$

$$B_2 = A_1 \otimes A_0^T.$$ 

(14)

As stated in [52], the condition of Theorem 4 covers the existing necessary conditions for the two delay parameters case. However, we report here an interesting example which points out the possible limitations of this condition.

**Example 1.** [39] Consider system (1) with $d = 1$ and:

$$A_0 = \begin{pmatrix} 0.5 & 0.5 \\ 0 & 0 \end{pmatrix}; \quad A_1 = \begin{pmatrix} 0 & 0.5 \\ -0.5 & 0.5 \end{pmatrix}$$

(16)

For this numerical example, one can compute:

$$\rho(A_0 + A_1) = 0.8660 < 1$$

and

$$\gamma(U, V) = 1.7442 \pm 1.9433i, 0.2558 \pm 0.2850i, 0, 0, inf, inf.$$ The necessary condition by [46] is fulfilled. However, up to the existing constructive routines (see next section) there is no numerical construction able to determine a $\mathcal{D}$-invariant set for this system. □

3.2. Sufficient conditions for $\mathcal{D}$-invariance

The converse problem of establishing sufficient conditions for the existence of $\mathcal{D}$-invariant sets has been stated in [52] with two tests that we recall here for completeness.

**Proposition 4.** [52] The existence of a $\mathcal{D}$-invariant $C$-set $\mathcal{P}$ is guaranteed for the system (1), if one of the following spectral norm based conditions holds:

1. The sum of the spectral norms of $A_i$, for $i \in \mathbb{Z}_{[0,d]}$, is subunitary:

$$\sum_{i=0}^{d} \sigma(A_i) < 1.$$  

2. In the case of nonsingular matrix $A_i$ for $i \in \mathbb{Z}_{[0,d]}$

$$(1 + \sigma(A_0^{-1} A_1) + \cdots + \sigma(A_0^{-1} A_d) ) \sigma(A_0) \leq 1$$

$$\vdots$$

$$(1 + \sigma(A_d^{-1} A_0) + \cdots + \sigma(A_d^{-1} A_{d-1}) ) \sigma(A_d) \leq 1.$$  

**Remark 3.** The sufficient condition (1) can be generalized by replacing the sum of the spectral norms by the sum of any other induced matrix norms.

Proposition 4 concentrates on the spectral norms of the matrices appearing in the dDDE (1). A different approach for establishing sufficient conditions is to exploit the structural properties of specific classes of candidate $\mathcal{D}$-invariant sets. We propose next a contribution in this sense with a sufficient condition for the existence of ellipsoidal $\mathcal{D}$-contractive sets for a dDDE. As it is often the case in this framework, the tests are based on LMIs.

**Theorem 5.** Considering the dynamical system (1), the existence of an ellipsoidal $\mathcal{D}$-invariant set is guaranteed if the following $d + 1$ LMIs hold for some $P = P^T > 0$:

$$\begin{bmatrix} A_0^T P A_0 & A_0^T P A_1 & \cdots & A_0^T P A_d \\ A_1^T P A_0 & A_1^T P A_1 & \cdots & A_1^T P A_d \\ \vdots & \vdots & \ddots & \vdots \\ A_d^T P A_0 & A_d^T P A_1 & \cdots & A_d^T P A_d \end{bmatrix} < 0$$

(18a)

$$\begin{bmatrix} A_0^T P A_0 & A_0^T P A_1 & \cdots & A_0^T P A_d \\ A_1^T P A_0 & A_1^T P A_1 - P & \cdots & A_1^T P A_d \\ \vdots & \vdots & \ddots & \vdots \\ A_d^T P A_0 & A_d^T P A_1 & \cdots & A_d^T P A_d \end{bmatrix} < 0$$

(18b)

$$\begin{bmatrix} A_0^T P A_0 & A_0^T P A_1 & \cdots & A_0^T P A_d \\ A_1^T P A_0 & A_1^T P A_1 & \cdots & A_1^T P A_d \\ \vdots & \vdots & \ddots & \vdots \\ A_d^T P A_0 & A_d^T P A_1 & \cdots & A_d^T P A_d - P \end{bmatrix} < 0$$

