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Bond graph/Digraph matching procedure and symbolic determination of state space descriptor systems

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Abstract In the context of structural analysis of descriptor systems (i.e. $E \dot{x}(t) = Ax(t) + Bu(t)$, with $E$ singular, also called singular systems, generalized state space systems, or differential-algebraic systems [1, 2, 3]), this paper proposes an extension of existing procedures to determine the state-space descriptor form from the bond graph model, relaxing simplified assumptions. Using this state-space representation, we also propose to build the exact corresponding digraph of a (bi)causal bond graph. This correspondence is used to rigorously adapt, by means of the digraph tool, the graphical computation of the characteristic polynomial $\det(sE - A)$ or the determinant of the Rosenbrock matrix $\det(P(s))$ directly from the bond graph model (not presented in this paper).

Key words Structural Analysis, Descriptor Systems, Bicausal Bond Graph, Causal Cycles and Causal Paths, Digraph.

1 Introduction

This paper is concerned with the structural analysis of bond graph models, aiming at extending the existing framework on state-space determination, and proposing a procedure that builds the exact corresponding directed-graph (i.e. digraph) of a (bi)causal bond graph. This extends the symbolic determination of state space representation procedures on LTI regular systems [4, 5, 6, 7, 8] to the class of LTI descriptor systems with regular matrix pencil. From the algebraic point of view, such a system can be described by:

$$\Sigma_d : \begin{cases} E \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

where $x(t) \in \mathbb{R}^n$ is the descriptor state vector, $u(t) \in \mathbb{R}^m$ the input vector, $y(t) \in \mathbb{R}^p$ the output vector and $E, A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}, C \in \mathbb{R}^{p \times n}, D \in \mathbb{R}^{p \times m}$, with a regular matrix pencil $(sE - A)$, that is $\det(sE - A) \neq 0$, and $s$ the Laplace variable.

Descriptor systems have been widely studied in the literature [1, 2] to answer to mechatronic modeling
requirements for representing algebraic constraint equations that may occur in electrical networks [2, 3] or mechanical systems [9], leading to Differential-Algebraic Equations (DAE) with singular matrix E in (1). Descriptor systems are also called singular system, generalized state space systems or differential algebraic systems [3]. Structural analysis on linear descriptor systems has been theoretically formalized on digraphs [10], on matroids [11] and on bond graphs [12, 13].

The class of systems targeted in this paper is the one represented by causal or bicausal bond graph models, without necessary integral preferred causality assignment, potentially with zero-order causal paths (ZCP) of type 1ZCP, 2ZCP, 3ZCP, 4ZCP in the classification of [14, 15], and R and/or I/C fields. For obvious reasons the only assumption enabling the proposed procedure to be applied is the respect of global causality constraints which guaranty a solvable junction structure (and excludes for instance ZCP of class 5). The regularity of the matrix pencil in (1) will be discussed and proven for this class of bond graph models.

The matching proposed in this paper is used to generalize graphical procedures on bond graph from the ones on digraph [4, 5, 13, 16, 17]. In fact, using the digraph and the related frameworks on regular [18, 19] and descriptor systems [10], this matching enables the characteristic polynomial det(sE − A) or, in the case of square systems (m = p), the determinant of the Rosenbrock matrix 1 det(P(s)) directly from the bond graph model to be determined [21]. These determinants are used for instance to graphically determine the structure at infinity of the system [22] or to identify the presence of impulsive modes on power or energy variables [21].

The present paper is organized as follows. Section 2 brings some background on the junction structure matrix and proposes some extensions to account for any power-variable of the system. Section 3 proposes to extend the symbolic determination of the state space linear descriptor system from a (bi)causal bond graph model, highlighting an important result about the regularity of its pencil sE − A. Section 4 presents the graphical procedure to obtain a digraph model that exactly matches to the bond graph model, and gives an illustrative example. The conclusion outlines some targeted applications and suggests some directions for future works.

