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A Core Theory of Delay Systems

Sébastien Boisgérault*

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Contents

Abstract	2
1 Introduction	3
2 The Initial Value Problem	4
2.1 Delay Operators	4
2.2 Well-Posedness	6
2.3 The FSA System	9
3 Graph-Theoretic Analysis	10
3.1 Block Diagrams	10
3.2 Causality Analysis	12
3.3 Block Diagram of the FSA System	15
4 Stability	15
4.1 Test in the Laplace Domain	15
4.2 Stability of the FSA System	18
Appendix	18
References	20

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Abstract

We introduce a framework for the description of a large class of delay-differential algebraic systems, in which we study three core problems: first we characterize abstractly the well-posedness of the initial-value problem, then we design a practical test for well-posedness based on a graph-theoretic representation of the system; finally, we provide a general stability criterion. We apply each of these results to a structure that commonly arises in the control of delay systems.

1 Introduction

Systems of delay-differential algebraic equations (DDAE) combine differential and algebraic equations with delayed variables in the right-hand side. Among them, we represent linear and time-invariant systems with a finite memory r as

$$\begin{aligned}\dot{x}(t) &= Ax_t + By_t \\ y(t) &= Cx_t + Dy_t\end{aligned}\tag{1}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$ and the notation z_t stands for the memory of the variable z at time t : the function defined by $z_t(\theta) = z(t + \theta)$ for any $\theta \in [-r, 0]$. The symbols A , B , C and D denote delay operators: linear and bounded operators from the space of continuous functions $C([-r, 0], \mathbb{C}^j)$ to \mathbb{C}^i for some integers i and j . Specifically, since $(x(t), y(t)) \in \mathbb{R}^{n+m}$, we require that the compound operator

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix}\tag{2}$$

is linear and continuous from $C([-r, 0], \mathbb{C}^{n+m})$ to \mathbb{C}^{n+m} .

This description is general enough to encompass systems of seemingly different nature – such as ordinary differential equations (e.g. $\dot{x}(t) = x(t)$), delay systems of retarded type (e.g. $\dot{x}(t) = x(t - 1)$) or neutral type (e.g. $\dot{y}(t) = \dot{y}(t - 1)$, rewritten as $\dot{x}(t) = 0$ and $y(t) = x(t) + y(t - 1)$), difference equations (e.g. $y(t) = y(t - 1)$) but also different types of delays: systems with single delays (e.g. $\dot{x}(t) = x(t) + x(t - 1)$), multiple delays (e.g. $\dot{x}(t) = x(t - 1) + x(t - 2)$), distributed delays (e.g. $\dot{x}(t) = \int_{t-1}^t x(\theta) d\theta$) or a combination thereof. We refer the reader to (Bensoussan et al. 2006, chap. 4) for a more comprehensive collection of examples.

However, competing theories of delay systems have been developed and used over time (Hale 1977; M. Delfour and Karakchou 1987; Salamon 1984); they may use different definitions of delay operators, of the solutions to the initial-value problem and may be restricted to systems of a certain type. For the newcomer in the field, this is an unintended source of complexity. To make the subject more widely accessible, we lay out in this paper a simple but general framework for the description of delay systems, based on a combination of linear algebra and measure theory. Then we develop on this foundation a core theory: we characterize the well-posedness of the initial-value problem, then we perform a graph-theoretic analysis of this issue and finally, we provide a general stability criterion.

We illustrate the results of each section with the same example: the so-called finite spectrum assignment (FSA) architecture which is used for the control of dead-time systems (Manitius and Olbrot 1979). This example – and the methods

that we use to study it – are actually representative of a large class of controllers for delay systems which ranges from the primeval Smith predictor to the general observer-predictor structure (Mirkin and Raskin 2003). A linear dead-time systems is governed by an ordinary differential equation $\dot{x}(t) = Ex(t) + Fu(t)$ but the only information available at time t is the delayed variable $x(t - T)$. If we apply the control $u(t) = -Gy(t)$ where $y(t)$ is the predicted value of $x(t)$ based on $x(t - T)$ and the history of u defined by

$$y(t) = e^{TE}x(t - T) + \int_{-T}^0 e^{-\theta E}Fu_t(\theta)d\theta$$

then the closed-loop dynamics is

$$\begin{aligned} \dot{x}(t) &= Ex(t) - FGy(t) \\ y(t) &= e^{TE}x(t - T) - \int_{-T}^0 e^{-\theta E}FGy_t(\theta)d\theta \end{aligned} \quad (3)$$

Since this system is not overly simplistic (it combines effectively differential and algebraic equations, its delays are discrete and distributed), it is a good testbed for the theory exposed in this paper.

2 The Initial Value Problem

2.1 Delay Operators

We have described general delays as operators applied to continuous functions: if $C(X, Y)$ denotes the space of continuous functions from X to Y , a delay operator with memory length r is a bounded linear operator from $C([-r, 0], \mathbb{C}^j)$ to \mathbb{C}^i for some integers i and j . But delays also have two alternate (and equivalent) representations – as measures and as convolution kernels – which we will use extensively in the next sections.

We start with a quick example of these representations: consider $y(t) = x(t - 1)$ where $x(t)$ is a scalar variable. We will associate to this delayed expression three distinct representations m_1 , m_2 and m_3 such that by construction

$$y(t) = m_1x_t = \int_{-1}^0 x_t(\theta) dm_2(\theta) = \int_0^1 x(t - \theta)dm_3(\theta). \quad (4)$$

The first one m_1 is an operator applied to continuous scalar functions defined on $[-1, 0]$; here, equation (4) clearly mandates that $m_1\chi = \chi(-1)$. The representations m_2 and m_3 are measures; equation (4) holds if m_2 is δ_{-1} , the Dirac

measure at $t = -1$ and m_3 is δ_1 , the Dirac measure at $t = 1$. Now if instead we consider the distributed delay $y(t) = \int_{t-1}^t x(\theta)d\theta$, equation (4) is satisfied with

$$m_1\chi = \int_{-1}^0 \chi(t) dt, \quad m_2 = dt|_{[-1,0]} \quad \text{and} \quad m_3 = dt|_{[0,1]}.$$

In general, we deal with vector-valued variables; in this broader context, the term “measure” may refer to several things. We use the term *scalar measure* for a complex-valued and countably additive function defined for the bounded Borel subsets of the real line (Fell and Doran 1988). A *matrix-valued measure* (resp. *vector-valued measure*) is a countably additive function defined for the bounded Borel subsets of the real line whose values are complex matrices (resp. vectors) of a fixed size. Equivalently, it is as a matrix (resp. vector) whose elements are scalar measures.

