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Necessary and sufficient stability condition for
time-delay systems arising from Legendre
approximation

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Abstract—Recently, necessary conditions of stability for
time-delay systems based on the handling of the Lyapunov-Krasovskii
functional have been studied in the literature giving rise to a
new paradigm. Interestingly, the necessary condition for stability
developed by Gomez et al. has been proven to be sufficient. It is
presented as a simple positivity test of a matrix issued from the
Lyapunov matrix. The present paper proposes an extension of
this result, where the uniform discretization of the state has been
replaced by projections on the first Legendre polynomials. Like
in Gomez et al., the stability is guaranteed regarding the sign of
the eigenvalues of a matrix, whose size is given analytically from
convergence arguments. Compared to them, by relying on the
supergeometric convergence rate of the Legendre approximation,
the required order to ensure stability can be remarkably reduced.
Thanks to this significant modification, it is possible to find
an outer estimate of the stability regions, which converges to
the expected stability regions with respect to the number of
projections, as illustrated in the example section.

Index Terms—Time-delay systems, Stability analysis,
Lyapunov-Krasovskii functionals, Approximation theory.

I. INTRODUCTION

Delays appear unavoidably as soon as time processing or
analog-to-digital converters interfere in the communication
between interconnected dynamical systems. Numerous
numerical methods have been deployed to consider this
latency and to analyze the stability of time-delay systems [25].
First of all, the D-partition [26] issued from the modulus-
argument calculation is simple to implement and indicates the
exact stability properties. Furthermore, stability areas can be
inferred using quasi-polynomials approaches and the set-up of
Mikhaylov diagrams [31]. Then, approximated models derived
from pseudo-spectral techniques such as collocation [4],
or tau [22] methods have also been prevalent. Besides
stability sets obtained using bifurcation analysis, the root
locus is outlined. In the Laplace domain, frequency-sweeping
delay-dependent tests have also been developed to avoid case-
by-case studies. The $\mathcal{H}_\infty$ analysis provides accurate stability
results [24] and even results in the design of controllers [13],
[29]. Lastly, in the time domain, it is well-known that the
existence of Lyapunov-Krasovskii functionals leads to a
necessary and sufficient condition of stability [23] even
though the sufficiency is usually not numerically tractable.
The implementation has only been recently made feasible
by discretizing the Lyapunov-Krasovskii functional [10].
Henceforth, tractable necessary and sufficient conditions can
be formulated as a positive definiteness test of a certain
matrix. This method has been applied to various classes of
delay systems with single [11], integral [5], neutral [15] or
multiple [1], [18] delay types. This paper focuses on this last
feature.

For linear finite-dimensional systems with state matrix $A$,
stability is equivalent to the positive definiteness of a
symmetric matrix $P$ solution of the so-called Lyapunov
equation $PA + A^TP = -I$. Concurrently, for linear infinite-
dimensional systems with operator $A$, stability is equivalent
to find a positive hermitian operator $P$ solution of the Lyapunov
equation $PA + A^T P = -I$ (see [8]). Certified implementation
techniques need to be developed [11], [27] to use such a
theoretical necessary and sufficient condition. From one side,
the necessity is directly obtained by the construction of an
approximated Lyapunov-Krasovskii functional [12]. On the
other side, the sufficiency is obtained asymptotically for
sufficiently large approximated orders [9]. In practice, the
approximation is realized by discretizing the Lyapunov matrix
appearing in the operator $P$. The interpolated functions are
selected on each evenly-spaced subinterval as polynomials
(see piece-wise linear or splines schema [20], [27], [28]) or
exponential kernels (see [9], [11], [12]). The latter technique
makes it possible to elegantly end up with point-wise
evaluations of the Lyapunov matrix $U$. Then, a necessary
and sufficient condition of stability is expressed as the
positive definiteness of a matrix, approximating $P$, of size
$n^*$ [17]. The estimation of the order $n^*$ to assess stability
has also been given in [19]. Nevertheless, this estimated
order seems extremely large, pessimistic, and limited by the
discretization schema, which leads us to the following
questions. Is it possible to extend the methodology to other
approximation techniques and to other support basis? Can the
numerical complexity of the numerical test be reduced? In
that direction, we propose here another way to approximate
the Lyapunov-Krasovskii functional following the idea of
projection on a Legendre polynomial basis [32]. The selection
of Legendre polynomials is already meaningful insofar
$\tau$-Legendre models are very efficient to perform convergent
simulations [30] or convergent stability estimates in the
linear matrix inequality framework [2]. By taking the benefits
of Legendre approximation, especially its supergeometric
convergence rate, new necessary and sufficient criterion of
stability is derived and the estimated order $n^*$ is notably
reduced compared to [19].
The article is organized as follows. Section II presents the complete Lyapunov-Krasovskii functional and recalls the necessity and sufficiency of the converse Lyapunov theorem. Section III is dedicated to the supergeometric convergence occurring when performing Legendre approximations. Then, our novel necessary and sufficient numerical condition of stability is exposed in Section IV. The last section deals with computational issues and performances evaluations of our stability test.

