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# 1 Index Reduction of Second Order, 2 Discrete Time Descriptor Systems \*

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5 Version 2 : 05/08  
6 Received: date / Accepted: date

7 **Abstract** This paper is devoted to the analysis of linear, second order *discrete*  
8 *time descriptor systems* (or singular difference equations (SiDEs) with control).  
9 Following the algebraic approach proposed in [10, 11], first we present a theo-  
10 retical framework to analyze the corresponding initial value problem for SiDEs,  
11 which is followed by the analysis of descriptor systems. We also describe numer-  
12 ical methods to analyze structural properties related to the solvability analysis  
13 of these systems. This work extends and completes the researches in [2, 14, 18].  
14 **Keywords:** Singular systems; Difference equation; Descriptor systems;  
15 Strangeness-index; Regularization; Feedback.  
16 **AMS Subject Classification:** 34A09, 34A12, 65L05, 65H10

## 17 1 Introduction and Preliminaries

In this paper we study second order, discrete time descriptor systems of the form

$$A_n x(n+2) + B_n x(n+1) + C_n x(n) + D_n u(n) = f(n), \quad \text{for all } n \geq n_0. \quad (1.1)$$

We will also discuss the initial value problem of the associated singular difference equation (SiDE)

$$A_n x(n+2) + B_n x(n+1) + C_n x(n) = f(n), \quad \text{for all } n \geq n_0, \quad (1.2)$$

together with some given initial conditions

$$x(n_0+1) = x_1, \quad x(n_0) = x_0. \quad (1.3)$$

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The second author was supported the National Foundation for Science and Technology Development (NAFOSTED) under the project number 101.01-2017.302. He also would like to thank the Vietnam Institute for Advanced Study in Mathematics (VIASM) for their kind hospitality during his research visit.

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18 Here the solution/state  $x = \{x(n)\}_{n \geq n_0}$ , the inhomogeneity  $f = \{f(n)\}_{n \geq n_0}$ ,  
 19 the input function  $u = \{u(n)\}_{n \geq n_0}$ , where  $x(n) \in \mathbb{R}^d$ ,  $f(n) \in \mathbb{R}^m$  and  $u(n) \in \mathbb{R}^p$   
 20 for each  $n \geq n_0$ . The coefficients contain three matrix sequences  $\{A_n\}_{n \geq n_0}$ ,  
 21  $\{B_n\}_{n \geq n_0}$ ,  $\{C_n\}_{n \geq n_0}$  which always take values in  $\mathbb{R}^{m,d}$ , and  $\{D_n\}_{n \geq n_0}$  which  
 22 take values in  $\mathbb{R}^{m,p}$ . We notice, that all the results in this paper also carry  
 23 over to the complex case, and they can also be easily extended to systems of  
 24 higher than second order, but for ease of notation and because this is the most  
 25 important case in practice, we restrict ourselves to the real, second order case.

26 The SiDE (1.2), on one side, can be consider as the resulting equation,  
 27 obtained by finite difference or discretization of some continuous-time DAEs or  
 28 constrained PDEs. One the other side, there are also many models/applications  
 29 in real-life, which lead to SiDEs, for example Leotief economic models, backward  
 30 Leslie model in biology, etc, see e.g. [1, 5, 9, 15].

31 While both DAEs and SiDEs of first order have been well-studied from both  
 32 theoretical and numerical sides, the same maturity has not been reached for  
 33 higher order systems. In classical literature [1, 5, 9], usually new variables are  
 34 introduced to present some chosen derivatives of the state variable  $x$  such that  
 35 a high order system can be reformulated as a first order one. This method,  
 36 however, is not only non-unique but also has presented some substantial dis-  
 37 advantages. As have been fully discussed in [14, 18] for continuous time sys-  
 38 tems, these disadvantages include: (1st) increase the index of the system, and  
 39 therefore the complexity of a numerical method to solve it; (2nd) increase the  
 40 computational effort, due to the bigger size of a new system; (3rd) affect the  
 41 controllability/observability of the corresponding descriptor system, since there  
 42 exist situations where a new system is uncontrollable while the original one is.  
 43 Therefore, the *algebraic approach*, which treats the system directly without re-  
 44 formulating it, has been presented in [14, 18, 22, 23] in order to overcome the  
 45 disadvantages mentioned above. Nevertheless, even for second order SiDEs, this  
 46 method has not yet been considered.