(18c)

**Proof.** In order to ensure that the set

$$\Psi = \{ x \in \mathbb{R}^n, x^T P x \leq 1 \}$$

is $\mathcal{D}$-invariant for the system described by the dDDE (1), one has to show that $x_{k+1} \in \Psi$, 7
∀x_k, x_{k-1}, \ldots, x_{k-d} ∈ Ψ, which is equivalent to the simultaneous verification of the d + 1 inequalities:

\[
x^T_{k+1} P x_{k+1} - x^T_k P x_k < 0
\]

\[
x^T_{k+1} P x_{k+1} - x^T_{k-1} P x_{k-1} < 0
\]

\[\vdots\]

\[
x^T_{k+1} P x_{k+1} - x^T_{k-d} P x_{k-d} < 0
\]

Exploiting the dDDE relationship one has:

\[
x^T_{k+1} P x_{k+1} - x^T_k P x_k = (A_0 x_k + A_1 x_{k-1} + \cdots + A_d x_{k-d})^T P (A_0 x_k + A_1 x_{k-1} + \cdots + A_d x_{k-d}) - x^T_k P x_k
\]

\[
= x^T_k (A_0^T P A_0 - P) x_k + x^T_k A_0^T P (A_1 x_{k-1} + \cdots + A_d x_{k-d}) + (A_1 x_{k-1} + \cdots + A_d x_{k-d})^T P (A_0 x_k + A_1 x_{k-1} + \cdots + A_d x_{k-d}) < 0.
\]

and in the equivalent matrix formulation:

\[
\begin{pmatrix}
    x_k \\
    x_{k-1} \\
    \vdots \\
    x_{k-d}
\end{pmatrix}^T
\begin{bmatrix}
    A_0^T P A_0 & A_0^T P A_1 & \cdots & A_0^T P A_d \\
    A_1^T P A_0 & A_1^T P A_1 & \cdots & A_1^T P A_d \\
    \vdots & \vdots & \ddots & \vdots \\
    A_d^T P A_0 & A_d^T P A_1 & \cdots & A_d^T P A_d
\end{bmatrix}
\begin{pmatrix}
    x_k \\
    x_{k-1} \\
    \vdots \\
    x_{k-d}
\end{pmatrix} < 0
\]

(20)

Analogously for the second inequality:

\[
\begin{pmatrix}
    x_k \\
    x_{k-1} \\
    \vdots \\
    x_{k-d}
\end{pmatrix}^T
\begin{bmatrix}
    A_0^T P A_0 & A_0^T P A_1 & \cdots & A_0^T P A_d \\
    A_1^T P A_0 & A_1^T P A_1 & \cdots & A_1^T P A_d \\
    \vdots & \vdots & \ddots & \vdots \\
    A_d^T P A_0 & A_d^T P A_1 & \cdots & A_d^T P A_d
\end{bmatrix}
\begin{pmatrix}
    x_k \\
    x_{k-1} \\
    \vdots \\
    x_{k-d}
\end{pmatrix} < 0
\]

(21)

up to the d + 1 inequality. We can conclude that the existence of a positive definite matrix \( P = P^T \) is a sufficient condition for the existence of an ellipsoidal \( \mathcal{D} \)-invariant set, and the proof is complete. □

Example 2. For illustration let us consider system (1) with only one delay parameter \( d = 1 \) and:

\[
A_0 = \begin{pmatrix} 0.35 & 0.13 \\ 0.51 & -0.01 \end{pmatrix}, A_1 = \begin{pmatrix} 0.51 & -0.01 \\ 0.03 & 0.51 \end{pmatrix}.
\]

(22)

The condition for the existence of a \( \mathcal{D} \)-contractive set proposed in Theorem 5 is fulfilled and the \( \mathcal{D} \)-contractive set exists as shown in Figure 1. Dashed black lines in Figure 1 represent the state trajectories starting from some points on the boundary of the ellipsoidal \( \mathcal{D} \)-contractive set with respect to the dDDE (1) with \( d = 1 \), \( A_0, A_1 \) given in (22). It is interesting to note that the sufficient condition \( \| A_0 \|_p + \| A_1 \|_p \leq 1 \) by [42, 44] does not hold for this numerical example.