Notations: Throughout the paper, \( N \) and \( \mathbb{R}^{m \times p} \) and \( \mathbb{S}^m \) denote the set of natural numbers, real matrices of size \( m \times p \) and symmetric matrices of size \( m \), respectively. For any square matrix \( M \in \mathbb{R}^{m \times m} \), \( M^T \) denotes the transpose of \( M \) and \( \mathcal{H}(M) \) stands for \( M + M^T \). For any matrix \( M \in \mathbb{S}^m \), \( M \succeq 0 \) means that \( M \) is positive definite (the eigenvalues of \( M \) are strictly positive). Furthermore, for any matrix \( M \) in \( \mathbb{R}^{m \times p} \), the 2-norm of \( M \) is \( |M| = \sqrt{\text{tr}(M^T M)} \), where \( \text{tr} \) defines the maximal eigenvalue. The vector \( u = \text{vec}(M) \) if \( \mathbb{R}^{m \times p} \) collocates the columns of \( M \) and the inverse operation is denoted \( \text{vec}^{-1} \) and verifies \( \text{vec}^{-1}(\text{vec}(M)) = M \). Moreover, \( I_m \) is the identity matrix of size \( m \), \( \delta_{jk} \) denotes the Kronecker delta, symbol \( \otimes \) represents the Kronecker product, \( \left[ \begin{array}{c} M_1 \\ M_2 \end{array} \right] \) for \( \left[ \begin{array}{c} M_1 \\ M_2 \end{array} \right] \) and \( \text{diag}(d_1, \ldots , d_m) \) is the diagonal matrix with diagonal coefficients \( d_1, \ldots , d_m \). We also declare functions \( e^r \) and \( \lceil r \rceil \) as the exponential and ceiling part of the real number \( r \), respectively. For functions \( f_1, f_2 \) from \( N \) to \( \mathbb{R} \), equivalence \( f_1(n) \sim f_2(n) \) means that \( \frac{f_1(n)}{f_2(n)} \) is finite as \( n \) tends to infinity. The set of piecewise continuous functions from \( [-h,0] \) to \( \mathbb{R}^m \) is denoted \( \mathcal{C}_{pw}(-h,0;\mathbb{R}^m) \). For any function \( \varphi \) in this set, the induced norm is \( \|\varphi\| = \sup_{[\tau-h,0]}|\varphi(\tau)|. \) Denote also \( \mathcal{C}_\infty(-h,0;\mathbb{R}^m) \), the set of smooth functions from \( [-h,0] \) to \( \mathbb{R}^m \). Finally, the Shimanov notation \( x_t: \{[-h,0) \rightarrow \mathbb{R}^m \} \tau \mapsto x_t(\tau) = x(t + \tau) \) will be used all along the paper.

II. LYAPUNOV NECESSARY AND SUFFICIENT STABILITY CONDITION FOR TIME-DELAY SYSTEMS

A. Time-delay system and Lyapunov-Krasovskii functional

Consider a linear time invariant time-delay system given by
\[
\dot{x}(t) = Ax(t) + A_dx(t-h), \quad \forall t \geq 0, \tag{1}
\]
where \( h > 0 \) is the delay and matrices \( A, A_d \) in \( \mathbb{R}^{m \times m} \) are constant and known. Such a system is initialized by \( x_0 = \varphi \in \mathcal{C}_{pw}(-h,0;\mathbb{R}^m) \) and, for any \( t \geq 0 \), \( x_t \in \mathcal{C}_{pw}(-h,0;\mathbb{R}^m) \) denotes the state of (1).

Definition 1: The trivial solution of system (1) is said to be exponentially stable if there exist \( \kappa \geq 1 \) and \( \alpha > 0 \) such that, for all \( t \geq 0 \) and \( x_0 \in \mathcal{C}_{pw}(-h,0;\mathbb{R}^m) \), \( \|x_t\| \leq \kappa e^{-\alpha t}\|x_0\| \) holds.

In order to study the stability of system (1), recall the Lyapunov-Krasovskii functional introduced in [23]:
\[
V(\varphi) = \int_{-h}^{0} \int_{-h}^{0} \left[ \frac{\varphi(\tau)}{\varphi(\tau)} \right]^T \Pi(\tau_1,\tau_2) \left[ \frac{\varphi(\tau)}{\varphi(\tau)} \right] d\tau_1 d\tau_2, \tag{2}
\]
for any \( \varphi \in \mathcal{C}_{pw}(-h,0;\mathbb{R}^m) \), where matrix \( \Pi \) is given by
\[
\Pi(\tau_1,\tau_2) = \begin{bmatrix}
\frac{U(0)}{\alpha^2} & \frac{1}{2\alpha} U^T(h + \tau_1) A_d \\
\frac{1}{2\alpha} U(h + \tau_1) A_d & \frac{1}{\alpha^2} U(h + \tau_1) A_d
\end{bmatrix} \tag{3}
\]
and where the Lyapunov matrix \( U \) in \( \mathbb{R}^{m \times m} \) is given by \( U = \text{vec}^{-1}(U_0) \) where \( U = \text{vec}(U) \) is given analytically by
\[
U(\tau) = \begin{cases}
\left[ \begin{array}{c}
I_m \\
0 \end{array} \right] e^{M \tau} N^{-1} \left[ \begin{array}{c}
0 \\
I_m \end{array} \right] & \text{if } \tau \geq 0, \\
\left[ \begin{array}{c}
I_m \\
0 \end{array} \right] e^{h(\tau + \tau)} N^{-1} \left[ \begin{array}{c}
0 \\
I_m \end{array} \right] & \text{if } \tau < 0,
\end{cases}
\tag{4}
\]
with
\[
M = \begin{bmatrix}
A^T \otimes I_m & A_d^T \otimes I_m \\
I_m \otimes A_d & -I_m \otimes A_d
\end{bmatrix},
\tag{5}
\]
\[
N = \begin{bmatrix}
I_m + I_m \otimes A_T A_d \otimes I_m & I_m \otimes A_d \\
I_m \otimes A_d & 0
\end{bmatrix} + \begin{bmatrix}
I_m \otimes A_d & 0 \\
0 & -I_m
\end{bmatrix} e^{hM}.
\]

The authors of [9] showed that it is the unique functional that satisfies
\[
V(x_t) = -\|x(t-h)\|^2, \tag{6}
\]
along the trajectories \( x_t \) of system (1).

Remark 1: The Lyapunov matrix \( U \) ensuring (6) is unique if and only if matrix \( N \) is non singular. As explained in [23], such a limitation is the Lyapunov condition and excludes all systems with eigenvalues \( s_1, s_2 \) satisfying \( |s_1 + s_2| = 0 \).

B. Necessary and sufficient stability condition

Under the Lyapunov condition, the authors of [9], [19], [23], [28] provide sufficient and necessary conditions for the exponential stability of system (1), which are recalled below.