47 Another motivation of this work comes from recent researches on the stability  
 48 analysis of high order, discrete time systems with time-dependent coefficients  
 49 [13, 19]. There, considered systems are in either strangeness-free form or linear  
 50 state-space form. Nevertheless, it is not always the case in applications, and  
 51 hence, a reformulation procedure is necessary.

52 Therefore, the main aim of this article is to set up a comparable framework  
 53 for second order SiDEs/descriptor systems. It is worth marking that the alge-  
 54 braic method proposed in [14, 18] is applicable theoretically but not numerically,  
 55 due to two reasons: (1) The condensed form of the matrix coefficients are really  
 56 big and complicated. (2) The system's transformations are not orthogonal, and  
 57 hence, not numerically stable. In this work, we will modify this method to make  
 58 it more concise and also be computable in a stable way.

59 The outline of this paper is as follows. After recalling some preliminary con-  
 60 cepts and some auxiliary lemmata, in Sections 2 and 3 we consecutively intro-  
 61 duce *index reduction procedures* for SiDEs and for descriptor systems. Resulting  
 62 systems from these procedures allow us to determine structural properties such  
 63 as existence and uniqueness of a solution, consistency and hidden constraints,  
 64 etc. For the numerical solution of these systems, in Section 4 we study the *differ-*  
 65 *ence array approach* in order to bring the original system to its strangeness-free  
 66 form. Finally, we finish with some conclusion.

68 In the following example we demonstrate some difficulties that may arise in  
69 the analysis of second order SiDEs.

*Example 1* Consider the following second order SiDE, motivated from Example 2, [18].

$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x(n+2) + \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x(n+1) + \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} x(n) = \begin{bmatrix} f_1(n) \\ f_2(n) \end{bmatrix}, \quad n \geq n_0. \quad (1.4)$$

Clearly, from the second equation  $[1 \ 0]x(n) = f_2(n)$ , we can shift forward the time  $n$  to obtain

$$[1 \ 0]x(n+1) = f_2(n+1) \quad \text{and} \quad [1 \ 0]x(n+2) = f_2(n+2).$$

Inserting these into the first equation of (1.4), we find out the hidden constraint  $f_2(n+2) + f_2(n+1) + [0 \ 1]x(n) = f_1(n)$ . Consequently, we obtain the following system, which possess a unique solution

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} x(n) = \begin{bmatrix} f_1(n) - f_2(n+2) - f_2(n+1) \\ f_2(n) \end{bmatrix}, \quad n \geq n_0.$$

70 Let  $n = n_0$  in this new system, we obtain a constraint that  $x(n_0)$  must obey. This  
71 example showed us some important facts. Firstly, one can use some shift operators  
72 and row-manipulation (Gaussian eliminations) to derive hidden constraints.  
73 Secondly, a solution only exists if an initial condition fulfills some consistency  
74 conditions.

For matrices  $Q \in \mathbb{R}^{q,d}$ ,  $P \in \mathbb{R}^{p,d}$ , the pair  $(Q, P)$  is said to *have no hidden redundancy* if

$$\text{rank} \left( \begin{bmatrix} Q \\ P \end{bmatrix} \right) = \text{rank}(Q) + \text{rank}(P).$$

75 Otherwise,  $(Q, P)$  is said to *have hidden redundancy*. The geometrical meaning  
76 of this concept is that the intersection space  $\text{span}(P^T) \cap \text{span}(Q^T)$  contains  
77 only the zero-vector  $\mathbf{0}$ . Here by  $\text{span}(P^T)$  (resp.,  $\text{span}(Q^T)$ ) we denote the real  
78 vector space spanned by the rows of  $P$  (resp., rows of  $Q$ ). We further notice  
79 that, if  $\begin{bmatrix} Q \\ P \end{bmatrix}$  is of full row rank then obviously, the pair  $(Q, P)$  has no hidden

80 redundancy. However, the converse is not true as is obvious for  $Q = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ ,

$$81 \quad P = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}.$$

82 **Lemma 1** ([7]) Suppose that for  $Q \in \mathbb{R}^{q,d}$ ,  $P \in \mathbb{R}^{p,d}$ , the pair  $(Q, P)$  has no  
83 hidden redundancy. Then, for any matrix  $U \in C^{q,q}$  and any  $V \in C^{p,p}$ , the pair  
84  $(UQ, VP)$  has no hidden redundancy.