2 Junction Structure Matrix

2.1 Original form

The physical and energetical characteristics of the bond graph representation are described in this paper with the junction structure matrix approach, originally introduced by Rosenberg [23, 24]. This approach enables a global consideration of the junction structure of the model and its causal implications on storage elements, dissipative elements, sources and sensors. It suits to the graphical approach of the structural analysis on bond graph. Also it is used in the present paper to determine the state-space equations for the class of bond graph models previously introduced.

From a condensed representation of the junction structure (Figure 1), the storage elements, the dissipation elements, the sources, the sensors, and the significant vectors of the causal form, Rosenberg has established the expression of the outputs of the junction structure in terms of the inputs. This leads to a state-space representation, regular (2) or singular (1), depending on the topology of the system and the level of equation substitution that could be achieved [23].

$$\Sigma : \begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

(2)

where $x(t) \in \mathbb{R}^n$ is the regular state vector, $u(t) \in \mathbb{R}^m$ the input vector, and $y(t) \in \mathbb{R}^p$ the output vector.

In Figure 1b condensed representation $x_I$ and $x_D$ are the vectors of energy variables for the storage elements respectively in integral and derivative causality, $z_I$ and $z_D$, the associated vectors of co-energy variables, and, $D_{in}$ and $D_{out}$, the vector of power variables respectively ingoing and outgoing the dissipative elements. The constitutive laws of the $I$, $C$ and $R$ fields can be expressed in terms of these vectors:

- with the assumption that the laws are explicit, the storage field can be characterized by

$$\begin{pmatrix} z_I \\ x_D \end{pmatrix} = \Phi_S(x_I, z_D)$$

(3)

Considering, in the present paper, mixed causality with linear laws leads to the following simplification [12]:

$$\begin{pmatrix} z_I \\ x_D \end{pmatrix} = \begin{pmatrix} S_i & S_{id} \\ S_{di} & S_d \end{pmatrix} \begin{pmatrix} x_I \\ z_D \end{pmatrix}$$

(4)

where $\begin{pmatrix} S_i & S_{id} \\ S_{di} & S_d \end{pmatrix}$ is invertible.
- the resistive field can be characterized by

\[ \mathbf{D}_{out} = \Phi_L (\mathbf{D}_{in}) \]  

that can be simplified in the linear case as:

\[ \mathbf{D}_{out} = \mathbf{L} \cdot \mathbf{D}_{in} \]  

With the assumption that there is no zero-order causal path of class 5 (5ZCP) [14, 15] leading to non solvable junction structure [12], if exist, are solved. With a more general view, this condition corresponds, on one hand, to the fact that the global Kirchhoff constraints on cycles and co-cycles of the junction structure are respected [25], and on the other hand that the basis order variable of junction structures is also respected [26]. With the constitutive laws (4) and (6), the equation (7) can be developed under several forms, e.g. generalized state-space form [23], standard implicit form [12], or descriptor state-space form [12, 27, 28, 29], the later one being of interest in this paper.

2.2 Extended form

In this section, we propose to extend the output vector of the junction structure with any power variable of the system. The main motivation of this extension is the important simplification of the bond graph inspection, by "breaking" causal loops when extracting the descriptor state-space form from (7).

The related extended significant vector is exposed in Figure 2, highlighting the variables \( x_\lambda \), that could be components of \( \mathbf{D}_{in} \) and/or \( \mathbf{D}_{out} \) for instance, or any other power variables on a bond in the junction structure. With

Figure 1: Condensed representation of multiport systems [23]

Figure 2: Condensed representation of multiport systems, considering additional power variables of the junction structure.
these new variables, the equation (7) becomes:

\[
\begin{pmatrix}
\dot{x}_I \\
\dot{z}_D \\
D_{in} \\
x_\lambda
\end{pmatrix} =
\begin{pmatrix}
S_{I1} & S_{I2} & S_{I3} & S_{I4} & S_{I5} \\
S_{D1} & S_{D2} & S_{D3} & S_{D4} & S_{D5} \\
S_{L1} & S_{L2} & S_{L3} & S_{L4} & S_{L5} \\
S_{S1} & S_{S2} & S_{S3} & S_{S4} & S_{S5}
\end{pmatrix}
\begin{pmatrix}
z_I \\
x_\lambda \\
z_D \\
x_\lambda
\end{pmatrix}
\]

\[
y = \begin{pmatrix}
S_{Z1} & S_{Z2} & S_{Z3} & S_{Z4} & S_{Z5}
\end{pmatrix}
\begin{pmatrix}
z_I \\
x_\lambda \\
z_D \\
x_\lambda
\end{pmatrix}
\]  \hspace{1cm} \text{(8)}