Lemma 1: If system (1) is exponentially stable, then there exists \( \eta > 0 \) such that,
\[
V(\varphi) \geq \eta \left( \|\varphi(0)\|^2 + \frac{1}{h} \int_{-h}^{0} \|\varphi(\tau)\|^2 d\tau \right), \forall \varphi \in \mathcal{C}_{pw}(-h,0;\mathbb{R}^m), \tag{7}
\]
where \( V \) is the functional defined by (2) satisfying (6).

Lemma 2: Assume that system (1) has an eigenvalue with a strictly positive real part. Then,
\[
\exists \varphi \in \mathcal{S}, \quad V(\varphi) \leq -\eta_0 = \frac{e^{-2rh}}{4r} - \cos^2(b_0) < 0, \tag{8}
\]
where the functional \( V \) is defined by (2) and (6) and where \( \mathcal{S} \) stands for the compact set given by
\[
\mathcal{S} = \left\{ \varphi \in \mathcal{C}_\infty(-h,0;\mathbb{R}^m); \|\varphi(k)\| = 1, \forall k \in \mathbb{N} \right\}, \tag{9}
\]
with a system dependent parameter \( r \) given by
\[
r = |A| + |A_d|, \tag{10}
\]
and with scalar \( b_0 \) the unique root on \([0, \frac{\pi}{2}]\) of the function \( g(b) := \sin^4(b)((hr)^2 + b^2) - (hr)^2 \). Proof: The proofs of both lemmas are postponed to Appendices A and B.
III. PRELIMINARIES ON LEGENDRE POLYNOMIALS

A. Legendre approximation

Legendre polynomials considered on $[-h, 0)$ are defined by

$$\forall k \in \mathbb{N}, \ell_k(\tau) = (-1)^k \sum_{j=0}^{k} \binom{k}{j} \left(\frac{\tau + h}{2}\right)^j \left(\frac{\tau - h}{2}\right)^j,$$  \hspace{1cm} (11)

where $\binom{k}{j}$ stands for the binomial coefficient [14]. These polynomials $\{\ell_k\}_{k \in \mathbb{N}}$ form an orthogonal sequence of functions, which spans the space of square-integrable functions [14].

For the sake of simplicity, introduce matrix $\ell_n$ in $\mathbb{R}^{nm \times m}$ given by

$$\ell_n(\theta) = \left[\ell_0(\theta)I_m \ell_1(\theta)I_m \ldots \ell_{n-1}(\theta)I_m\right]^\top, \forall \theta \in [-h, 0].$$  \hspace{1cm} (12)

For any function $\varphi$ in $C_{pu}(-h, 0; \mathbb{R}^m)$ and any approximation order $n \in \mathbb{N}^*$, let us decompose

$$\forall \tau \in [-h, 0], \varphi(\tau) = \ell_n(\tau)\Phi_n + \tilde{\varphi}_n(\tau),$$  \hspace{1cm} (13)

where $\varphi_n(\tau) = \ell_n(\tau)\Phi_n$ is the polynomial approximation and $\tilde{\varphi}_n(\tau) = \varphi(\tau) - \varphi_n(\tau)$ is the residual error. The vector $\Phi_n$ represents the normalized $n$ first polynomial coefficients of the function $\varphi$ and is defined by

$$\Phi_n = \text{diag}\left(\frac{1}{h}, \ldots, \frac{2n-1}{h}\right) \otimes I_m \int_{-h}^{0} \ell_n(\tau)\varphi(\tau)d\tau \in \mathbb{R}^{nm},$$  \hspace{1cm} (14)

where $I_m = \int_{-h}^{0} \ell_n(\tau)\ell_n(\tau)d\tau$ is the diagonal Gram-Schmidt normalization matrix.

In the sequel, the objective is to prove that the Legendre approximation $\tilde{\varphi}_n$ converges uniformly towards $\varphi$ with respect to $\tau$ and also to quantify its convergence rate on $S$.

B. Convergence of the Legendre remainder

In light of the polynomial approximation theory [6], it results in an important convergence lemma.

**Lemma 3:** For any function $\varphi$ in $S$, the approximation error $\tilde{\varphi}_n$ in (13) verifies, for any $\varepsilon > 0$,

$$\|\tilde{\varphi}_n\| \leq \varepsilon, \forall n \geq N(\varepsilon),$$  \hspace{1cm} (15)

where $N(\varepsilon)$ is given by

$$N(\varepsilon) = \max\left\{4, \left[\frac{3}{2} + \mu e^{1+\mathcal{W}\left(-\frac{\log(\varepsilon)}{\mu e}\right)}\right]\right\},$$  \hspace{1cm} (16)

where

$$\mu = \frac{hr}{2}, \quad \rho = \sqrt{\frac{2[\mu]}{\pi^3}} \frac{1}{\mu^2} \left(\frac{\mu e}{\left[\mu + \frac{1}{2}\right]}\right)^{\left[\mu + \frac{1}{2}\right]},$$  \hspace{1cm} (17)

and where the Lambert function [7]

$$\mathcal{W} : \mathbb{R}_+ \rightarrow \mathbb{R}_+, \quad z \mapsto \mathcal{W}(z) = y,$$

where $y$ is uniquely defined by the relation $ye^y = z$.