85 **Lemma 2** ([7]) Consider  $k+1$  full row rank matrices  $R_0 \in \mathbb{R}^{r_0,d}, \dots, R_k \in$

86  $\mathbb{R}^{r_k,d}$ , and assume that for  $j = k, \dots, 1$  none of the matrix pairs  $\left( R_j, \begin{bmatrix} R_{j-1} \\ \vdots \\ R_0 \end{bmatrix} \right)$

87 has a hidden redundancy. Then,  $\begin{bmatrix} R_k \\ \vdots \\ R_0 \end{bmatrix}$  has full row rank.

88 Lemma 3 below will be very useful later for our analysis, in order to remove  
89 hidden redundancy in the coefficients of (1.2).

**Lemma 3** Consider two matrix sequences  $\{P_n\}_{n \geq n_0}$ ,  $\{Q_n\}_{n \geq n_0}$  which take values in  $\mathbb{R}^{m,d}$ , and assume that they satisfy the constant rank assumptions

$$\text{rank}(Q_n) = r_Q, \quad \text{and} \quad \text{rank} \left( \begin{bmatrix} P_n \\ Q_n \end{bmatrix} \right) = r_{[P;Q]}, \quad \text{for all } n \geq n_0 .$$

90 Then, there exists a matrix sequence  $\left\{ \begin{bmatrix} S_n & 0 \\ Z_n^{(1)} & Z_n^{(2)} \end{bmatrix} \right\}_{n \geq n_0}$  in  $\mathbb{R}^{p,p+q}$  such that  
91 the following conditions hold.

92 i)  $S_n \in \mathbb{R}^{r_{[P;Q]} - r_Q, p}$ ,  $Z_n^{(1)} \in \mathbb{R}^{p - r_{[P;Q]} + r_Q, p}$ ,  $Z_n^{(2)} \in \mathbb{R}^{p - r_{[P;Q]} + r_Q, q}$ ,

93 ii)  $\begin{bmatrix} S_n \\ Z_n^{(1)} \end{bmatrix} \in \mathbb{R}^{p,p}$  is orthogonal, and  $Z_n^{(1)} P_n + Z_n^{(2)} Q_n = 0$ ,

94 iii) the matrix  $S_n P_n$  has full row rank, and the pair  $(S_n P_n, Q_n)$  has no hidden  
95 redundancy.

*Proof.* First using SVD we factorize  $Q_n$  and then partition  $P_n$  conformably to get

$$U_1^T Q_n V_1 = \begin{bmatrix} \Sigma_n & 0 \\ 0 & 0 \end{bmatrix}, \quad \text{and} \quad P_n V_1 = [P_{n,1} \ P_{n,2}], \quad (1.5)$$

where the matrices  $U_1 = [U_{11} \ U_{12}] \in \mathbb{R}^{q,q}$ ,  $V_1 = [V_{11} \ V_{12}] \in \mathbb{R}^{d,d}$  are orthogonal and  $\Sigma_n \in \mathbb{R}^{r_Q, r_Q}$  is diagonal. Now we use a second SVD to factorize  $P_{n,2}$  and to find an orthogonal matrix  $U_2^T = \begin{bmatrix} S \\ Z_n^{(1)} \end{bmatrix} \in \mathbb{R}^{p,p}$  such that  $U_2^T P_{n,2} = \begin{bmatrix} P_{n,12} \\ 0 \end{bmatrix}$ , where  $P_{n,12}$  has full row rank. Thus, we obtain

$$\begin{bmatrix} S_n & 0 \\ Z_n^{(1)} & 0 \\ 0 & U_{11}^T \\ 0 & U_{12}^T \end{bmatrix} \begin{bmatrix} P_n \\ Q_n \end{bmatrix} [V_{11} \ V_{12}] = \begin{bmatrix} P_{n,11} & P_{n,12} \\ P_{n,21} & 0 \\ \Sigma_n & 0 \\ 0 & 0 \end{bmatrix} \begin{matrix} r_{[P;Q]} - r_Q \\ p - r_{[P;Q]} + r_Q \\ r_Q \\ q - r_Q \end{matrix} .$$