Given a matrix-valued measure L , we define the integral of the vector function $\phi : \mathbb{R} \rightarrow \mathbb{C}^j$ with respect to L as the vector of \mathbb{C}^i whose k -th element is given by

$$\forall k \in \{1, \dots, j\}, \quad \left[\int dL \phi \right]_k = \sum_{l=1}^j \int \phi_l dL_{kl}.$$

Riesz’s representation theorem provides a bijection between delay operators with memory length r and matrix-valued measures supported on $[-r, 0]$. We may therefore use the same symbol L to denote both objects, and this convention yields for any $\phi \in C([-r, 0], \mathbb{C}^j)$

$$L\phi = \int dL \phi.$$

The nature of the argument of L (function or set) determines without ambiguity which representation of L is used in a given context. When the argument is a set, we will also drop the parentheses when it improves readability: typically, instead of $L(\{0\})$, we will use the lighter notation $L\{0\}$.

Let L^* refer to the measure obtained by symmetry of the measure L around $t = 0$, which is defined for any bounded Borel set B by $L^*(B) = L(-B)$. This construct obviously provides a new bijection between delay operators with memory length r and matrix-valued measures supported on $[0, r]$, which is related to the representation of delay operators as convolutions.

We say that a measure μ is *limited on the left* if its support is included in $[-r, +\infty)$ for some $r \in \mathbb{R}$, *causal* if its support is included in $[0, +\infty)$ and *strictly causal* if additionally $\mu\{0\} = 0$. The convolution $\mu * \nu$ of two scalar measures limited on the left μ and ν is the scalar measure defined for any bounded Borel set B by

$$(\mu * \nu)(B) = \int \nu(B - x)d\mu(x).$$

The convolution of two vector or matrix-valued measures limited on the left combines the scalar convolution and the linear algebra product: for example, the convolution of two measures L and M with values in $\mathbb{C}^{i \times j}$ and $\mathbb{C}^{j \times k}$ is the measure $L * M$, with values in $\mathbb{C}^{i \times k}$, given by

$$(L * M)(B)_{\alpha, \beta} = \sum_{\gamma=1}^j (L_{\alpha, \gamma} * M_{\gamma, \beta})(B).$$

We may identify a (locally integrable) function f with the measure given by

$$f(B) = \int_B f(t) dt.$$

This enables us to define the convolution of functions and measures, which may be identified with a function, and of two functions, which may be identified with an absolutely continuous function. Additionally, we implicitly extend by 0 functions that are defined on a proper subset of real line – for example functions defined on $[-r, 0]$ or $[-r, +\infty)$. With these conventions, if L is a delay operator with memory r and $\phi \in C([-r, +\infty), \mathbb{C}^i)$ then for almost every $t > 0$

$$L\phi_t = (L^* * \phi)(t).$$

2.2 Well-Posedness

Several concepts of solution exist for the problem formally described by

$$\forall t > 0 \quad \begin{cases} \dot{x}(t) &= Ax_t + By_t \\ y(t) &= Cx_t + Dy_t \end{cases}, \quad (x(0), x_0, y_0) = (\phi, \chi, \psi). \quad (5)$$

A classic (or continuous) solution is a pair of continuous functions (x, y) from $[-r, +\infty)$ to \mathbb{C}^{n+m} such that \dot{x} exists, is continuous on $(0, +\infty)$ and satisfies the system equations and initial conditions. In this setting, necessarily the functions χ and ψ are continuous, $\phi = \chi(0)$ and the consistency condition

$$\psi(0) = C\chi + D\psi$$

holds.

The continuity assumption ensures that the right-hand side of the system equations is defined for any time t ; if we relax this assumption, we may generalize the concept of solution to locally integrable solutions (Salamon 1984; M. Delfour and Karakchou 1987).

Let L be a delay operator with memory length r and L^* the corresponding convolution kernel. For any continuous function z defined on $[-r, +\infty)$, we have $Lz_t = (L^* * z)(t)$ for any $t > 0$ but the right-hand side of this equation is still properly defined – as a locally integrable function of t – if z is merely

locally integrable, a strong incentive to rewrite the initial value problem as a convolution equation. Using the Heaviside step function e , defined by

$$e(t) = \begin{cases} 1 & \text{if } t \geq 0, \\ 0 & \text{otherwise,} \end{cases}$$

we may also rewrite the (integral form of) the differential equation as a convolution equation. We end up with the following definition:

Definition – Solution of the Initial Value Problem. *A pair of locally integrable functions (x, y) , defined on $[-r, +\infty)$, with values in \mathbb{C}^{n+m} , is a (locally integrable) solution of the DDAE initial value problem (5) if*

$$(x(0^+), x_0, y_0) = (\phi, \chi, \psi)$$

and if there is a $f \in \mathbb{C}^n$ such that

$$\begin{bmatrix} x \\ y \end{bmatrix} (t) = \begin{bmatrix} e * A^* & e * B^* \\ C^* & D^* \end{bmatrix} * \begin{bmatrix} x \\ y \end{bmatrix} (t) + \begin{bmatrix} f \\ 0 \end{bmatrix} \text{ for a.e. } t > 0.$$

This definition makes sense even if the initial function data is not continuous, does not meet the consistency condition, or if $\chi(0) \neq \phi$. It is however consistent with the concept of continuous solution when such solutions exist. The constant vector f is uniquely determined by the initial value: we have $x(0^+) = \phi$ if and only if $f = \phi - (e * A^* * \chi + e * B^* * \psi)(0)$.