One could note that the sub-matrices from (8) are generally not the same as the ones of (7). Furthermore, the differences arise from the non-resolution a priori of some causal loops - the ones involving the variables from \(x_\lambda\) in (8) - which could be 2ZCP, 3ZCP and/or 4ZCP [14, 15], whereas these loops are necessarily solved in (7).

3 Symbolic determination of the descriptor state space form

In this section, we propose a LTI descriptor state space form (1) symbolically determined from the bond graph model, using the linear storage field (4) and linear dissipation field (6) laws, together with the extended consideration on input and output relations from the junction structure proposed in (8).

The constitutive laws of dissipation fields (6) enable the outputs from the resistive elements \(D_{out}\) with respect to the other inputs of the junction structure to be expressed,

\[
D_{out} = (I - L S_{33})^{-1}
\]

\[
\begin{align*}
L S_{31} z_I \\
+ L S_{32} \dot{x}_D \\
+ L S_{34} x_\lambda \\
+ L S_{35} u,
\end{align*}
\]  \hspace{1cm} \text{(9)}

assuming that the 2ZCP and 3ZCP involving the resistive elements \(R\) are solvable, to insure the invertibility of the matrix \((I - L S_{33})\). Then, introducing (9) in (8), we obtain:

\[
\begin{pmatrix}
\dot{x}_I \\
\dot{z}_D \\
x_\lambda
\end{pmatrix} =
\begin{pmatrix}
S_{I1} & S_{I2} & S_{I3} & S_{I4} & S_{I5} \\
S_{D1} & S_{D2} & S_{D3} & S_{D4} & S_{D5} \\
S_{L1} & S_{L2} & S_{L3} & S_{L4} & S_{L5} \\
S_{S1} & S_{S2} & S_{S3} & S_{S4} & S_{S5}
\end{pmatrix}
\begin{pmatrix}
z_I \\
x_\lambda \\
z_D \\
x_\lambda
\end{pmatrix}
\]

\[
y = \begin{pmatrix}
S_{Y1} & S_{Y2} & S_{Y3} & S_{Y4} & S_{Y5}
\end{pmatrix}
\begin{pmatrix}
z_I \\
x_\lambda \\
z_D \\
x_\lambda
\end{pmatrix}
\]  \hspace{1cm} \text{(10)}

Using the constitutive laws of storage fields (4), the equation (10) can be re-written in terms of energy variables \(x_I\) and \(x_D\), power variables \(x_\lambda\), and in term of the output vector \(y\). In a first step, the variable \(z_D\) are expressed as,

\[
z_D = (I - S_{D4} S_{id})^{-1}
\]

\[
\begin{align*}
S_{D1} x_I \\
+ S_{D2} \dot{x}_D \\
+ S_{D4} x_\lambda \\
+ S_{D5} u
\end{align*}
\]  \hspace{1cm} \text{(12)}

Then, using (12), the equation (10) can be put on the LTI descriptor state space form (1) as follows:

\[
\begin{pmatrix}
I_I & E_{ID} & 0 \\
0 & E_D & 0 \\
0 & E_{\lambda D} & 0
\end{pmatrix}
\begin{pmatrix}
\dot{x}_I \\
\dot{z}_D \\
x_\lambda
\end{pmatrix} =
\begin{pmatrix}
A_I & 0 & A_{I\lambda} \\
A_{D1} & I_D & A_{D\lambda} \\
A_{\lambda I} & 0 & I_\lambda + A_{\lambda}\lambda
\end{pmatrix}
\begin{pmatrix}
x_I \\
x_D \\
x_\lambda
\end{pmatrix}
+ \begin{pmatrix}
B_I \\
B_D \\
B_{\lambda}
\end{pmatrix} u
\]