Since  $P_{n,12}$  has full row rank,  $S_n P_n = [P_{n,11} \ P_{n,12}] V_1^{-1}$  also has full row rank. Moreover, one sees that

$$\text{rank} \left( \begin{bmatrix} S_n P_n \\ Q_n \end{bmatrix} \right) = \text{rank} ([0 \ P_{n,12}]) + \text{rank} ([\Sigma_n \ 0]) = \text{rank}(S_n P_n) + \text{rank}(Q_n),$$

which follows that the pair  $(S_n P_n, Q_n)$  has no hidden redundancy.

Finally, setting  $Z_n^{(2)} := -P_{n,21} \Sigma_n^{-1} U_{11}^T$ , we obtain

$$Z_n^{(1)} P_n + Z_n^{(2)} Q_n = ([P_{n,21} \ 0] - P_{n,21} \Sigma_n^{-1} [\Sigma_n \ 0]) V_1^{-1} = 0,$$

96 which completes the proof.  $\square$

97 *Remark 1* i) In the special case, where  $P_n$  has full row rank and the pair  $(P_n, Q_n)$   
 98 has no hidden redundancy, we will adapt the notation of an empty matrix and  
 99 take  $S_n = I_p$ ,  $Z_n^{(1)} = [ ]^{0,p}$ ,  $Z_n^{(2)} = [ ]^{0,q}$ .  
 100 ii) Furthermore, we notice that the matrices  $U_1, U_2, V_1$  in the proof of Lemma  
 101 3 are orthogonal. Therefore, in case that the singular values of  $Q_n$  are neither  
 102 too small nor too big, then  $\Sigma_n^{-1}$  is well-conditioned, and hence we can stably  
 103 compute the matrix  $Z_n^{(2)}$ . Both matrices  $Z_n^{(1)}$  and  $Z_n^{(2)}$  will play the key role in  
 104 our *index reduction procedure* presented in the next section.

105 For any given matrix  $M$ , by  $M^T$  we denote its transpose. By  $T_0(M)$  we  
 106 denote an orthogonal matrix whose columns span the left null space of  $M$ . By  
 107  $T_\perp(M)$  we denote an orthogonal matrix whose columns span the vector space  
 108  $\text{range}(M)$ . From basic linear algebra, we have the following three lemmata.

**Lemma 4** *The matrix  $\begin{bmatrix} T_\perp^T(M) \\ T_0^T(M) \end{bmatrix}$  is nonsingular, the matrix  $T_\perp^T(M) M$  has full row rank, and the following identity holds*

$$\begin{bmatrix} T_\perp^T(M) \\ T_0^T(M) \end{bmatrix} M = \begin{bmatrix} T_\perp^T(M) M \\ 0 \end{bmatrix}.$$

109 *Proof.* A simple proof can be found, for example, in [6].  $\square$

**Lemma 5** *Given four matrices  $\check{A}, \check{B}, \check{C}$  in  $\mathbb{R}^{m,d}$  and  $\check{D}$  in  $\mathbb{R}^{m,p}$ . Let us consider the following matrices whose columns span orthogonal bases of the associated vector spaces*

$$\begin{aligned} T_1 & \text{ basis of kernel}(\check{A}^T), & \text{and } T_{1,\perp} & \text{ basis of range}(\check{A}), \\ W_1 & \text{ basis of kernel}(T_1^T \check{D})^T, & \text{and } W_{1,\perp} & \text{ basis of range}(T_1^T \check{D}), \\ & & J_D & := W_{1,\perp}^T T_1^T \check{D}, \\ J_{B_1} & := W_1^T T_1^T \check{B}, & \text{and } J_{B_2} & := W_{1,\perp}^T T_{1,\perp}^T \check{B}, \\ J_{C_1} & := W_1^T T_1^T \check{C}, & \text{and } J_{C_2} & := W_{1,\perp}^T T_{1,\perp}^T \check{C}, \\ T_2 & \text{ basis of kernel}(J_{B_1}^T), & \text{and } T_{2,\perp} & \text{ basis of range}(J_{B_1}), \\ T_3 & \text{ basis of kernel}(J_{B_2}^T), & \text{and } T_{3,\perp} & \text{ basis of range}(J_{B_2}), \\ T_4 & \text{ basis of kernel}(T_2^T J_{C_1})^T, & \text{and } T_{4,\perp} & \text{ basis of range}(T_2^T J_{C_1}). \end{aligned}$$