**Proof:** According to [33, Theorem 2.5], an upper bound of the Legendre approximation error $\|\tilde{\varphi}_n\| = \sup_{[-h, 0]} |\tilde{\varphi}_n(\tau)|$ is given by

$$\|\tilde{\varphi}_n\| \leq \sqrt{\frac{\pi^3}{2[\mu]} \frac{\mu n - [\mu]}{(n - \frac{3}{2}) \ldots (1 + \frac{1}{2})}}, \forall n \geq 4,$$  \hspace{1cm} (18)

using Legendre polynomials properties and $n - 1 - [\mu]$ successive integrations by parts. Applying the logarithm to (18) leads to

$$\log \left(\sqrt{\frac{2[\mu]}{\pi^3}} \|\tilde{\varphi}_n\|\right) \leq (n - [\mu]) \log (\mu) - \sum_{k=1+\lfloor \mu \rfloor}^{n-2} \log \left(k + \frac{1}{2}\right).$$  \hspace{1cm} (19)

Since the log function being monotonically increasing, we obtain

$$\sum_{k=1+\lfloor \mu \rfloor}^{n-2} \log \left(k + \frac{1}{2}\right) \geq \int_{x=\mu}^{n-2} \log (x + \frac{1}{2}) dx,$$$$

= \left[\left(x + \frac{1}{2}\right) \log (x + \frac{1}{2})\right]_{\mu}^{n-2},$$

where $e$ denotes the exponential of 1. Reordering the terms and introducing $\rho$ in (17), inequality (19) becomes

$$\log (\rho \|\tilde{\varphi}_n\|) \leq - \left(n - \frac{3}{2}\right) \log \left(\frac{\mu e}{\mu e}\right).$$  \hspace{1cm} (20)

Denoting $y_n := \log \left(\frac{n - \frac{3}{2}}{\mu e}\right)$, we look for the orders $n$ such that the upper bound given by (20) is bounded by $\log (\lambda e)$. Then, the following inequality need to be satisfied

$$-y_n e^{y_n} \leq \frac{\log (\rho e)}{\mu e}.\hspace{1cm} (21)$$

From Lambert function definition [7], it boils down to

$$y_n := \log \left(\frac{n - \frac{3}{2}}{\mu e}\right) \geq W\left(-\frac{\log (\rho e)}{\mu e}\right).\hspace{1cm} (22)$$

Therefore, the orders for which the previous inequality holds satisfy

$$n \geq \frac{3}{2} + \mu e^{1+W\left(-\frac{\log (\rho e)}{\mu e}\right)}.$$  \hspace{1cm} (23)

Together with the initial constraint $n \geq 4$ to employ (18), the expression of $N(\varepsilon)$ is retrieved, which concludes the proof. \hfill \Box

This result allows us to estimate an order that ensures that $\|\tilde{\varphi}_n\|$ is upper bounded by $\varepsilon > 0$, for any $\varphi$ in $S$. The relation between $\varepsilon$ and such a minimal order $N(\varepsilon)$ is depicted in Fig. 1. As expected for smooth functions [3], [33], the uniform convergence of Legendre approximation is supergeometric which means that $\varepsilon = N^{-1}(n) \sim e^{-n \log(n)}$ as emphasized in formula (18).
A. Approximated Lyapunov-Krasovskii functional

In this section, in order to construct an approximated Lyapunov-Krasovskii functional, the complete Lyapunov-Krasovskii functional given by (2) is regarded for particular functions \( \phi \), taken from subsets of \( C_{pw}(-h, 0; \mathbb{R}^{m}) \). For instance, we consider here the space spanned by the \( n \) first Legendre polynomial, and we take support on the \( n \) first Legendre coefficients of \( \phi \) denoted \( \Phi_n \) and expressed in (14).

Let the approximated Lyapunov-Krasovskii functional at order \( n \)

\[
V_n(\varphi) = \begin{bmatrix} \varphi(0) \\ \Phi_n \end{bmatrix}^T \begin{bmatrix} \Phi_n & \varphi(0) \end{bmatrix},
\]

for any \( \varphi \in C_{pw}(-h, 0; \mathbb{R}^{m}) \), with matrix

\[
P_n = \begin{bmatrix} U(0) & Q_n \\ T_n & I - T_n^{-1} \end{bmatrix}.
\]

In the previous expression, we have

\[
Q_n = \int_{-h}^{0} U^T(h + \tau) A_d e_n^T(\tau) d\tau,
\]

\[
T_n = \int_{-h}^{0} \int_{-h}^{0} e_n(\tau_1) A_d^T U(\tau_1 - \tau_2) A_d e_n^T(\tau_2) d\tau_1 d\tau_2.
\]

Remark 2: Note that \( V_n \) does not involve the Legendre remainder \( \tilde{\varphi}_n \). Functional \( V_n \) is an approximation of the Lyapunov-Krasovskii functional \( V \) defined by (2).

Based on the previous section on polynomial approximation, the convergence of this approximated functional towards the complete Lyapunov-Krasovskii functional given by (2) will be established in the next section.

B. Convergence of the approximated Lyapunov-Krasovskii functional

Define the Lyapunov-Krasovskii functional remainder as

\[
\tilde{V}_n(\varphi) = V(\varphi) - V_n(\varphi), \quad \forall \varphi \in C_{pw}(-h, 0; \mathbb{R}^{m}).
\]

Applying expansion (13), this remainder is rewritten as

\[
\tilde{V}_n(\varphi) = \int_{-h}^{0} \int_{-h}^{0} \begin{bmatrix} \varphi(0) \\ \varphi(\tau_1) \\ \varphi_n(\tau_2) \end{bmatrix}^T \Delta_n(\tau_1, \tau_2) \tilde{\varphi}_n(\tau_2) d\tau_1 d\tau_2,
\]

where

\[
\Delta_n(\tau_1, \tau_2) = \begin{bmatrix} \frac{2}{h} U^T(h + \tau_2) A_d \\ 2A_d^T U(\tau_1 - \tau_2) A_d \\ -A_d^T U(\tau_1 - \tau_2) A_d \end{bmatrix}.
\]

The main idea is now to prove, at least in the compact subset \( S \) of \( C_{pw}(-h, 0; \mathbb{R}^{m}) \) given by (9), that the approximated Lyapunov-Krasovskii functional \( V_n \) given by (24) converges towards the complete Lyapunov-Krasovskii functional \( V \) given by (2) with a guaranteed and quantified convergence rate.