Figure 1: \( \mathcal{D} \)-contractive set for the dDDE (1) with \( d = 1 \), \( A_0, A_1 \) given in (22).

3.3. Necessary and sufficient algebraic conditions for Polyhedral \( \mathcal{D} \)-invariant sets

The problem of finding convex \( \mathcal{D} \)-invariant sets can benefit whenever particular structural properties are enforced. It is the case of polyhedral sets, for which necessary and sufficient conditions exist as resumed by the following theorem.

Theorem 6. [54] Let a delay difference equation be described by (1). There exists \( \mathcal{P} \) a polyhedral \( \mathcal{D} \)-contractive set containing the origin:

\[
\mathcal{P} = \{ x \in \mathbb{R}^n \mid F x \leq 1 \}
\]

(23)

with \( F \in \mathbb{R}^{r \times n} \), described by its minimal half space representation, if and only if there exist \( d + 1 \) real matrices \( H_i \in \mathbb{R}^{r \times r} \), for \( i = 0, \cdots, d \), with non-negative elements and a positive \( \epsilon < 1 \), such that:

\[
FA_i = H_i F
\]

(24a)

\[
\left( \sum_{i=0}^d H_i \right) 1_r \leq \epsilon 1_r
\]

(24b)

Clearly, if the requirement on \( \epsilon \) being strictly smaller than 1 is relaxed to non-strict inequality, then (24) represents a necessary and sufficient condition for the existence of a \( \mathcal{D} \)-invariant set.

3.4. Relationship between \( \mathcal{D} \)-invariance and dDDE stability

In this subsection we aim at complementing the overview of the necessary and sufficient conditions with a theoretical result that establishes a link between the stability in presence of time-varying delay and the existence of \( \mathcal{D} \)-invariant sets.
Theorem 7. The dDDE (2) admits a proper \(\mathcal{D}\)-invariant set if and only if the time-varying dDDE (3) is delay-independent stable.

Proof. We prove next the case of dDDE with only two delay parameters, \(x(k+1) = A_0x(k-d_0) + A_1x(k-d_1)\), the case of finite number of delays (2) being a direct generalization. The proof of the "only if" implication builds on the fact that the existence of a \(\mathcal{D}\)-invariant set \(\mathcal{P}\) is equivalent with the set inclusion:

\[
A_0\mathcal{P} \cup A_1\mathcal{P} \subset \mathcal{P}
\]  

(25)

Thus for initial conditions \(x(k) \in \mathcal{P}\) for \(k \in \mathbb{Z}_{(−\infty,0]}\) one has \(x(1) \in \mathcal{P}\) independent of the delay realization \(d_0(0),d_1(0)\) in \(\mathbb{N}\). By induction, given a positive index \(i \in \mathbb{N}\), if \(x(k) \in \mathcal{P}\) for \(k \in \mathbb{Z}_{(−\infty,i]}\) then \(x(i+1) \in \mathcal{P}\) independent of the delay realization \(d_0(i),d_1(i)\) in \(\mathbb{N}\) which implies that the trajectories are bounded \(x(k) \in \mathcal{P}, \forall k \in \mathbb{N}_++\). Stability for any initial condition follows from property (1) of Proposition 1. By homogeneity, \(\mathcal{D}\)-invariance is preserved by scaling and as such, there always exists a \(\mathcal{D}\)-invariant set which contains a given initial condition of the dDDE.