110 *Then, the following assertions hold true.*

111 i) *The matrices  $\begin{bmatrix} T_{i,\perp} \\ T_i \end{bmatrix}$ ,  $i = 1, \dots, 4$ ,  $\begin{bmatrix} W_{1,\perp} \\ W_1 \end{bmatrix}$  are orthogonal.*

112 ii) *The matrices  $T_{1,\perp}^T \check{A}$ ,  $T_{2,\perp}^T J_{B_1}$ ,  $T_{3,\perp}^T J_{B_2}$ ,  $T_{4,\perp}^T T_2^T J_{C_1}$ , and  $J_D$  have full row rank.*

iii) *Moreover, there exists a nonsingular matrix  $\check{U}$  such that*

$$\check{U} \left[ \check{A} \ \check{B} \ \check{C} \mid \check{D} \right] = \left[ \begin{array}{ccc|c} \check{A}_1 & \check{B}_1 & \check{C}_1 & \check{D}_1 \\ 0 & \check{B}_2 & \check{C}_2 & 0 \\ 0 & 0 & \check{C}_3 & 0 \\ 0 & 0 & 0 & 0 \\ \hline 0 & \check{B}_4 & \check{C}_4 & \check{D}_4 \\ 0 & 0 & \check{C}_5 & \check{D}_5 \end{array} \right], \quad (1.6)$$

113 where the matrices  $\check{A}_1, \check{B}_2, \check{B}_4, \check{C}_3, \begin{bmatrix} \check{D}_4 \\ \check{D}_5 \end{bmatrix}$  have full row rank.

*Proof.* The first two claims followed directly from Lemma 4. To prove the third claim, we construct the desired matrix  $\check{U}$  as follows

$$\check{U} := \begin{bmatrix} I & & & \\ & I & & \\ & & T_{4,\perp}^T & \\ & & T_4^T & \\ & & & I \end{bmatrix} \cdot \begin{bmatrix} I & & & \\ & T_{2,\perp}^T & & \\ & T_2^T & & \\ & & & T_{3,\perp}^T \\ & & & T_3^T \end{bmatrix} \cdot \begin{bmatrix} I & \\ & W_{1,\perp}^T \end{bmatrix} \cdot \begin{bmatrix} T_{1,\perp}^T \\ T_1^T \end{bmatrix}.$$

Thus, we have that

$$\check{U} \begin{bmatrix} \check{A} & \check{B} & \check{C} & \check{D} \end{bmatrix} = \begin{bmatrix} T_{1,\perp}^T \check{A} & T_{1,\perp}^T \check{B} & T_{1,\perp}^T \check{C} & T_{1,\perp}^T \check{D} \\ 0 & T_{2,\perp}^T J_{B1} & T_{2,\perp}^T J_{C1} & 0 \\ 0 & 0 & T_{4,\perp}^T T_2^T J_{C1} & 0 \\ 0 & 0 & 0 & 0 \\ \hline 0 & T_{3,\perp}^T J_{B2} & T_{3,\perp}^T J_{C2} & T_{3,\perp}^T J_D \\ 0 & 0 & T_3^T J_{C2} & T_3^T J_D \end{bmatrix}.$$

114 Due to the parts i) and ii), we see that this is exactly the desired form (1.6).  $\square$

115 **Lemma 6** Let  $P \in \mathbb{R}^{p,d}$ ,  $Q \in \mathbb{R}^{q,d}$  be two full row rank matrices and  $p + q \leq d$ .  
116 Then, the following assertions hold true.