Lemma 4: For any \( \varphi \) in \( S \) and \( \eta > 0 \), we have

\[
|\tilde{V}_n(\varphi)| \leq \eta, \quad \forall n \geq N(\varepsilon(\eta)),
\]

where the Lyapunov-Krasovskii remainder \( \tilde{V}_n \) is given by (28), where the order \( N(\varepsilon) \) is described in (16) and where

\[
\varepsilon(\eta) = \frac{\kappa_1 + \kappa_2}{\kappa_2 + 1} + \sqrt{\left( \frac{\kappa_1 + \kappa_2}{\kappa_2 + 1} \right)^2 + \frac{\eta}{h(\kappa_2 + 1)}},
\]

with scalars \( \kappa_1 \) and \( \kappa_2 \) given by

\[
\kappa_1 = \max_{[0,h]} |U(\tau) A_d|, \quad \kappa_2 = h \max_{[-h,h]} |A_d^T U(\tau) A_d|.
\]

Proof: An upper bound of \( |\tilde{V}_n| \) is obtained by

\[
|\tilde{V}_n(\varphi)| \leq 2(\kappa_1 |\varphi(0)| + \kappa_2 ||\varphi||) \int_{-h}^{0} |\tilde{\varphi}_n(\tau)| d\tau \frac{2^h}{h} \left( \int_{-h}^{0} |\tilde{\varphi}_n(\tau)| d\tau \right)^{\frac{2}{h}} + \int_{-h}^{0} |\tilde{\varphi}_n(\tau)|^2 d\tau.
\]

Hence, having \( \varphi \) in \( S \), \( ||\varphi|| = \sup_{[-h,0]} |\varphi(\tau)| \leq 1 \) so that

\[
\frac{1}{h} |\tilde{V}_n(\varphi)| \leq 2(\kappa_1 + \kappa_2) ||\tilde{\varphi}_n|| + (\kappa_2 + 1) ||\tilde{\varphi}_n||^2.
\]

We obtain

\[
|\tilde{V}_n(\varphi)| \leq \eta\text{ under the following quadratic constraint}
\]

\[
-\frac{\eta}{h(\kappa_2 + 1)} + 2 \frac{\kappa_1 + \kappa_2}{\kappa_2 + 1} ||\tilde{\varphi}_n|| + ||\tilde{\varphi}_n||^2 \leq 0,
\]

which is satisfied for \( ||\tilde{\varphi}_n|| \leq \varepsilon(\eta) \). The conclusion is finally drawn thanks to Lemma 3, which states that \( ||\tilde{\varphi}_n|| \leq \varepsilon(\eta) \) holds for any \( n \) greater than \( N(\varepsilon(\eta)) \).

Remark 3: Notice that the maximal values \( \kappa_1 \) and \( \kappa_2 \) can easily be computed by grid search with an equally-spaced grid.

Contrary to previous works based on discretization procedures with exponential kernels [9], [19] or splines [20], [28] which were limited to algebraic convergence rates, we take the benefits of Legendre polynomial approximation to obtain a supergeometric convergence rate on the remainder \( |\tilde{V}_n(\varphi)| \). Therefore, the proposed convergence property of the remainder will be the key to design a stability test for time-delay systems that extends and enhances existing results.
C. Necessary and sufficient stability test

Lemmas 1 and 2 applied to the approximated Lyapunov-Krasovskii functional \( V_n \) defined by (24) provide a new necessary and sufficient condition of stability for system (1).

**Theorem 1:** System (1) is exponentially stable if and only if matrix \( P_{N(E(\eta_0))} \) in (25) is positive definite where \( N, E \) are defined in (16), (31), respectively.

**Proof:** Assume that system (1) is exponentially stable. For any vector \([\bar{\phi}_n]\in \mathbb{R}^{(n+1)m}\) and \( n \in \mathbb{N} \), define function \( \phi \) as
\[
\phi(\tau) = \begin{cases} 
\ell_n^+ (\tau) \Phi_n, & \forall \tau \in [-h, 0), \\
0, & \text{if } \tau = 0.
\end{cases}
\]
Applying Lemma 1 with \( \phi \) given above, there exists \( \eta > 0 \) such that
\[
V(\phi) = V_n(\phi) = [\bar{\phi}_n]^\top P_n [\bar{\phi}_n] \geq \eta [\bar{\phi}_n]^2,
\]
which yields \( P_n > 0 \), for all \( n \in \mathbb{N} \), since \( x \) and \( \Phi_n \) are any independent vectors.

Concerning the sufficiency, assume by contradiction that system (1) is not exponentially stable, and that \( P_{N(E(\eta_0))} \) is not positive definite. This means that there exists a characteristic root of (1) with a given order \( n \) such that
\[
V(\phi) = V_n(\phi) = [\bar{\phi}_n]^\top P_n [\bar{\phi}_n] \leq -\eta_0 + V_n(\phi),
\]
with \( \eta_0 \) given by (8). Finally, the convergence presented in Lemma 4 with \( \eta = \eta_0 \) leads to
\[
V_n(\phi) = [\bar{\phi}_n]^\top P_n [\bar{\phi}_n] \leq 0, \quad \forall n \geq N(E(\eta_0)),
\]
which contradicts \( P_{N(E(\eta_0))} > 0 \).

The proposed theorem provides a numerical test to guarantee stability or instability of time-delay systems, which follows the following sequence.

1) Compute \( n^* = N(E(\eta_0)) \) with \( \eta_0 \) given by (8).
2) Evaluate each element of matrix \( P_{n^*} \).
3) Test the positivity of matrix \( P_{n^*} \) to state the stability.

Notice that this necessary and sufficient condition is formulated as in [19] on the positivity of matrix \( P_n \) for a given order \( n \).

As a background result, a hierarchical sufficient condition for instability of system (1) is also formulated below.

**Corollary 1:** If there exists \( n \in \mathbb{N} \) such that matrix \( P_n \) given by (25) is not definite positive then system (1) is not exponentially stable. Moreover, if this statement holds at an order \( n \), then it also holds at the order \( n + 1 \).