For the "if" part of the proof, consider the initial conditions for the system (3) to be contained in a compact set \(\mathcal{P}\) containing the origin in its interior. Formally, the initial conditions and the time-varying delay realization can be described by the functions:

\[
x^\mathcal{P} : \mathbb{Z}_{(−\infty,0]} \to \mathcal{P}
\]

\[
d_0 : \mathbb{N}_+ \to \mathbb{Z}_{(−\infty,0]}
\]

\[
d_1 : \mathbb{N}_+ \to \mathbb{Z}_{(−\infty,0]}
\]  

(26)  

(27)

Having as an objective the construction of the reachable set from \(\mathcal{P}\), let us denote the state at time instant \(k \in \mathbb{Z}\) by \(x(k,x^\mathcal{P},d_0,d_1)\) as the solution of (3) with respect to the initial conditions \(x^\mathcal{P}\) and time-varying delay realizations \(d_0(\cdot),d_1(\cdot)\). With this notation, the reachable set from \(\mathcal{P}\) via (3) is defined as:

\[
\mathcal{R}(\mathcal{P}) = \{x \in \mathbb{R}^n| \exists k \in \mathbb{N}_+, x^\mathcal{P}(\cdot), d_0(\cdot), d_1(\cdot) \text{ s.t. } x = x(k,x^\mathcal{P},d_0,d_1)\}
\]  

(28)

Coming back to the proof, the objective is to show that \(\mathcal{P}_r = \mathcal{P} \cup \mathcal{R}(\mathcal{P})\) is a proper \(\mathcal{D}\)-invariant set. The fact that the origin is contained in the interior of \(\mathcal{P}_r\) is inherited from the properties of \(\mathcal{P}\). The boundedness of the set \(\mathcal{R}(\mathcal{P})\) is ensured by the stability assumption and will be inherited by \(\mathcal{P}_r\). What remains to be proved is the invariance of \(\mathcal{P}_r\). Three possibilities should be discussed:

\begin{itemize}
  \item \(x(k-d_0(k)) \in \mathcal{P}\) and \(x(k-d_1(k)) \in \mathcal{P}\): in this case the state \(x(k+1)\) is part of the one step reachable set and subsequently \(x(k+1) \in \mathcal{R}(\mathcal{P}) \subset \mathcal{P}_r\).
  \item \(x(k-d_0(k)) \in \mathcal{P}\) and \(x(k-d_1(k)) \in \mathcal{R}(\mathcal{P})\) (with delay indices which can be interchanged): this case corresponds to a reachable state \(x(k-d_1(k)) \in \mathcal{R}(\mathcal{P})\) combined with a large (pseudo-infinite) delay \(d_0(k)\). By consequence the state realizations \(x(k+1)\) will represent a subset of the reachable set and \(\mathcal{R}(\mathcal{P}) \subset \mathcal{P}_r\).
  \item \(x(k-d_0(k)) \in \mathcal{R}(\mathcal{P})\) and \(x(k-d_1(k)) \in \mathcal{R}(\mathcal{P})\) (with delay indices which can be interchanged): again, via reachability \(x(k+1) \in \mathcal{R}(\mathcal{P}) \subset \mathcal{P}_r\) with the particular case \(d_0(k) = d_1(k)\) which deserves a special treatment. Indeed, for the restriction \(d_0(k) = d_1(k)\), the state dynamics (3) reduces to \(x(k+1) = (A_0 + A_1)x(k-d_1(k))\). But this realization is only a particular case of the general time-varying delay realization \(d_0(k) \neq d_1(k)\) for which \(x(k-d_0(k)) = x(k-d_1(k))\) which is covered by the reachable set construction and the proof is complete. \(\square\)
\end{itemize}

Remark 4. The sets containing the forward trajectories, as those used in the argument of the proof, are non-convex and lead to computationally demanding constructions, from a practical point of view. In the next section we describe the corresponding algorithm and subsequently reinforce the convexity by exploiting property (5) of Proposition 1.

4. Construction of \(\mathcal{D}\)-invariant sets based on set iterations

We address now the construction procedures for the case \(x(k+1) = A_0x(k) + A_dx(k-d)\) supposing that it admits a \(\mathcal{D}\)-invariant set. The general form (1) follows similarly. We use the fact that existence of \(\mathcal{D}\)-invariant sets is exactly equivalent, by Lemma (3), to the verification of \(A_0\mathcal{P} \cup A_d\mathcal{P} \subseteq \mathcal{P}\). To simplify the explanation, we first define the forward mapping:

\[
\Phi : \text{Com}(\mathbb{R}^n) \to \text{Com}(\mathbb{R}^n)
\]

\[
\Phi(\mathcal{P}) = A_0\mathcal{P} \cup A_d\mathcal{P}
\]  

(29)

and the mapping based on the union:

\[
\Psi : \text{Com}(\mathbb{R}^n) \to \text{Com}(\mathbb{R}^n)
\]

\[
\Psi(\mathcal{P}) = \bigcup(\mathcal{P}, \Phi(\mathcal{P})).
\]  

(30)
Note that even if $\mathcal{P}$ is convex, $\Psi(\mathcal{P})$ is not necessarily convex.