\[
y = \begin{pmatrix}
C_I & C_{D\frac{d\lambda}{dt}} & C_\lambda
\end{pmatrix}
\begin{pmatrix}
x_I \\
x_D \\
x_\lambda
\end{pmatrix}
+ \begin{pmatrix}
D\end{pmatrix} u
\]  \hspace{1cm} \text{(13)}
with \( H_D = (I - S_{DI}S_{sd})^{-1} \), and

\[
\begin{align*}
E_{ID} &= -[S_{ID} + S_{II}S_{sd}H_DS_{DD}] \\
E_D &= [S_{II}H_DS_{DD}] \\
E_{\lambda D} &= [S_{\lambda D} + S_{\lambda I}S_{sd}H_DS_{DD}] \\
A_I &= [S_{II}S_I + S_{II}S_{sd}H_DS_{DD}S_{DI}] \\
A_{\lambda I} &= [S_{II}S_{\lambda I}H_DS_{D\lambda}] \\
A_{DI} &= -[S_{dI} + S^{-1}_dH_DS_{DD}S_{DI}] \\
A_{D\lambda} &= -[S_{dI}H_DS_{D\lambda}] \\
A_M &= -[S_{\lambda I}S_{I} + S_{\lambda I}S_{sd}H_DS_{DD}S_{DI}] \\
A_{\lambda} &= -[S_{\lambda I}S_{I} + S_{\lambda I}S_{sd}H_DS_{D\lambda}] \\
B_I &= [S_{IU} + S_{II}S_{sd}H_DS_{DU}] \\
B_D &= -[S^{-1}_dH_DS_{DU}] \\
B_{\lambda} &= -[S_{\lambda I}S_{I} + S_{\lambda I}S_{sd}H_DS_{DU}] \\
C_I &= [S_{YI}S_I + S_{YI}S_{sd}H_DS_{DD}S_{DI}] \\
C_D &= [S_{YD} + S_{YI}S_{sd}H_DS_{DD}] \\
C_{\lambda} &= [S_{Y\lambda I}S_{I} + S_{Y\lambda I}S_{sd}H_DS_{D\lambda}] \\
D &= [S_{YU} + S_{YI}S_{sd}H_DS_{DU}]
\end{align*}
\]

The LTI state-space form (13) is thus generic to any causal or bicausal bond graph model with linear constitutive laws, where the only assumption is that the junction structure is solvable. The bond graph model can have fields of type \( I, C, IC \) and/or \( R \), zero-order causal paths of type \( 1ZCP, 2ZCP, 3ZCP, 4ZCP \) [14, 15], and integral, derivative or mixed preferred causality assignment. An important result is that the pencil \((sE - A)\) of (13) is regular, as we structurally have \( \det(sE - A) \neq 0 \) in (13). This justifies the restriction of the algebraic framework to this class of systems in this paper (equation (1)), and also ensures the existence and theunicity of a solution [2]. It can be underlined that the form (13) is consistent with the several simplified forms previously mentioned [7, 12, 13, 17, 28, 29, 30], and thus proposes a generic and common framework to these simplified cases.

The form (13) will be used in section 4 to develop the procedure to build the matching digraph of a bond graph model (proposition 1).

4 Bond graph/ Digraph matching

In the present paper, the directed-graph (digraph) characterization of square matrices is understood as the Cauchy-Coates representation [10, 18, 19] \(^2\). In the sequel, considering the digraph representation of the matrix pencil \( sE - A \) of (13), an edge between two vertices related to the matrix \( E \) (resp. \( A \)) is called an \( E \)-edge (resp. \( A \)-edge). By extension, considering the complete digraph of the system (13), an edge with (resp. without) a LAPLACE operator is called an \( E \)-edge (resp. \( A \)-edge).

4.1 Generic digraph construction from a bond graph model

4.1.1 Procedure

**Proposition 1.** Considering the (bi)causal bond graph model representation of LTI descriptor system (13), the matching digraph - represented in Figure 3 - is obtained from the following steps:

1. Each energy variable related to a storage element with integral causality, each energy variable related to a storage element with derivative causality, each power variable chosen in the state vector, each input variable related to a flow or effort source and each output variable related to a flow or effort sensor is respectively represented by a state vertex \( x_I \), \( x_D \), and \( x_{\lambda} \), input vertex \( u_i \) and output vertex \( y_i \) on the digraph.