117 i) There exists a matrix  $F \in \mathbb{R}^{d,d}$  such that  $H := \begin{bmatrix} P \\ QF \end{bmatrix}$  has full row rank.

118 ii) For any  $G \in \mathbb{R}^{q,d}$ , there exists a matrix  $F \in \mathbb{R}^{d,d}$  such that  $\begin{bmatrix} P \\ G + QF \end{bmatrix}$  has  
119 full row rank.

*Proof.* i) First we consider the SVDs of  $P$  and  $G$  that reads

$$U_P P V_P = [\Sigma_P \ 0_{p,d-p}], \quad U_Q Q V_Q = [\Sigma_Q \ 0_{q,d-q}],$$

where  $\Sigma_P, \Sigma_Q$  are nonsingular, diagonal matrices, and  $0_{p,d-p}$  (resp.  $0_{q,d-q}$ ) are the zero matrix of size  $p$  by  $d-p$  (resp.  $q$  by  $d-q$ ).

By choosing  $F := V_Q \begin{bmatrix} 0 & I_q \\ I_{d-q} & 0 \end{bmatrix} V_P^{-1}$  we see that

$$\begin{bmatrix} U_P & 0 \\ 0 & U_Q \end{bmatrix} \begin{bmatrix} P \\ QF \end{bmatrix} V_P = \begin{bmatrix} U_P P V_P \\ U_Q Q F V_P \end{bmatrix} = \begin{bmatrix} \Sigma_P & 0_{p,d-p-q} & 0_{p,q} \\ 0_{q,p} & 0_{p,d-p-q} & \Sigma_Q \end{bmatrix},$$

and hence, the claim i) is proven.

ii) Clearly, in case that the matrix  $F$  is very big, then  $G$  is only a small perturbation, and hence for sufficiently large  $\eta$ , by choosing

$$F := \eta V_Q \begin{bmatrix} 0 & I_q \\ I_{d-q} & 0 \end{bmatrix} V_P^{-1},$$

120 we obtain the full row rank property of  $\begin{bmatrix} P \\ G + QF \end{bmatrix}$ .  $\square$

121 *Remark 2* It should be noted that, the proof of Lemmata 5 and 6 are construc-  
 122 tive, and all the matrices  $T_{i,\perp}$ ,  $T_i$ ,  $i = 1, \dots, 4$ ,  $W_{1,\perp}$ ,  $W_1$  and  $F$  can be stably  
 123 computed.

## 124 2 Strangeness-index of second-order SiDEs

125 In this section, we study the solvability analysis of the second-order SiDE (1.2)  
 126 and of its corresponding IVP (1.2)–(1.3). Many regularization procedures and  
 127 their associated index concepts have been proposed for first order systems, see  
 128 the survey [17] and the references therein. Nevertheless, for second order sys-  
 129 tems, only the strangeness-index has been proposed for only continuous but  
 130 not discrete time systems in [18, 23]. Thus, it is our purpose to construct a  
 131 comparable regularization and index concept for system (1.2).

Let

$$M_n := [A_n \ B_n \ C_n], \quad X(n) := \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix},$$

we call  $\{M_n\}_{n \geq n_0}$  the *behavior matrix sequence* of system (1.2). Thus, (1.2) can be rewritten as

$$M_n X(n) = f(n), \quad \text{for all } n \geq n_0. \quad (2.1)$$

Clearly, by scaling (1.2) with a pointwise nonsingular matrix sequence  $\{P_n\}_{n \geq n_0}$  in  $\mathbb{R}^{d,d}$ , we obtain a new system

$$[P_n A_n \ P_n B_n \ P_n C_n] X(n) = P_n f(n), \quad \text{for all } n \geq n_0, \quad (2.2)$$

132 without changing the solution space. This motivates the following definition.

133 **Definition 1** Two behavior matrix sequences  $\{M_n = [A_n \ B_n \ C_n]\}_{n \geq n_0}$  and  
 134  $\{\tilde{M}_n = [\tilde{A}_n \ \tilde{B}_n \ \tilde{C}_n]\}_{n \geq n_0}$  are called (*strongly*) *left equivalent* if there exists a  
 135 pointwise nonsingular matrix sequence  $\{P_n\}_{n \geq n_0}$  such that  $\tilde{M}_n = P_n M_n$  for all  
 136  $n \geq n_0$ . We denote this equivalence by  $\{M_n\}_{n \geq n_0} \stackrel{\ell}{\sim} \{\tilde{M}_n\}_{n \geq n_0}$ . If this is the  
 137 case, we also say that two SiDEs (1.2), (2.2) are left equivalent.