**Remark 5.** We enumerate here some useful properties of the mappings defined in (29-30):

1. If a given set $\mathcal{P}$ (convex or not) is $\mathcal{D}$-invariant for (1), then $\Phi(\mathcal{P}) \subseteq \mathcal{P}$.
2. $k$-iterates over the family of sets is set-wise non decreasing ($\Psi^{k-1}(\mathcal{P}) \subseteq \Psi^k(\mathcal{P}), \forall k \geq 1$) with $\Psi^k(\mathcal{P}) = \Psi(\Psi^{k-1}(\mathcal{P}))$ for $k > 0$ and $\Psi^0(\mathcal{P}) = \mathcal{P}$.
3. If $\mathcal{P}$ is $\mathcal{D}$-invariant for (1) then $\Phi^k(\mathcal{P})$ is set-wise non increasing ($\Phi^k(\mathcal{P}) \subseteq \Phi^{k-1}(\mathcal{P}), \forall k \geq 1$).

### 4.1. Basic set-iterates procedure for the construction of $\mathcal{D}$-invariant sets

We describe in this part the basic steps of an iterative construction of $\mathcal{D}$-invariant sets. Under the assumption that such an invariant set exists for the system (1), we can always scale it using property (1) of Proposition (1) such that it encompasses the initial set $\mathcal{Q}$. Using the theoretical properties shown above, an algorithmic routine based on non-convex sets mapping is proposed for the computation of $\mathcal{D}$-invariant sets with respect to (1). This algorithm considers as an input argument an arbitrary bounded set $\mathcal{Q}$ containing the origin ([55, 56]).

**Algorithm 1:** Basic (non-convex) set-iterates procedure.

**Data:** A bounded set $\mathcal{Q} \in \mathbb{R}^n$ containing the origin; the matrices $A_0, A_d \in \mathbb{R}^{n \times n}$ describing the system (1)

**Result:** $\mathcal{R}$ a $\mathcal{D}$-invariant set  

$\mathcal{R}_0 = \mathcal{Q}$;  

$\mathcal{R}_1 = \Phi(\mathcal{Q}) = A_0\mathcal{Q} \oplus A_d\mathcal{Q}$;  

$i = 1$;

while $\mathcal{R}_i \not\subset \mathcal{R}_{i-1}$ do 

$\mathcal{R}_{i+1} = \Psi(\mathcal{R}_i) = \bigcup(\mathcal{R}_i, A_0\mathcal{R}_i \oplus A_d\mathcal{R}_i)$;  

$i = i + 1$;

end 

Return $\mathcal{R} = \mathcal{R}_i$  

(Alternatively, $\mathcal{R} = \text{Conv}(\mathcal{R}_i)$ can represent the output if a unique convex set is needed.)