**Proof:** Relying on the necessity part of the proof of Theorem 1, the sufficient condition for instability is trivial. The hierarchy can then be proven because matrix \( P_{n+1} \) at order \( n+1 \) can be written as
\[
P_{n+1} = \begin{bmatrix} P_n & \bar{i} \int_{-h}^0 U(\tau + h) A_d \bar{u}(\tau) d\tau \\
\bar{d} \int_{-h}^0 A_d \bar{u}(\tau) U(\tau - h) A_d \bar{u}(\tau) d\tau + \frac{1}{2}\bar{e} \vec{I}_m \end{bmatrix},
\]
If \( P_n \) is not positive definite then \( P_{n+1} \) cannot be positive definite.

Interestingly, Corollary 1 suggests an algorithm to solve the instability test. It consists in testing \( P_n > 0 \) from \( n = 1 \) to \( n = n^* \). If \( P_n \) is not definite positive, then the system is unstable. Once order \( n^* = N(E(\eta_0)) \) is reached, the system is necessarily stable.

It remains to solve the important problem of the numerical computation of matrix \( P_n \), which is necessarily to implement the algorithm. This is detailed in the next section.

V. COMPUTATIONAL ISSUES

A. Numerical issues

To perform the numerical test presented above, each coefficient of matrix \( P_n \) given by (25) needs to be evaluated numerically. It is worth noticing that this problem is not encountered in [19] since the matrix to be evaluated contains point-wise evaluations of the Lyapunov matrix \( U \). Here, the situation is more complicated since the matrix \( P_n \) given by (25) is equal to
\[
P_n = \begin{bmatrix} 
U(0) & Q_{1}^\top A_d & \cdots & Q_{n-1}^\top A_d \\
* & A_d(T_{00} + \tilde{T}_{00}) A_d + h I_m & \cdots & * \\
* & \cdots & \cdots & * \\
* & \cdots & \cdots & * \\
\end{bmatrix},
\]
where \( Q_k, T_{jk}, \tilde{T}_{jk} \) are the Legendre coefficients of the Lyapunov matrix \( U \) given in the vector form as follows
\[
\begin{align*}
Q_k &= \text{vec}(Q_k) = [t_{m+1}^0 \Gamma_k N^{-1}[\text{vec}(I_m)]] \\
T_{jk} &= \text{vec}(T_{jk}) = [t_{m+1}^0 \Gamma_{jk} N^{-1}[\text{vec}(I_m)]] \\
\tilde{T}_{jk} &= \text{vec}(\tilde{T}_{jk}) = [t_{m+1}^0 \Gamma_{jk}^\circ N^{-1}[\text{vec}(I_m)]]
\end{align*}
\]
where \( \Gamma_k, \tilde{\Gamma}_{jk} \) and \( \Gamma_{jk}^\circ \) are defined by
\[
\begin{align*}
\Gamma_k &= \int_{-h}^0 e^{(h+\tau)M} L_k(\tau) d\tau, \quad \forall k \in \{0, \ldots, n-1\}, \\
\tilde{\Gamma}_{jk} &= \int_{-h}^0 \left( \int_{-h}^{\tau_1} e^{(\tau_1-\tau_2)M} L_k(\tau_2) d\tau_2 \right) L_j(\tau_1) d\tau_1, \\
\Gamma_{jk}^\circ &= \int_{-h}^0 \left( \int_{-h}^{\tau_1} e^{(\tau_2-\tau_1)M} L_k(\tau_2) d\tau_2 \right) L_j(\tau_1) d\tau_1, \\
&\forall (j,k) \in \{0, \ldots, n-1\}^2.
\end{align*}
\]

The question of the numerical implementation of these integral terms in a reasonable time is then raised. Such computations can be done analytically by computer algebra systems but may turn out to be a tough task, especially for large \( n \) or \( m \). For instance, for \( m = 4 \) and \( n = 100 \), the exact calculation of \( P_n \) can take days on a basic computer. An alternative computation through inductive relations is proposed to face the problem and make our results tractable.

B. Iterative calculation of Legendre exponential coefficients

The numerical issue resumes to the calculation of Legendre polynomials coefficients of exponential matrices (41)-(43). To perform this computation recursively, the following relations can be used.
Proposition 1: If $M$ is a non singular matrix, then matrices $\Gamma_k$ in (41) can be computed by the recursive relation
\[
\Gamma_k = \Gamma_{k-2} - \frac{2(2k-1)}{h}M^{-1}\Gamma_{k-1}, \quad \forall k \geq 2,
\]
initialized with
\[
\begin{bmatrix}
\Gamma_0 \\
\Gamma_1 \\
\Gamma_2
\end{bmatrix} = \begin{bmatrix}
M^{-1}(e^{hM}-I_{2m^2}) \\
M^{-1}(e^{hM}+I_{2m^2})-\frac{h}{2}M^{-1}I_0
\end{bmatrix}.
\]

Proposition 2: For any matrix $M$ and for matrices $\bar{\Gamma}_{jk}$, $\bar{\Gamma}_{jk}$ expressed in (42),(43), the following equality holds
\[
\bar{\Gamma}_{jk} = (-1)^j+i^k\bar{\Gamma}_{jk}, \quad \forall (j, k) \in \{0, \ldots, n-1\}^2.
\]

Proposition 3: If $M$ is a non singular matrix, then matrices $\bar{\Gamma}_{jk}$ in (42) can be computed by the following relations
\[
\bar{\Gamma}_{jk} = \left\{ \begin{array}{l}
(1)^{j+k}\bar{\Gamma}_{kj}, \\
\bar{\Gamma}_{jk-2} + M^{-1}(I_{2m^2})-\frac{h^2}{2}M^{-1}\delta_{jk-1},
\end{array} \right\} \forall k < j,
\]
initialized with
\[
\begin{bmatrix}
\bar{\Gamma}_{00} \\
\bar{\Gamma}_{01} \\
\bar{\Gamma}_{11}
\end{bmatrix} = \begin{bmatrix}
M^{-1}(\Gamma_0-hI_{2m^2}) \\
-M^{-1}\Gamma_1 \\
M^{-1}(\frac{h}{2}I_{2m^2})-\frac{h}{2}I_{2m^2}
\end{bmatrix}.
\]

Proof: The proofs of these propositions are respectively postponed to Appendices C, D and E. □

Remark 4: Notice that the case where $M$ is singular has not been discussed in this paper. It could for instance be treated using the Jordan canonical form.