**Lemma 7** Consider the behavior matrix sequence  $\{M_n\}_{n \geq n_0}$  of system (1.2). Then, for all  $n \geq n_0$ , we have that

$$\{M_n\}_{n \geq n_0} \stackrel{\ell}{\sim} \left\{ \begin{bmatrix} A_{n,1} & B_{n,1} & C_{n,1} \\ 0 & B_{n,2} & C_{n,2} \\ 0 & 0 & C_{n,3} \\ 0 & 0 & 0 \end{bmatrix} \right\}_{n \geq n_0}, \quad \begin{matrix} r_{2,n} \\ r_{1,n} \\ r_{0,n} \\ v_n \end{matrix} \quad (2.3)$$

138 where the matrices  $A_{n,1}$ ,  $B_{n,2}$ ,  $C_{n,3}$  on the main diagonal have full row rank.  
 139 Here the numbers  $r_{2,n}$ ,  $r_{1,n}$ ,  $r_{0,n}$ ,  $v_n$  are row-sizes of the block rows of  $M_n$ .  
 140 Furthermore, these numbers are invariant under left equivalent transformations.  
 141 Thus, we can call them the local characteristic invariants of the SiDE (1.2).

*Proof.* The block diagonal form (2.3) is obtained directly by consecutively compressing the block columns  $A_n$ ,  $B_n$ ,  $C_n$  of  $M_n$  via Lemma 4. In details, we have

that

rows of  $A_{n,1}$  form the basis of the space  $\text{range}(A_n^T)$ ,  
rows of  $B_{n,2}$  form the basis of the space  $\text{range}(T_0^T(A_n) B_n)^T$ ,  
rows of  $C_{n,3}$  form the basis of the space  $\text{range}\left(T_0^T\left(\begin{bmatrix} A_n \\ B_n \end{bmatrix}\right) C_n\right)^T$ .

142 Moreover, from (2.3), we obtain the following identities

$$\begin{aligned} r_{2,n} &= \text{rank}(A_n), \\ r_{1,n} &= \text{rank}([A_n B_n]) - \text{rank}(A_n), \\ r_{0,n} &= \text{rank}([A_n B_n C_n]) - \text{rank}([A_n B_n]), \end{aligned}$$

143 which proves the second claim.  $\square$

144 Analogous to the continuous-time case, we will apply an *algebraic approach*  
145 (see [2, 18]), which aims to reformulate (1.2) into a so-called *strangeness-free*  
146 form, as stated in the following definition.

**Definition 2** ([13]) System (1.2) is called *strangeness-free* if there exists a pointwise nonsingular matrix sequence  $\{P_n\}_{n \geq n_0}$  such that by scaling the SiDE (1.2) at each point  $n$  with  $P_n$ , we obtain a new system of the form

$$\begin{aligned} \hat{r}_2 & \begin{bmatrix} \hat{A}_{n,1} \\ 0 \\ 0 \\ 0 \end{bmatrix} x(n+2) + \begin{bmatrix} \hat{B}_{n,1} \\ \hat{B}_{n,2} \\ 0 \\ 0 \end{bmatrix} x(n+1) + \begin{bmatrix} \hat{C}_{n,1} \\ \hat{C}_{n,2} \\ \hat{C}_{n,3} \\ 0 \end{bmatrix} x(n) = \begin{bmatrix} \hat{f}_1(n) \\ \hat{f}_2(n) \\ \hat{f}_3(n) \\ \hat{f}_4(n) \end{bmatrix}, \text{ for all } n \geq n_0, \\ \hat{r}_1 & \\ \hat{r}_0 & \\ \hat{v} & \end{aligned} \tag{2.4}$$

147 where the matrix  $\begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix}$  has full row rank for all  $n \geq n_0$ .