Now, because of its algebraic component, DDAE system (5) may have no solutions or multiple solutions; we may actually easily exhibit an algebro-differential system (with no delay) with this property. Consider

$$\forall t > 0, \begin{cases} \dot{x}(t) & = & 0 \\ y(t) & = & x(t) + y(t) \end{cases} \quad (6)$$

It is defined formally by $n = 1$, $m = 1$, for example $r = 1$ (any nonnegative value is admissible) and for any $\chi \in C^0([-r, 0], \mathbb{R}^n)$ and $\psi \in C^0([-r, 0], \mathbb{R}^m)$, $A\chi = 0$, $B\psi = 0$, $C\chi = \chi(0)$, and $D\psi = \psi(0)$. If $x(0^+) = 0$, then $x(t) = 0$ for $t > 0$ is the unique solution of the first system equation; the second equation becomes $y(t) = y(t)$, hence arbitrary values of $y(t)$ for $t > 0$ satisfy it: multiple solutions exist. On the contrary, if $x(0^+) = 1$, then necessarily $x(t) = 1$ for any $t > 0$; the second equation would become $y(t) = 1 + y(t)$, which no function $y(t)$ may satisfy: there are no solutions to this initial value problem.

A simple assumption that yields uniqueness of the solution is *explicitness*: a DDAE system is explicit if the right-hand side of its algebraic equation only depends strictly causally on the variable y , that is, if $D\{0\} = 0$. For example, system (6) is not explicit since $y(t) = x(t) + y(t)$ and thus $D\{0\} = 1$; replace this equation with $y(t) = x(t) + y(t - \epsilon)$ for any $\epsilon > 0$ and it becomes explicit.

However this assumption is too conservative for some practical use cases, including the kind of sound composition of input-output systems that is exposed in the next section. Therefore we state in this section a well-posedness theorem applicable under a more general assumption that ensures that the system is merely equivalent to an explicit system.

Let I_p be the $p \times p$ identity matrix and X be the product space

$$X = \mathbb{C}^n \times L^2([-r, 0], \mathbb{C}^{n+m}). \quad (7)$$

endowed with the norm

$$\|(\phi, \chi, \psi)\|_X^2 = |\phi|^2 + \int_{-r}^0 |\chi(t)|^2 + |\psi(t)|^2 dt. \quad (8)$$

Theorem – Well-posedness. *DDAE systems such that $I_m - D\{0\}$ is invertible are well-posed in the product space X : when $(\phi, \chi, \psi) \in X$, there is a unique solution (x, y) to the initial value problem (5), it belongs to $L^2_{\text{loc}}([-r, +\infty), \mathbb{C}^{n+m})$, x is continuous and for any $t > 0$ there is some $\alpha > 0$ such that*

$$\|(x(t), x_t, y_t)\|_X \leq \alpha \|(\phi, \chi, \psi)\|_X. \quad (9)$$

It is classic that this result holds when the system is explicit (Salamon 1984; M. Delfour and Karakchou 1987; Boisgérault 2013). Therefore, to prove the general case, we only need to demonstrate the following lemma:

Lemma – Equivalent Systems. *Assume that the matrix $J := I_m - D\{0\}$ associated to system (5) is invertible and denote E and F the delay operators $E := J^{-1}C$ and $F := J^{-1}(D - D|_{\{0\}})$. The original initial value problem and the one defined by*

$$\forall t > 0 \quad \begin{cases} \dot{x}(t) &= Ax_t + By_t \\ y(t) &= Ex_t + Fy_t \end{cases}, \quad (x(0), x_0, y_0) = (\phi, \chi, \psi). \quad (10)$$

have the same solutions but the latter system is always explicit:

$$F\{0\} = 0.$$

Proof. We may decompose the convolution kernel of D into two components:

$$D^* = (D|_{[-r, 0]})^* = (D|_{\{0\}})^* + (D|_{[-r, 0]})^*.$$

The first component is instantaneous: for any locally integrable function y

$$((D|_{\{0\}})^* * y)(t) = D\{0\}y(t);$$

the second component is strictly causal:

$$(D|_{[-r,0]})^*\{0\} = D|_{[-r,0]}\{0\} = 0.$$

Consequently, a pair of locally integrable functions $(x, y) : [-r, +\infty) \rightarrow \mathbb{R}^{n+m}$ is a solution of the DDAE system equations (5) if and only if they satisfy, almost everywhere for $t > 0$, the equations

$$\begin{aligned} x(t) &= (A^* * x)(t) + (B^* * y)(t) + f \\ y(t) - D\{0\}y(t) &= (C^* * x)(t) + ((D|_{[-r,0]})^* * y)(t) \end{aligned}$$

By assumption the matrix $J = I_m - D\{0\}$ is invertible, thus the second equation is equivalent to

$$y(t) = (E^* * x)(t) + (F^* * y)(t)$$

with $E = J^{-1}C$ and $F = J^{-1}D|_{[-r,0]}$. The functions x and y are solutions of the original DDAE system if and only if they are solutions – with the same initial values – of the system whose delay operators are A , B , E and F . This new DDAE system is explicit by construction: $F\{0\} = J^{-1}D|_{[-r,0]}\{0\} = 0$. ■

2.3 The FSA System

The delay operators A , B and C of system (3) are defined by $Ax_t = Ex(t)$, $By_t = -FGy(t)$ and $Cx_t = e^{TE}x(t - T)$. Since the Dirac measure δ_τ at $\tau \in \mathbb{R}$ satisfies $\delta_\tau\varphi = \varphi(\tau)$, we have $A = E\delta_0$, $B = -FG\delta_0$, $C = e^{TE}\delta_{-T}$ or equivalently

$$A^* = E\delta_0, \quad B^* = -FG\delta_0, \quad C^* = e^{TE}\delta_T. \quad (11)$$

The fourth delay operator D is defined by

$$Dy_t = - \int_0^T e^{\theta E} FGy(t - \theta) d\theta = -(e^{\theta E} FGd\theta|_{[0,T]} * y)(t),$$

hence

$$D^* = -e^{\theta E} FGd\theta|_{[0,T]}. \quad (12)$$

In particular

$$D^*\{0\} = - \int_{\{0\}} e^{\theta E} FGd\theta = 0,$$

thus this DDAE system is explicit and therefore it is well-posed.