These propositions allows to propose a numerical solution to compute the integral terms (41)-(43) by induction. The induction requires a number of operations in the range of $n^2$ and $m^2$. The necessary and sufficient condition of stability for time-delay systems presented in this paper becomes numerically tractable in a reasonable time. In the last section, our numerical test for stability is performed on two examples.

VI. APPLICATION TO NUMERICAL EXAMPLES

A. Presentation of the examples

Example 1: Consider (1) with $A = 1$ and $A_d = -2$.

Example 2: Consider (1) with $A = \begin{bmatrix}
0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 \\
-10-K & 10 & 0 & 0 \\
5 & -15 & 0 & -0.25
\end{bmatrix}$, for any $K > 0$.

B. Numerical results on stability analysis

Theorem 1 is applied for point-wise values of delays to evaluate the stability of these systems with respect to the delay. The results are gathered in Table I and II for Example 1 and 2, respectively. For both systems, one can see that the maximal allowable delay $h$, which guarantees the stability, can be given with a precision $0.001$.

The estimated order $n^* = N(\epsilon(\eta_0))$ for our necessary and sufficient test of stability is reported on Tables I and II. For Example 2, a map of orders in the $(K, h)$ plan is depicted in Fig 2. One can see that order $n^*$ increases as parameters $h$ and $K$ increase. The exact limit between stable (on the right) and unstable (on the left) regions obtained using D-partition is superposed. As emphasized in Remark 1, this line is excluded to our criterion.

The CPU time spent to compute $n^*$, evaluate the components of matrix $P_n$ by induction and test the positivity of $P_n^*$, is also reported on both Tables. The computational load is obviously increasing as the order increases. Actually, the processing time is not only impacted by the positivity test but also by the calculation of the integrals terms in matrix $P_n$. By using the iterative relation given by Propositions 1, 2 and 3, this time grows with the square of the order $n^*$.

Other positivity tests of finite size $n^*$ for stability of time-delay systems have also been developed by Gomez et al. and Zhabko et al. A brief comparison with [19] is presented in Table I. The approximation of the complete Lyapunov-Krasovskii functional is realized by Legendre polynomial approximation here, whereas a discretization with exponential kernels is used in [19]. From the supergeometric convergence rate of Legendre approximation, our estimation of the minimal order $n^* = N(\epsilon)$ satisfies $\epsilon \sim e^{-n \log(n)}$. In [19], this estimation is limited to $\epsilon \sim \frac{1}{n^2}$. On Table I, this order of magnitude significant difference is highlighted. Notice that this advantage has to be balanced with the shape of the matrix $P_n$ and the ease of computing its components.

Finally, recalling the sufficient criterion of instability mentioned in Corollary 1. It permits to detect some unstable systems by testing $P_n > 0$ for a limited number of order. For Example 2, for low orders $n = \{1, \ldots, 5\}$, the corresponding unstable areas with respect to parameters $(K, h)$, denoted $\{U_1, \ldots, U_6\}$, are drawn on Fig. 3. These tests are not time consuming and already get an accurate information on the unstable regions of Example 2. Notice that $U_1$ already spans the main areas of instability. The hierarchical structure $U_1 \subset \cdots \subset U_6$ is also verified. More interestingly, the hard-to-reach areas are located when the eigenvalues crosses the imaginary axis from the left-half plane to the right-half plane (see red lines).

VII. CONCLUSIONS AND PERSPECTIVES

This paper has been devoted to the formulation of a necessary and sufficient stability condition of time-delays systems, extending the result of paper [19]. It derives from the positivity of the complete Lyapunov-Krasovskii functional, where an
approximation of the Lyapunov matrix has been considered. Stability can then be linked with the positive definiteness of a certain matrix of finite size $n^*$, which depends on systems parameters. The originality of our work relies on the approximation techniques which employs polynomial coefficients instead of discretized elements. The supergeometric convergence rate satisfied by Legendre polynomials demonstrates that our approach requires smaller orders $n^*$ than in [19]. Based on recurrence relations satisfied by Legendre polynomials, our stability criterion can be easily implemented by induction.

Our approach still to be generalized to distributed [9] and neutral [15] time-delay systems. New tracks of research would be to cover other classes of delay systems and more complex infinite-dimensional systems by this methodology. The Legendre polynomial approximation technique could also be deployed to problems of controller or observer synthesis for infinite-dimensional systems.

APPENDIX

A. Proof of Lemma 1

Proof: The proof can be found in [9, Th. 3] with $W = I_m$ and follows arguments given in [21, Th. 5.19]. Firstly, we introduce the functional

$$ W(\varphi) = V(\varphi) - \int_{-h}^{0} \begin{bmatrix} \varphi(0) \\ \varphi(\tau) \end{bmatrix}^T \begin{bmatrix} \frac{\partial^2}{\partial \tau^2} I_m & 0 \\ 0 & \frac{1}{2} I_m \end{bmatrix} \begin{bmatrix} \varphi(0) \\ \varphi(\tau) \end{bmatrix} d\tau. $$

(49)

Functional $V$ is built so that the time derivative along the trajectories $x_t$ of system (1) gives (6). Then, the time derivative of the functional $W$ gives

$$ W(x_t) = - \begin{bmatrix} x(t) \\ x(\tau-h) \end{bmatrix}^T \begin{bmatrix} \eta_0 H(A) + \frac{1}{2} I_m & \eta_0 A_d \\ \frac{1}{2} I_m & \frac{1}{2} I_m \end{bmatrix} \begin{bmatrix} x(t) \\ x(\tau-h) \end{bmatrix}, $$