2. **Submatrix** \( I_1 \): for each state vertex \( x_I \) of the digraph is associated a unitary cycle of type \( E \)-edge, with a unitary weight.

3. **Submatrix** \( E_{ID} \): on the bond graph, each causal path between the derivative of an energy variable \( x_D \), and the derivative of an energy variable \( x_I \) is represented on the digraph by an \( E \)-edge between the state vertex \( x_D \), and the state vertex \( x_I \), with a weight \( e_{ID} \), equal to the opposite of the static gain of the causal path.

4. **Submatrix** \( E_D \): each causal path between the derivative of an energy variable \( x_D \), and an energy variable \( x_{\lambda} \), is represented on the digraph by an \( E \)-edge between the state vertex \( x_D \), and the state vertex \( x_{\lambda} \), with a weight \( e_{DI} \), equal to the static gain of the causal path.

5. **Submatrix** \( E_{\lambda D} \): each causal path between the derivative of an energy variable \( x_D \), and a power variable \( x_{\lambda} \), is represented on the digraph by an \( E \)-edge between the state vertex \( x_D \), and the state vertex \( x_{\lambda} \), with a weight \( e_{\lambda D} \), equal to the static gain of the causal path.

6. **Submatrix** \( A_I \): each causal path between an energy variable \( x_I \), and the derivative of an energy variable
10. Submatrix $A_{1A}$: each causal path between a power variable $x_{\lambda}$ and the derivative of an energy variable $x_{I}$ is represented on the digraph by an $A$-edge between the state vertex $x_{\lambda}$ and the state vertex $x_{I}$, with a weight $a_{1\lambda}$, equal to the static gain of the causal path.

7. Submatrix $A_{1A}$: each causal path between a power variable $x_{\lambda}$ and the derivative of an energy variable $x_{I}$ is represented on the digraph by an $A$-edge between the state vertex $x_{\lambda}$ and the state vertex $x_{I}$, with a weight $a_{1\lambda}$, equal to the static gain of the causal path.

8. Submatrix $A_{DI}$: each causal path between an energy variable $x_{I}$ and an energy variable $x_{D}$ is represented on the digraph by an $A$-edge between the state vertex $x_{I}$ and the state vertex $x_{D}$, with a weight $a_{DI}$, equal to the opposite of the static gain of the causal path.

9. Submatrix $I_{D}$: to each state vertex $x_{D}$ of the digraph is associated a unitary cycle of type $A$-edge, with a unitary weight.

10. Submatrix $A_{DI}$: each causal path between a power variable $x_{\lambda}$ and the derivative of an energy variable $x_{D}$ is represented on the digraph by an $A$-edge between the state vertex $x_{\lambda}$ and the state vertex $x_{D}$, with a weight $a_{D\lambda}$, equal to the opposite of the static gain of the causal path.

11. Submatrix $A_{1\lambda}$: each causal path between an energy variable $x_{I}$ and a power variable $x_{\lambda}$ is represented on the digraph by an $A$-edge between the state vertex $x_{I}$ and the state vertex $x_{\lambda}$, with a weight $a_{1\lambda}$, equal to the opposite of the static gain of the causal path.

12. Submatrix $I_{\lambda}$: to each state vertex $x_{\lambda}$ of the digraph is associated a unitary cycle of type $A$-edge, with a unitary weight.

13. Submatrix $A_{\lambda}$: each causal path between a power variable $x_{\lambda}$ and a power variable $x_{\lambda}$ is represented on the digraph by an $A$-edge between the state vertex $x_{\lambda}$ and the state vertex $x_{\lambda}$, with a weight $a_{\lambda\lambda}$, equal to the opposite of the static gain of the causal path.

14. Submatrix $B_{1}$: each causal path between an input variable $u_{i}$ and the derivative of an energy variable $x_{I}$ is represented on the digraph by an $A$-edge between the input vertex $u_{i}$ and the state vertex $x_{I}$, with a weight $b_{1i}$, equal to the static gain of the causal path.