3 Graph-Theoretic Analysis

Block diagrams used in control theory provide an alternative to the system-of-equations approach for the modelling of dynamical systems: in this context, dynamical systems are specified as diagrams made of blocks and wires as in figure 1. Blocks are independent input-output systems with their own behaviors: each block specifies how to compute the values of its output variables given the values of its input variables (and optionally of some local initial state). In figure 1, each rectangle is a block and the symbol it holds defines its behavior. Wires connect these blocks to build larger systems: they specify how the input of each block is computed as a combination of the outputs of all blocks.

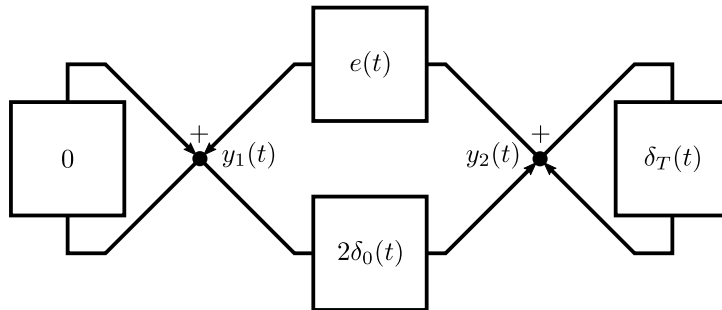


Figure 1: Block diagram of a simple delay system.

This classic description – actually a graph representation in disguise – offers new insights into the structural properties of dynamical systems. In section 3.1, we complete for delay systems the informal definition of block diagrams that we have given so far and expose the connexion with graph theory. Then, in section 3.2, we focus on causality to design sound composition rules of subsystems, that ensure the well-posedness of the initial-value problem for the global system.

3.1 Block Diagrams

For linear time-invariant systems, block diagrams can always be decomposed into elementary input-output systems with a scalar input $u(t)$ and a scalar output $y(t)$, defined for $t > 0$, and characterized by a scalar convolution kernel k^* . Additionally, for delay systems, only two types of elementary subsystems are required: the integrator, whose kernel is the Heaviside step function e , and (scalar) finite-memory delay operators. By definition, the behavior of an elementary system is governed by

$$y(t) = (k^* * u)(t) + (k^* * \omega)(t), \quad t > 0;$$

where ω , a scalar function defined on $[-r, 0]$ that represents a finite memory of past input values, determines the system initial state.

A block diagram embeds a collection of such systems into a (directed, labeled) graph structure: each vertice i refers to an output variable y_i and each edge $j \rightarrow i$ is labeled with a scalar kernel $k^* = K_{ij}^*$; the global initial state is determined by the collection of functions ω_{ij} for every edge $j \rightarrow i$. The functional analytic representation of this system is

$$y_i(t) = \sum_j (K_{ij}^* * y_j)(t) + \sum_j (K_{ij}^* * \omega_{ij})(t), \quad t > 0. \quad (13)$$

For example, the block diagram depicted in figure 1 has two nodes – associated to variables y_1 and y_2 – and four edges – whose labels are $K_{11}^* = 0$, $K_{12}^* = e$, $K_{21}^* = 2\delta_0$ and $K_{22}^* = \delta_T$. Then $y_1(t)$, defined for $t > 0$, satisfies

$$\begin{aligned} y_1(t) &= (0 * y_1)(t) + (e * y_2)(t) + (0 * \omega_{11})(t) + (e * \omega_{12})(t) \\ &= (e * y_2)(t) + (e * \omega_{12})(t) \end{aligned}$$

which, since $(e * \omega_{12})(t)$ is a constant c for $t > 0$, is the integral form of the differential equation $\dot{y}_1(t) = y_2(t)$ with initial value $y_1(0) = c$. On the other hand $y_2(t)$, also defined for $t > 0$, satisfies

$$y_2(t) = (2\delta_0 * y_1)(t) + (\delta_T * y_2)(t) + (2\delta_0 * \omega_{21})(t) + (\delta_T * \omega_{22})(t)$$

and since $(2\delta_0 * \omega_{21})(t) = (\delta_T * \omega_{22})(t) = 0$ for any $t > r$, eventually the equation $y_2(t) = 2y_1(t) + y_2(t - T)$ holds and the behavior of the system is ruled by a DDAE system. This is not accidental, since the following result holds:

Theorem. *Every DDAE system has a canonical block diagram representation.*

Proof. Suppose that for some constant vector f

$$\begin{cases} x(t) = (e * A^* * x)(t) + (e * B^* * y)(t) + f \\ y(t) = (C^* * x)(t) + (D^* * y)(t) \end{cases}, \quad t > 0$$

and

$$(x(0^+), x|_{[-r,0]}, y|_{[-r,0]}) = (\phi, \chi, \psi).$$

The vector f is uniquely determined by the initial state. We may define the auxiliary variable $w(t)$ for $t > 0$ by $w(t) = (A^* * x)(t) + (B^* * y)(t)$, and rewrite the first equation of the DDAE system as

$$x(t) = (e * w)(t) + (e * v)(t), \quad t > 0$$

for any function $v : [-r, 0] \mapsto \mathbb{C}^n$ such that $(e * v)(0) = f$. It is now plain that the variable $z(t) = (x(t), w(t), y(t))$ defined for $t > 0$, the matrix kernel and the initial state

$$K^* = \begin{bmatrix} 0 & eI_n & 0 \\ A^* & 0 & B^* \\ C^* & 0 & D^* \end{bmatrix}, \quad \omega_{ij} = (\chi, v, \psi)_j \quad (14)$$

satisfy

$$z_i(t) = \sum_j (K_{ij}^* * z_j)(t) + \sum_j (K_{ij}^* * \omega_{ij}), t > 0. \quad (15)$$

which is the functional analytic representation of a block diagram. ■

3.2 Causality Analysis

We say that a scalar input-output system is *causal* (resp. *strictly causal*) if its convolution kernel is causal (resp. strictly causal). For example, since the kernel of the integrator is the function e , it is strictly causal:

$$e\{0\} = \int_{\{0\}} e(t) dt = 0.$$

A (discrete or pure) delay of $T > 0$ seconds satisfies $y(t) = u(t - T)$; its kernel is $k^* = \delta_T$ and it is strictly causal. Similarly, a distributed delay with bounded delay r , defined by the equation

$$y(t) = \int_{-r}^0 h(\theta)u(t + \theta)d\theta$$

for some locally integrable function $h : [-r, 0] \rightarrow \mathbb{R}$, is also strictly causal. A gain $y(t) = ku(t)$ is causal, but it is strictly causal only if $k = 0$.

Definition – Causality Loop. *A causality loop of a block diagram with kernel matrix K^* of size $n \times n$ is a finite sequence of integers i_0, i_1, \dots, i_j in $\{1, \dots, n\}$ such that $i_0 = i_j$ and for any $\ell \in \{0, \dots, j - 1\}$, the system with kernel $k_\ell^* = K_{i_{\ell+1}i_\ell}^*$ is not strictly causal.*

The block diagram of figure 1 has no causality loop, since the only element of its matrix K^* which is not strictly causal is $K_{21}^* = 2\delta_0$ which is not on the matrix diagonal. The following result is applicable in this case:

Theorem. *If the block diagram representation (14)-(15) of DDAE system (5) has no causality loop then $D\{0\}$ is nilpotent.*

Proof. Let K^* be the kernel matrix of the block diagram representation of DDAE system (5). In this proof, for any matrix-valued measure M , we denote M_0 the matrix $M\{0\}$. Let $\mathcal{A}[K]$ be the adjacency matrix of the block diagram directed graph, stripped off its strictly causal edges:

$$\mathcal{A}[K]_{ij} = \begin{cases} 1 & \text{if } [K_0]_{ij} \neq 0 \\ 0 & \text{otherwise.} \end{cases}$$

The graph-theoretic concept of loop corresponds to our definition of causality loop. Element (i, j) of the p -th power $\mathcal{A}[K]^p$ of the adjacency matrix is the

number of distinct paths with p edges between vertices j and i . If there is a loop of length p in the block diagram, there is also a loop whose length is an arbitrary multiple of p , hence the adjacency matrix is not nilpotent. Conversely, if the adjacency matrix is not nilpotent, there is a pair (i, j) such that $\mathcal{A}[K]_{ij}^n \neq 0$; since any path between j and i goes through $n + 1$ vertices and the graph has only n distinct vertices, at least one of them is repeated: the path necessarily contains a loop.

Now, if $\mathcal{A}[K]^p = 0$ for some $p \in \mathbb{N}^*$ then $[K_0]^p = 0$. Indeed, since

$$\mathcal{A}[K]_{ij}^p = \sum_{\ell_1, \dots, \ell_{p-1}} \mathcal{A}[K]_{i\ell_1} \dots \mathcal{A}[K]_{\ell_{p-1}j}$$

and all elements of the adjacency matrix are non-negative, if its p -th power is zero, there is for each term in the sum above at least one factor $\mathcal{A}[K]_{\nu\mu}$ which is zero and thus $[K_0]_{\nu\mu}^p = 0$. Consequently, since

$$[K_0]_{ij}^p = \sum_{\ell_1, \dots, \ell_{p-1}} [K_0]_{i\ell_1} \dots [K_0]_{\ell_{p-1}j},$$

the p -th power of K_0 is also zero. But for any integer $p \geq 2$, we have

$$[K_0]^p = \begin{bmatrix} 0 & 0 & 0 \\ A_0 & 0 & B_0 \\ C_0 & 0 & D_0 \end{bmatrix}^p = \begin{bmatrix} 0 & 0 & 0 \\ B_0[D_0]^{p-2}C_0 & 0 & B_0[D_0]^{p-1} \\ [D_0]^{p-1}C_0 & 0 & [D_0]^p \end{bmatrix}.$$

Consequently, if the block diagram has no causality loop, $\mathcal{A}[K]$ is nilpotent and the matrix $D_0 = D\{0\}$ is also nilpotent. ■

Note that the absence of causality loop does not necessarily provide explicitness: for example, the delay operator D of the DDAE system governed by $y_1(t) = y_2(t)$ and $y_2(t) = y_1(t - 1)$ satisfies

$$D\{0\} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

which is not zero. However, it is plain that its block-diagram representation has no causality loop: it contains only one subsystem which is not strictly causal and this subsystem connects variable y_1 to the other variable y_2 . This is sufficient to ensure its well-posedness:

Corollary. *A DDAE system without causality loop is well-posed.*

Proof. If DDAE system (5) has no causality loop, the matrix $D\{0\}$ is nilpotent and $[D\{0\}]^m = 0$. Consequently, the matrix $I_m - D\{0\}$ is invertible:

$$[I_m - D\{0\}]^{-1} = I_m + D\{0\} + \dots + [D\{0\}]^{m-1}.$$

The well-posedness theorem of the previous section provides the conclusion. ■

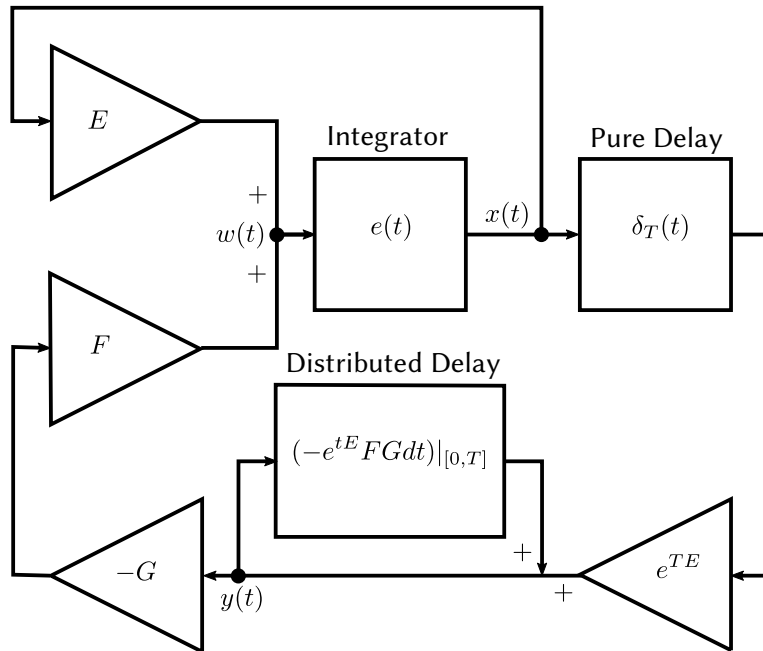


Figure 2: FSA closed-loop system block diagram. Each rectangular block characterizes an input-output system by its (function or measure) convolution kernel. A triangle with symbol K represents a gain: a system whose output $y(t)$ and input $u(t)$ satisfy $y(t) = Ku(t)$.

3.3 Block Diagram of the FSA System

In the block diagram representation of system (3), the auxiliary variable w satisfies

$$w(t) = Ex(t) - FGy(t), \quad t > 0.$$

Since $x(t) = e * w(t) + e * v(t)$ is the only (system of) equation(s) with some variable w_j in the right-hand side or some variable x_i in the left hand-side, there is a unique outgoing edge of w_j (with a nonzero label) which is also the unique incoming edge of the variable x_j . This label is the function e , strictly causal, thus no causality loop may contain either of the components of the state variables w or x . Now the possible causality loops that could remain would contain only vertices which are components of the state variable y . But the edge between y_j and y_i is H_{ij}^* where $H^* = (-e^{tE}FGdt)|_{[0,T]}$. Again, this kernel is a function and hence is strictly causal. Thus no such loop may exist and the initial-value problem is well-posed.

This analysis may also be carried out graphically on figure 2. Once the integrator, the pure delay and the distributed delay – which are strictly causal subsystems – are removed from the diagram, it is plain that there is no loop and thus that the system has no causality loop.

4 Stability

4.1 Test in the Laplace Domain

There is a general and simple test in the Laplace domain for the exponential stability of a DDAE system. It can also be used to determine its growth bound, defined as the infimum of the σ^+ such that for any initial value, there is a $\kappa > 0$ such that the solution satisfies $\|(x(t), x_t, y_t)\|_X \leq \kappa e^{\sigma^+ t}$ for any $t > 0$; exponential stability corresponds to a negative growth bound.

A few definitions are in order first: the Laplace transform $\mathcal{L}L$ of a compactly supported measure L is the function defined on \mathbb{C} by

$$\mathcal{L}L(s) = \int e^{-st} dL(t).$$

The Laplace transform of L^* satisfies $\mathcal{L}L^*(s) = L(t \mapsto e^{st})$. The *characteristic matrix* of the system is then defined as

$$\Delta(s) = \begin{bmatrix} sI_n & 0 \\ 0 & I_m \end{bmatrix} - \mathcal{L} \begin{bmatrix} A^* & B^* \\ C^* & D^* \end{bmatrix} (s) \quad (16)$$

and its *characteristic function* as its determinant.

We may now state the criterion:

Theorem. *The growth bound σ of a DDAE system without causality loop is:*

$$\sigma = \sup \{ \Re(s) \mid s \in \mathbb{C}, \det \Delta(s) = 0 \}. \quad (17)$$

The proof of this theorem is carried out in (Boisg erault 2013) in the case of explicit systems. Here, we expose the proof in the general case.

Lemma. *DDAE systems (5) and (10) have simultaneously singular characteristic matrices $\Delta(s)$ and $\Delta^\dagger(s)$.*

Proof. Let $J = I_m - D\{0\}$, $E = J^{-1}C$ and $F = J^{-1}(D - D|_{\{0\}})$; we have

$$\mathcal{L}E^* = J^{-1}\mathcal{L}C^* \quad \text{and} \quad \mathcal{L}F^* = J^{-1}(\mathcal{L}D^* - D\{0\})$$

and since $I_m + J^{-1}D\{0\} = J^{-1}$, $I_m - \mathcal{L}F^* = J^{-1}(I_m - \mathcal{L}D^*)$. Finally,

$$\Delta^\dagger(s) = \begin{bmatrix} sI_n & 0 \\ 0 & I_m \end{bmatrix} - \mathcal{L} \begin{bmatrix} A^* & B^* \\ E^* & F^* \end{bmatrix} (s) = \begin{bmatrix} I_n & 0 \\ 0 & J^{-1} \end{bmatrix} \Delta(s)$$

and the characteristic matrices are simultaneously singular. ■

The determinant and adjugate of the characteristic matrix Δ^\dagger have a quasipolynomial structure:

$$\det \Delta^\dagger(s) = \sum_{i=0}^n c_i(s)s^i, \quad \text{adj} \Delta^\dagger(s) = \sum_{i=0}^n C_i(s)s^i \quad (18)$$

where the c_i (resp. C_i) are entire functions (resp. matrices of entire functions) bounded on any right-hand plane. It is plain that the leading coefficient of the characteristic function is given by $c_n(s) = \det \Delta_0(s)$ where Δ_0 is the characteristic matrix of the system $y(t) = Fy_t$.