**Convergence and finite determinedness analysis:** First, it can be proved that Algorithm 1 constructs a non-decreasing sequence that converges to a $\mathcal{D}$-invariant set. Indeed, the algorithm is based on the set mapping $\mathcal{R}_{i+1} = \Psi(\mathcal{R}_i)$ which satisfies $\mathcal{R}_{i+1} \supset \mathcal{R}_i$. Thus the sequence $\mathcal{R}_i$ is non-decreasing in the sense of set inclusion. On the other hand, since the $\mathcal{D}$-invariance is scalable (using property (1) of Proposition (1)), the hypothesis of existence of a $\mathcal{D}$-invariant set $\mathcal{P}$ containing $\mathcal{Q}$ ensures $\mathcal{Q} \subset \mathcal{R}_i \subset \mathcal{P}$. Since any set $\mathcal{R}_i$ provided by the algorithm is a subset of $\mathcal{P}$, hence $\Psi(\mathcal{R}_i)$ is also a subset of $\mathcal{P}$. In conclusion, the algorithm provides a sequence of sets $\mathcal{R}_i$ which is non-decreasing by inclusion and limited from above by $\mathcal{P}$. Hence the sequence admits a limit which is $\mathcal{D}$-invariant (by the structure of the algorithm) and proper (because limited from above by $\mathcal{P}$ which is a fixed point with respect to the mapping $\Psi(\cdot)$). Secondly, the finite determinedness can be formally proved. Given the (delay-independent) asymptotic stability of system (1) with matrices $A_0$ and $A_d$, there exists a finite number of time steps $t_{\text{max}}$ such that the trajectories initiated in $\mathcal{Q}$ end up in $\mathcal{P}$. The algorithm is collecting the trajectories initiated in $\mathcal{Q}$, which is a subset of $\mathcal{P}$, and thus $t_{\text{max}}$ represents an upper bound for the number of iterations. This completes the convergence analysis of the algorithm.

Note that the iterations and the limit set are non-convex and this is related to the union operation performed by the mapping in $\Psi(\cdot)$.

**Example 3.** [39] Let us consider the following dynamical system:

$$x(k+1) = \begin{bmatrix} 0.1 & 0 \\ 0.4 & 0.1 \end{bmatrix} x(k) + \begin{bmatrix} 0.1 & -0.2 \\ 0.4 & 0.5 \end{bmatrix} x(k-d),$$

(31)

Consider the initialization set $\mathcal{Q}$ as the $\infty$-norm unit ball in $\mathbb{R}^2$. A non-convex $\mathcal{D}$-invariant set is obtained iteratively by applying Algorithm 1 with 4 iterations.

Figure 2 presents this invariant set (the left one), and the image (the right one) of this set by the forward mapping $\Phi(\cdot)$. Figure 3 presents the Convex
hull of the obtained non-convex \( \mathcal{D} \)-invariant set and shows that it is \( \mathcal{D} \)-invariant as theoretically proved in property (5) of Proposition 1.

Figure 3: Graphical illustration of the convex \( \mathcal{D} \)-invariant set for the Example 3.

4.2. Convex set-iterates procedure for the construction of \( \mathcal{D} \)-invariant sets

We describe briefly in this part the main steps of an iterative construction of \( \mathcal{D} \)-invariant sets while manipulating only convex sets. This algorithmic routine was proposed by [44], but we recall it here in light of Theorem 7 and Algorithm 1. Let us define the two mappings:

\[
\Omega : \text{ComC}(\mathbb{R}^n) \to \text{ComC}(\mathbb{R}^n) \\
\Omega(\mathcal{P}) = A_0 \mathcal{P} \oplus A_d \mathcal{P} \tag{32}
\]

\[
\Xi : \text{ComC}(\mathbb{R}^n) \to \text{ComC}(\mathbb{R}^n) \\
\Xi(\mathcal{P}) = \text{Conv}(\mathcal{P}, \Omega(\mathcal{P})) \tag{33}
\]

Given a convex set \( \mathcal{P} \in \text{ComC}(\mathbb{R}^n) \), the sequence \( \Xi^k(\mathcal{P}), k > 0 \) converges toward a convex \( \mathcal{D} \)-invariant set [44]. The main objective of this procedure remains the same as the previous one: enlarge the set as much as possible with the Convex hull operation, while keeping it included in a \( \mathcal{D} \)-invariant superset.

Algorithm 2: Convex set-iterates converging to a \( \mathcal{D} \)-invariant set.