15. Submatrix $B_{D}$: each causal path between an input variable $u_{i}$ and an energy variable $x_{D}$ is represented on the digraph by an $A$-edge between the input vertex $u_{i}$ and the state vertex $x_{D}$, with a weight $b_{Di}$, equal to the opposite of the static gain of the causal path.

16. Submatrix $B_{\lambda}$: each causal path between an input variable $u_{i}$ and a power variable $x_{\lambda}$ is represented on the digraph by an $A$-edge between the input vertex $u_{i}$ and the state vertex $x_{\lambda}$, with a weight $b_{\lambda i}$, equal to the opposite of the static gain of the causal path.

17. Submatrix $C_{I}$: each causal path between an energy variable $x_{I}$ and an output variable $y_{j}$ is represented on the digraph by an $A$-edge between the state vertex $x_{I}$ and the output vertex $y_{j}$, with a weight $c_{Ij}$, equal to the static gain of the causal path.

18. Submatrix $C_{D}$: each causal path between the derivative of an energy variable $x_{D}$ and an output variable $y_{j}$ is represented on the digraph by an $E$-edge between the state vertex $x_{D}$ and the output vertex $y_{j}$, with a weight $c_{Dj}$, equal to the static gain of the causal path.

19. Submatrix $C_{\lambda}$: each causal path between a power variable $x_{\lambda}$ and an output variable $y_{j}$ is represented on the digraph by an $A$-edge between the state vertex $x_{\lambda}$ and the output vertex $y_{j}$, with a weight $c_{\lambda j}$, equal to the static gain of the causal path.

20. Submatrix $D$: each causal path between an input variable $u_{i}$ and an output variable $y_{j}$ is represented on the digraph by an $A$-edge between the input vertex $u_{i}$ and the output vertex $y_{j}$, with a weight $d_{ji}$, equal to the static gain of the causal path.
On the digraph representation of Figure 3 obtained from proposition 1, the following remarks can be highlighted:

- the edges of type $E$ are:
  - either loops on the energy variable $x_I$ vertices,
  - or the edges out-going from the energy variable $x_D$ vertices,
- the edge of type $A$ are all the other edges.

4.1.2 Example

We consider the model of Figure 4, specifying a flow source and an effort sensor. According to (13), the state-space representation is given in (15).

\[
\begin{bmatrix}
1 & 0 & 1 \\
0 & 1 & -1 \\
0 & 0 & 0 \\
\end{bmatrix}
\begin{bmatrix}
\dot{p}_{11} \\
\dot{p}_{12} \\
\dot{p}_D \\
\end{bmatrix}
=
\begin{bmatrix}
0 & 0 & 0 \\
0 & 0 & 0 \\
-I_d & I_d & 1 \\
\end{bmatrix}
\begin{bmatrix}
p_{11} \\
p_{12} \\
p_D \\
\end{bmatrix}
+
\begin{bmatrix}
0 \\
0 \\
-I_d \\
\end{bmatrix}
u
\]

\[
y = \begin{bmatrix}
0 & 0 & \frac{d(\cdot)}{dt} \\
\end{bmatrix}
\begin{bmatrix}
p_{11} \\
p_{12} \\
p_D \\
\end{bmatrix}
+ \begin{bmatrix}
0 \\
\end{bmatrix}
u \quad (15)
\]

The corresponding digraph obtained with proposition 1 is given in Figure 5.
If we augment the state vector in (15) with the power variable $\lambda$ specified in Figure 4 and according to (13), the equation system can be written as (16), equivalent to (15).

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \dot{p}_{I_1} \\ \dot{p}_{I_2} \\ \dot{p}_D \\ \dot{\lambda} \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 0 & 1 \\ -I_d & I_d & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} p_{I_1} \\ p_{I_2} \\ p_D \\ \lambda \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ -I_d \\ 0 \end{bmatrix} u$$

$$y = \begin{bmatrix} 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} p_{I_1} \\ p_{I_2} \\ p_D \\ \lambda \end{bmatrix} + [0] u \quad (16)$$

The corresponding digraph obtained with proposition 1 is given in Figure 6.