Lemma – Zero Clusters. *Let H_σ be the open right half-plane $\{s \in \mathbb{C} \mid \Re(s) > \sigma\}$ and let Z_η be the set of points whose distance to the zeros of $\det \Delta_0$ is at most η .*

For any $\sigma \in \mathbb{R}$ and $\epsilon > 0$, there is a $\eta > 0$ such that:

- i. *any connected component Λ of the set Z_η is bounded,*
- ii. *Λ is a subset of $H_{\sigma-\epsilon}$ whenever $\Lambda \cap H_\sigma \neq \emptyset$,*
- iii. *$|\det \Delta_0|$ has a positive lower bound on $H_{\sigma-\epsilon} \setminus Z_\eta$.*

The proof (from (Boisg erault 2013)) is given in the appendix for the sake of completeness.

Lemma – Characteristic Function Zeros. *Let $\sigma \in \mathbb{R}$. If the function $\det \Delta_0$ has an infinite number of zeros on H_σ , the function $\det \Delta^\dagger$ has an infinite number of zeros on $H_{\sigma-\epsilon}$ for any $\epsilon > 0$.*

Proof. Suppose that $\det \Delta_0$ has an infinite number of zeros in H_σ and let $\eta > 0$ be as in the zero clusters lemma. The zeros of $\det \Delta_0$ are isolated, thus the collection of components Λ of Z_η such that $\Lambda \cap H_\sigma \neq \emptyset$, which is locally finite by construction, is also infinite: for any compact set K , there is a Λ such that $\Lambda \cap K = \emptyset$.

Since $\det \Delta^\dagger(s)$ has a quasi-polynomial structure with leading coefficient $\det \Delta_0(s) s^n$ and since $\det \Delta_0(s)$ has a positive lower bound on $H_{\sigma-\epsilon} \setminus Z_\eta$, there is a compact set K such that $|\det \Delta^\dagger(s) - s^n \det \Delta_0(s)| < |s^n \det \Delta_0(s)|$ whenever $s \in H_{\sigma-\epsilon} \setminus Z_\eta$ and $s \notin K$. Rouché's theorem is applicable to any of the components Λ in the complement of K , each of which contains at least one zero of $\det \Delta_0$. Thus, $\det \Delta^\dagger$ has an infinite number of zeros in $H_{\sigma-\epsilon}$. ■

We can associate to the initial-value problem (5), or (10) which is equivalent, a one-parameter family of operators $(\exp(\mathcal{A}t))_{t \geq 0}$ defined by

$$(x(t^+), x_t, y_t) = \exp(\mathcal{A}t)(\phi, \chi, \psi) \text{ for } t \geq 0, (\phi, \chi, \psi) \in X.$$

It is a strongly continuous semigroup on the product space X ; its infinitesimal generator \mathcal{A} is defined by

$$\mathcal{A}(\phi, \chi, \psi) = (A\chi + B\psi, \dot{\chi}, \dot{\psi})$$

on the domain

$$\{(\phi, \chi, \psi) \in \mathbb{C}^n \times W^{1,2}([-r, 0], \mathbb{C}^{n+m}) \mid \chi(0) = \phi, \psi(0) = E\chi + F\psi\}.$$

The resolvent operator $(sI - \mathcal{A})^{-1}$ exists if and only if $\Delta^\dagger(s)$ is non-singular and moreover, for any real number σ there are constants κ_σ and λ_σ such that

$$\|(sI - \mathcal{A})^{-1}\| \leq \kappa_\sigma \|\Delta^\dagger(s)^{-1}\| + \lambda_\sigma \quad (19)$$

if $\Re s \geq \sigma$ and $\det \Delta^\dagger(s) \neq 0$ (see Salamon (1984)).

We may now prove the main result:

Proof. Let $s(\mathcal{A})$ be the spectral bound of \mathcal{A} . We show that for any $\sigma > s(\mathcal{A})$, $\|\Delta^\dagger(s)^{-1}\|$ is bounded on \overline{H}_σ . Thus, by inequality (19), $\|(sI - \mathcal{A})^{-1}\|$ is similarly bounded and the Gearhart-Prüss theorem proves the result.

The quasi-polynomial structure of the adjugate matrix provides for some $\kappa \geq 0$

$$\|\text{adj } \Delta^\dagger(s)\| \leq \kappa(1 + |s|^n)$$

on \overline{H}_σ . Since the resolvent operator is defined on $H_{\sigma-\epsilon}$ for any $\epsilon > 0$ such that $s(\mathcal{A}) < \sigma - \epsilon$, $\det \Delta^\dagger$ has no zero in $H_{\sigma-\epsilon}$. Thus, by the characteristic function zeros lemma, $\det \Delta_0$ has at most a finite number of zeros on $H_{\sigma-\epsilon/2}$ and thus on \overline{H}_σ . By the zero clusters lemma, on \overline{H}_σ and away from these zeros, $|\det \Delta_0|$

has a positive lower bound κ' . It follows from the quasipolynomial structure of $\det \Delta^\dagger$ that

$$|\det \Delta^\dagger(s)| \geq \frac{\kappa'}{2}(1 + |s|^n)$$

on \overline{H}_σ except on some compact set K ; by continuity of $\det \Delta^\dagger$, this estimate still holds on all of \overline{H}_σ with a possibly smaller κ' . Finally, for any $s \in \overline{H}_\sigma$, $\|\Delta^\dagger(s)^{-1}\| = \|\text{adj } \Delta^\dagger(s)\|/|\det \Delta^\dagger(s)| \leq 2\kappa/\kappa'$. ■