**Data:** A convex set \( \mathcal{Q} \in \mathbb{R}^n \) containing the origin in the interior; the matrices \( A_0, A_d \in \mathbb{R}^{n \times n} \)

**Result:** \( \mathcal{R} \) Convex \( \mathcal{D} \)-invariant set

\( \mathcal{R}_0 = \mathcal{Q} \);

\( \mathcal{R}_1 = \Phi(\mathcal{Q}) = A_0 \mathcal{Q} \oplus A_d \mathcal{Q} \);

\( i = 1; \)

**while** \( \mathcal{R}_i \not\subset \mathcal{R}_{i-1} \) **do**

\( \mathcal{R}_{i+1} = \Xi(\mathcal{R}_i) = \text{Conv}(\mathcal{R}_i, A_0 \mathcal{R}_i \oplus A_d \mathcal{R}_i); \)

\( i = i + 1; \)

**end**

Return \( \mathcal{R} = \mathcal{R}_i \)

This algorithm, unlike the previous one, manipulates convex sets with all their computational advantages. At each iteration, the convex hull of the union of the present set and the forward mapping of the same set \( \mathcal{R}_i \) are obtained.

4.3. Complexity and speed of convergence

In this section, we point to the possible extension of Algorithms 1-2 in order to improve the convergence speed. Instead of performing one forward mapping in each iteration before checking \( \mathcal{D} \)-invariance, \( N \) forward mappings are performed in each iteration. This seems to be efficient in the sense that we can reduce the complexity and the number of iterations.

Algorithm 3: Auxiliary set-iterates procedure.

**Data:** A bounded convex set containing the origin \( \mathcal{Q} \in \mathbb{R}^n \); the matrices \( A_0, A_d \in \mathbb{R}^{n \times n} \); \( N \) the number of forward mappings in one iteration

**Result:** \( \mathcal{R} \) Convex \( \mathcal{D} \)-invariant set

\( \mathcal{R}_0 = \mathcal{Q} \);

\( \mathcal{R}_1 = \Omega(\mathcal{Q}) = A_0 \mathcal{Q} \oplus A_d \mathcal{Q} \);

\( Aux_1 = \mathcal{R}_0 \);

\( i = 1; \)

**while** \( \mathcal{R}_i \not\subset \mathcal{R}_{i-1} \) **do**

\( \text{for } m = 1 : N \text{ do} \)

\( Aux_{m+1} = \Phi(Aux_m) \)

**end**

\( Aux = [Aux_1, Aux_2, \ldots, Aux_{N+1}]; \)

\( \mathcal{R}_i = \text{Conv}(Aux); \)

\( \mathcal{R}_{i+1} = \Omega(\mathcal{R}_i); \)

\( i = i + 1; \)

\( Aux_1 = \mathcal{R}_i; \)

**end**

Return \( \mathcal{R} = \mathcal{R}_i \)
\textbf{Example 4.} [39] Let us consider the following dynamical system:

\[ x(k+1) = \begin{bmatrix} 0.2 & 0.1 \\ 0 & 0.6 \end{bmatrix} x(k) + \begin{bmatrix} 0.5 & 0 \\ 0.1 & 0.3 \end{bmatrix} x(k-d). \] (34)

Let

\[ Q = \left\{ x \in \mathbb{R}^2 \mid \begin{bmatrix} \sqrt{2} & -\sqrt{2} \\ -\sqrt{2} & \sqrt{2} \end{bmatrix} x \leq \begin{bmatrix} 0.5 \\ 1 \end{bmatrix} \right\} \]

be the initialization set. By applying Algorithm 3 with \( N = 2 \) and Algorithm 2, two different \( D \)-invariant sets are obtained for the dynamical system (34) in \( 2 \times (N = 2) \) and 18 iterations, respectively. Figure 4 presents these sets. Dashed black lines represent the state trajectories starting from the vertices of these sets with respect to the dynamics (34).

\[ Q = \left\{ \begin{array}{c} x_1 \in \mathbb{R}^2 \mid \\ \begin{bmatrix} \sqrt{2} & -\sqrt{2} \\ -\sqrt{2} & \sqrt{2} \end{bmatrix} x \leq \begin{bmatrix} 0.5 \\ 1 \end{bmatrix} \end{array} \right\} \]

be the initialization set. By applying Algorithm 3 with \( N = 2 \) and Algorithm 2, two different \( D \)-invariant sets are obtained for the dynamical system (34) in \( 2 \times (N = 2) \) and 18 iterations, respectively. Figure 4 presents these sets. Dashed black lines represent the state trajectories starting from the vertices of these sets with respect to the dynamics (34).