![Figure 6: Corresponding digraph model of Figure 4 bond graph with the additional state variable $\lambda$](image)

**Remark:** for the sake of clarity, Figure 6 digraph shows the $E$-edges with the Laplace operator in their respective weights, to clearly highlight these edges.

### 4.1.3 Consideration of zero-order causal cycles

The digraph corresponding to a bond graph model requires the determination of the static gains of the causal paths, which potentially can be generalized $^3$, depending on the state variables chosen in (13) (i.e. with or without power variables $x_\lambda$ in addition of the energy variables $x_I$ and $x_D$). First, the presence of all the energy variables (i.e. dependent and independent) in the state vector means that all 1ZCP and 3ZCP involving $I$ and/or $C$ elements are not solved a priori, and are "cut" by the fact of having to explicitly express one of the variables belonging to each of these causal loops. Secondly, the absence of power variables $x_\lambda$ in the state vector requires to solve all the "other ZCP", i.e. 2ZCP, 3ZCP involving only $R$ elements, and also 4ZCP. The case of 2ZCP and 3ZCP involving only $R$ elements is detailed in [31, 32], noticing that structurally in (9),

$$(I - LS_{33})^{-1} = \sum_{l=0}^{\infty} (LS_{33})^l \quad (17)$$

This corresponds to a causal loop that is browsed an infinite number of times, without modification of its order. It is also worth noticing that the form (17) leads to the solvability condition of Rosenberg & André [33], to ensure the convergence of the series.

The important idea of augmenting the state vector with power variables is the potential gain of simplicity when inspecting the bond graph by reducing the number of zero-order causal cycles to be considered to build the state-space formulation (13) as well as the corresponding digraph (proposition 1).

### 4.2 Digraph/bond matching between digraph cycles and bond graph cycles

By definition and construction with the proposition 1, an edge on the digraph corresponds on the bond graph to a causal path between two given variables of the state vector (or their derivatives with respect to time), without any other chosen state variable occurring in this path. Given the fact that a digraph path is an oriented sequence of edges such that the initial vertex of each succeeding edge is the final vertex of the preceding edge [10], the corresponding notion in bond graph is a causal path going through the different variables associated with the vertices involved in the digraph oriented path. A digraph cycle is a specific oriented closed path in which the initial vertex and the final vertex are the same and in which no vertex is reached more than once, except the initial-final vertex. In bond graph, considering the variables of the state vector, the digraph cycle corresponds to a causal cycle in which a variable associated with a digraph vertex is reached only once. This correspondence is illustrated in Figure 7 on example page 7 from corresponding digraph of Figure 5.
5 Conclusion

For the class of LTI descriptor systems (1), we have proposed in this paper an extension of the junction structure matrix (section 2.2), adding any power variable of interest, and leading to the proposition of an extended symbolic determination of the descriptor state space system from a (bi)causal bond graph without preferred causality assignment, with potentially zero-order causal paths of type 1ZCP, 2ZCP, 3ZCP, 4ZCP, and R and/or I/C fields (section 3). An important result on the regularity of the matrix pencil $sE - A$ for this class of systems has been highlighted. A bond graph/digraph matching has then been proposed (section 4), with a dedicated digraph construction procedure (proposition 1). This work could be seen as the core material to adapt the existing digraph procedure [10, 19] for the graphical determination of determinants on the bond graph model [22]. Before this, requires a completion of the bond graph/digraph matching introduced in section 4.2 by introducing, first, the notion of supplemented digraph, involving feedback edges from each output to each input [19]. Previous considerations on cycles correspondence can naturally be generalized to the input variables $u_i$ and output variables $y_i$ of the system (i.e., on one hand, the related vertex in the digraph, and on the other hand, source and sensor elements in the bond graph). The digraph/bond graph matching can then be established between the supplemented digraph cycles involving the input and output vertices, and the input/output causal paths on the corresponding bond graph, and then completed by the correspondence of cycle family on the supplemented digraph and bond graph family as defined in [21].

References


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