4.2 Stability of the FSA System

Since the kernels of FSA system (3) are given by (11) and (12), its characteristic matrix satisfies

$$\Delta(s) = \left[\begin{array}{c|c} sI_n - E & FG \\ \hline -e^{(sI_n - E)T} & I_n + P(s)FG \end{array} \right]$$

where $P(s)$ is the analytic extension to \mathbb{C} of the meromorphic function

$$s \mapsto [sI_n - E]^{-1}(I_n - e^{(sI_n - E)T}).$$

A straightforward computation provides

$$\Delta(s) \left[\begin{array}{c|c} I_n & 0 \\ \hline I_n & I_n \end{array} \right] = \left[\begin{array}{c|c} sI_n - E + FG & FG \\ \hline P(s)(sI_n - E + FG) & I_n + P(s)FG \end{array} \right]$$

and the right-hand side of this equation can be factored into

$$\left[\begin{array}{c|c} I_n & 0 \\ \hline P(s) & I_n \end{array} \right] \left[\begin{array}{c|c} I_n & FG \\ \hline 0 & I_n \end{array} \right] \left[\begin{array}{c|c} sI_n - E + FG & 0 \\ \hline 0 & I_n \end{array} \right].$$

Thus, the characteristic function satisfies

$$\det \Delta(s) = \det(sI_n - E + FG)$$

and its roots are the eigenvalues of $E - FG$. When the delay-free system $\dot{x}(t) = Ex(t) + Fu(t)$ is controllable, finite-dimensional control theory provides a gain matrix G that assigns these eigenvalues to n arbitrary locations in the complex plane, and thus we may achieve a “finite spectrum assignment” of the closed-loop system. In particular, it’s possible to select a gain matrix G which exponentially stabilizes system (3).

Appendix

We provide in this appendix a proof of the zero clusters lemma. First we derive some properties of $\det \Delta_0$ by expressing it as the Laplace transform of a complex

measure μ . Let Σ_m be the set of permutations of $\{1, \dots, m\}$ and

$$\det_* M = \sum_{\sigma \in \Sigma_m} \text{sgn}(\sigma) M_{1, \sigma(1)} * \dots * M_{m, \sigma(m)}.$$

As $\Delta_0(s) = I_m - \mathcal{L}F^*(s)$, $\det \Delta_0 = \mathcal{L}\mu$ where $\mu = \det_*(\delta_0 I_m - F^*)$. The complex measure μ is a sum of convolution products of m complex measures supported on $[0, r]$, hence it is supported on $[0, mr]$. Consequently, $\det \Delta_0$ is an entire function that satisfies the inequality

$$|\det \Delta_0(s)| \leq |\mu|([0, mr]) \max(1, \exp(-\Re(s)mr)). \quad (20)$$

Since $F\{0\} = 0$, we also have $\mu\{0\} = 1$, which yields

$$\lim_{\Re s \rightarrow +\infty} \det \Delta_0(s) = 1. \quad (21)$$

We may now proceed to the proof:

Proof (Zero Clusters Lemma). The function $z \mapsto \det \Delta_0(iz)$ meets the assumptions of theorem VIII in (Levinson 1940) and is of exponential type mr . Thus, the number of distinct zeros $N(\rho)$ of $\det \Delta_0$ whose modulus is less than ρ is such that $\limsup_{\rho \rightarrow +\infty} N(\rho)/\rho \leq 2mr/\pi$. If the component of Z_η that contains a zero s is unbounded, there are at least $n + 1$ zeros in the closed disk centered at s of radius $2\eta n$ and thus $\limsup_{\rho \rightarrow +\infty} N(\rho)/\rho \geq 1/2\eta$. Consequently, if $\eta < \pi/4mr$, every connected component of Z_η is bounded.

The proofs of statements ii. and iii. use a similar argument. In each case we consider a sequence s_n of complex numbers of bounded real part and the functions $f_n(s) = \det \Delta_0(s + i\Im s_n)$. Since inequality (20) holds, these functions are locally uniformly bounded, hence there is a subsequence of s_n which converges to some real number x and, by Montel's theorem, a corresponding subsequence of f_n that converges locally uniformly to an entire function f_∞ . By (21), f_∞ is not identically zero, thus by Hurwitz's theorem, if m is the multiplicity of x if $f_\infty(x) = 0$, or 0 otherwise, for any sufficiently small $\delta > 0$, $\det \Delta_0$ has exactly m zeros in the open disk $B(s_n, \delta)$ for an infinite number of values of n .

To prove ii., we assume the existence of a sequence Λ_n of connected components of Z_{η_n} where $\eta_n \rightarrow 0$ and such that $\Lambda_n \cap H_\sigma \neq \emptyset$ but $\Lambda_n \not\subset H_{\sigma-\epsilon}$. The Λ_n are eventually bounded and there is a tuple of zeros in Λ_n with a first element such that $\Re s + \eta_n > \sigma$, a last element such that $\Re s - \eta_n \leq \sigma - \epsilon$ and a distance between consecutive points that is at most $2\eta_n$. Let s_n to be the first element of this tuple; the real part of this sequence is bounded by (21). But for any $\delta > 0$, the number of zeros in $B(s_n, \delta)$ converges to $+\infty$, a contradiction with the previous paragraph.

To prove iii., we assume the existence of a sequence s_n in $H_{\sigma-\epsilon} \setminus Z_\eta$ such that $\det \Delta_0(s_n) \rightarrow 0$ when $n \rightarrow +\infty$; the real part of this sequence is bounded and

we may select s_n such that $\Re s_n \rightarrow x$. Now, $f_\infty(x) = \lim_{n \rightarrow +\infty} f_n(\Re s_n) = 0$, thus for some $\delta < \eta$ and for some value of n , there is at least one zero of $\det \Delta_0$ in $B(s_n, \delta)$, which is a contradiction. ■

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