It becomes clear that, under the assumption that a \( D \)-invariant set exists, an efficient construction exists. We can also use the algorithmic construction (Algorithm 2) as an induced tool to check if a \( D \)-invariant set can/cannot be obtained, whenever the dDDE satisfies the necessary conditions for the existence of such invariant sets. To illustrate this idea, Example 1, which raises a doubt about the sufficiency of the matrix-pencil based conditions [46], will be discussed in the sequel. By computing the set iterations up to strict inclusion into the initial one, convergence/divergence can be inferred. If the initial set \( Q \) for Algorithm 2 is the \( \infty \)-norm unit ball in \( \mathbb{R}^2 \) and the dDDE is given by the matrices in Example 1, then after 4 iterations one obtains the sequence in Figure 5. The set iteration can be stopped as long as \( Q \) is a strict subset of \( P_4 \). This represents a proof by construction that forward set iterations diverge and the system does not admit a \( D \)-invariant set.

A pictorial overview of the relation between different kinds of stability and existence of \( D \)-invariant sets is given in Figure 6. Solid black lines represent implications that have been proved herein. Solid yellow lines represent previous results and dashed lines with question marks represent open problems. Dashed lines with a cross between two statements show that the first property does not necessarily imply the second.

5. Extensions of \( D \)-invariance

As mentioned in the introduction, two main approaches exist in the literature dealing with positive invariant sets for discrete time-delay difference equations; an invariant set for the dDDE can be computed
either in an extended state space, or in the original state space (in this latter case, it is called $\mathcal{D}$-invariant set). The concept of cyclic invariance [37] can be exploited to compute, instead of a rigid set in $(\mathbb{R}^n)^{d+1}$ or $\mathbb{R}^n$ as in the two aforementioned approaches, a tuple of invariant sets; thus offering a certain degree of flexibility.

**Definition 9.** A $(d+1)$-tuple of sets $\{\Omega_0, \ldots, \Omega_d\}$ is called cyclic $\mathcal{D}$-invariant with respect to (1) if:

\[
\begin{align*}
A_0\Omega_0 \oplus A_1\Omega_1 \oplus \cdots \oplus A_d\Omega_d & \subseteq \Omega_d; \\
A_0\Omega_d \oplus A_1\Omega_0 \oplus \cdots \oplus A_{d-1}\Omega_{d-1} & \subseteq \Omega_{d-1}; \\
\vdots \\
A_0\Omega_1 \oplus A_1\Omega_2 \oplus \cdots \oplus A_d\Omega_0 & \subseteq \Omega_0.
\end{align*}
\]

A generalization of the cyclic invariance notion to invariant family of sets was proposed by [38, 36].

**Definition 10.** A family of $(d+1)$-tuples of sets $\mathcal{F} \subset (\mathbb{R}^n)^{d+1}$ is an invariant family with respect to (1) if for any tuple $\{\Omega_0, \Omega_1, \ldots, \Omega_d\} \in \mathcal{F}$ there exists a set $\Omega_* \subset \mathbb{R}^n$ such that $\{\Omega_0, \ldots, \Omega_d\} \in \mathcal{F}$ and $A_0\tilde{\Omega}_0 \oplus A_1\tilde{\Omega}_1 \oplus \cdots \oplus A_d\tilde{\Omega}_d \subseteq \Omega_*$.

The link between the two main representations for discrete time-delay difference equations and their invariant sets has received recently a unifying characterization via set factorization [33]. The reader is referred to this work for geometrical details on the Cartesian product of sets in relationship with positive invariance for time-delay systems.

6. Conclusion

This paper discusses positive invariance for discrete time-delay systems. Necessary and/or sufficient conditions for the existence of $\mathcal{D}$-invariant sets have been gathered and discussed. The relationship between $\mathcal{D}$-invariance and stability has been studied for discrete delay difference equations (dDDEs). The construction of $\mathcal{D}$-invariant sets via set iterations has been shown to benefit from set convexification, despite the fact that set forward mappings based on the original dDDE lead to a non-convex $\mathcal{D}$-invariant set.

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