(50)

for which there exists a sufficiently small $\eta_0 > 0$ such that $W(x_t) \leq 0$. Then, integrating from 0 to $\infty$ and assuming the exponential stability of (1) yields $W(x_0) \geq 0$, for any initial conditions $x_0 = \varphi$ in $C_{pu}(-h, 0; \mathbb{R}^m)$. Thus, (7) holds with $\eta = \min(\eta_0, \frac{h}{2})$. □

B. Proof of Lemma 2

Proof: The following proof is close to [19, Appendix] and to [28, Lemma 5]. The only difference relies on the definition of set $S$ that has been extended to $C_\infty$ instead of $C_1$. Let us denote $s_0 = \alpha + i\beta$ be the eigenvalue with positive real part of system (1). According to [16], there exists a vector $C = C_1 + iC_2$ such that $|C_2| \leq |C_1| = 1$, $C_2^* C_1 = 0$ and that

$$(s_0 I_m - A - A_d e^{-h s_0}) C = 0.$$  

Consequently, we have $\alpha \leq |s_0| \leq |A| + |A_d| = r$, and

$$\bar{x}(t) = e^{\alpha t} (\cos(\beta t) C_1 - \sin(\beta t) C_2), \quad \forall t \in \mathbb{R},$$

is a solution to system (1). Then, thanks to (6), the derivatives of $V(\bar{x}_t)$ along the trajectories of (1) yields to

$$\dot{V}(\bar{x}_t) = -|\bar{x}(t-h)|^2. $$

(51)

After some calculations developed in [19], we also obtain that

$$V(x_0) \leq -\frac{e^{-2\alpha h}}{4\alpha} \cos^2(b_0) \leq -\frac{e^{-2r h}}{4r} \cos^2(b_0) = -\eta_0. $$

(52)

Let $\varphi = \bar{x}_0$, which belongs to $C_\infty(-h, 0; \mathbb{R}^m)$ and satisfies $|\varphi(0)| = |\bar{x}(0)| = 1$. We finally prove by induction that $|\varphi(k)(\tau)| \leq r^k e^{\alpha k}$, for any $\tau \in [-h, 0]$. Initially, $|\bar{x}(\tau)| \leq e^{\alpha \tau}$ holds on $[-h, 0]$. Then, assuming $|\bar{x}(k)(\tau)| \leq r^k e^{\alpha k}$, since $\bar{x}$ satisfies (1) and is infinitely differentiable, we obtain

$$|\varphi(k+1)(\tau)| \leq |A| |\varphi(k)(\tau-h)| + |A_d| |\varphi(k)(\tau-h)| \leq r^{k+1} e^{\alpha \tau}. $$

Therefore, $\|\varphi(k)\| = \sup_{[-h, 0]} |\varphi(k)(\tau)| \leq r^k$ for any $k \in \mathbb{N}$. □

C. Proof of Proposition 1

Proof: The proof is based on the relation

$$l_k' - l_{k-2} = \frac{2(2k-1)}{h} l_{k-1}, \quad \forall k \geq 2, $$

(53)

satisfied by Legendre polynomials [14]. To compute $\Gamma_k$, an integration by parts leads to

$$\Gamma_k - \Gamma_{k-2} = \int_{-h}^{0} e^{(h+\tau)M}(l_k - l_{k-2})(\tau) d\tau,$$

$$= -\frac{2(2k-1)}{h} M -1 \int_{-h}^{0} e^{(h+\tau)M}l_{k-1}(\tau) d\tau + M^{-1} \left[ e^{(h+\tau)M}(l_k - l_{k-2})(\tau) \right]_{-h}^{0},$$

and knowing that $l_k(-h) = l_{k-2}(-h) = (-1)^k$ and $l_k(0) = l_{k-2}(0) = 1$, the last term vanishes. Moreover, for $k \in \{0, 1\}$, we directly obtain

$$\Gamma_0 = \int_{-h}^{0} e^{(h+\tau)M} d\tau = M^{-1} (e^{h M} - I_{2m^2}), $$

$$\Gamma_1 = \int_{-h}^{0} e^{(h+\tau)M} \left( \frac{2\tau+1}{h} \right) d\tau = M^{-1} (e^{h M} + I_{2m^2}) - \frac{2}{h} M^{-1} \Gamma_0,$$

which concludes the proof. □
D. Proof of Proposition 2

Proof: The successive changes of variables $\tau'_2 = -(\tau_2 + h)$ and $\tau'_1 = (\tau_1 + h)$ directly lead to

$$
\Gamma_{jk} = \int_{-h}^{0} \left( \int_{0}^{-(\tau_1 + h)} e^{(\tau_1 - \tau_2) M} l_k(\tau_2 - h) d\tau_2 \right) l_j(\tau_1) d\tau_1,
$$

which means that, using the parity properties of Legendre polynomials, i.e.

$$
l_k(-\tau - h) = (-1)^k l_k(\tau) \text{ for all } \tau \in [-h, 0].
$$

E. Proof of Proposition 3

Proof: As in Appendix C, an integration by parts and (53) ensure that $\Gamma_{jk}$ satisfies the recursive relation

$$
\Gamma_{jk} - \bar{\Gamma}_{jk-2} = \int_{-h}^{0} \left( \int_{0}^{-(\tau_1 + \tau_2)} e^{(\tau_1 - \tau_2) M} l_k(\tau_2 - h) d\tau_2 \right) l_j(\tau_1) d\tau_1,
$$

and $\tau'_2 = \tau_1$, the following equations hold

$$
\bar{\Gamma}_{jk} = \int_{-h}^{0} \left( \int_{0}^{-(\tau_1 + \tau_2)} e^{(\tau_1 - \tau_2) M} l_j(\tau_2) l_k(\tau_1) d\tau_2 d\tau_1,
$$

An integration by parts yields the initial values.

References