148 *Remark 3* We notice that, if the SiDE (1.2) is of the strangeness-free form (2.4),  
149 then the existence and uniqueness of the solution  $\{x(n)\}_{n \geq n_0}$  can be achieved  
150 if and only if  $\hat{r}_2 + \hat{r}_1 + \hat{r}_0 = d$ . Furthermore, either the last block row equation  
151 of (2.4) do not appear, i.e.  $\hat{v} = 0$ , or  $\hat{f}_4(n) = 0$  for all  $n \geq n_0$ .

152 In order to perform an algebraic approach, an additional assumption below  
153 is usually needed.

154 **Assumption 1.** Assume that the local characteristic invariants  $r_{2,n}$ ,  $r_{1,n}$ ,  $r_{0,n}$   
155 become global, i.e., they are constant for all  $n \geq n_0$ . Furthermore, assume that

156 two matrix sequences  $\left\{ \begin{bmatrix} A_{n,1} \\ B_{n,2} \\ C_{n,3} \end{bmatrix} \right\}_{n \geq n_0}$  and  $\left\{ \begin{bmatrix} B_{n,2} \\ C_{n,3} \end{bmatrix} \right\}_{n \geq n_0}$  have constant rank

157 for all  $n \geq n_0$ .

*Remark 4* Following directly from the proof of Lemma 7, we see that Assump-  
tion 1 is satisfied if and only if five following constant rank conditions are satisfied

$$\begin{aligned} \text{rank}(A_n) &\equiv \text{const.}, \quad \text{rank}([A_n B_n]) \equiv \text{const.}, \quad \text{rank}([A_n B_n C_n]) \equiv \text{const.}, \\ \text{rank}(T_0^T(A_n) B_n) &\equiv \text{const.}, \quad \text{rank}\left(T_0^T\left(\begin{bmatrix} A_n \\ B_n \end{bmatrix}\right) C_n\right) \equiv \text{const.} \end{aligned} \tag{2.5}$$



158 *Remark 5* In the context of continuous-time systems, the quantities  $r_2$ ,  $r_1$ , and  
 159  $r_0$  are the dimensions of the second order derivative part, the first order deriva-  
 160 tive part, and the algebraic part, respectively. Furthermore,  $r_2 + r_1$  is exactly  
 161 the degree of freedoms of the considered system.

Let us call the number

$$r_u := 3r_2 + 2r_1 + r_0$$

the *upper rank* of system (1.2). Clearly,  $r_u$  is invariant under left equivalence transformations. Rewrite (2.1) block row-wise, we obtain the following system for all  $n \geq n_0$ .

$$A_{n,1}x(n+2) + B_{n,1}x(n+1) + C_{n,1}x(n) = f_1(n), \quad r_2 \text{ equations}, \quad (2.6a)$$

$$B_{n,2}x(n+1) + C_{n,2}x(n) = f_2(n), \quad r_1 \text{ equations}, \quad (2.6b)$$

$$C_{n,3}x(n) = f_3(n), \quad r_0 \text{ equations}, \quad (2.6c)$$

$$0 = f_4(n), \quad v \text{ equations}. \quad (2.6d)$$

Since the matrices  $A_{n,1}$ ,  $B_{n,2}$ ,  $C_{n,3}$  have full row rank, the number of scalar difference equations of order 2 (resp. 1, and 0) in (1.2) is exactly  $r_2$  (resp.  $r_1$  and  $r_0$ ), while  $v$  is the number of redundant equations. Now we are able to define the shift-forward operator  $\Delta$ , which acts on some or whole equations of system (2.6). This operator maps each equation of system (2.6) at the time instant  $n$  to the equation itself at the time  $n+1$ , for example

$$\Delta : C_{n,3}x(n) = f_3(n) \mapsto C_{n+1,3}x(n+1) = f_3(n+1). \quad (2.7)$$

Clearly, under Assumption 1, this shift operator can be applied to equations of system (2.6). In order to reveal all hidden constraints of (2.6) we propose the idea, that for each  $j = 1, 2$ , we use equations of order less than  $j$  to reduce the number of scalar equations of order  $j$ . This task will be performed in Lemmata 9 and 10 below. In details, if the matrix pair  $(B_{n,2}, C_{n+1,3})$  has hidden redundancy then we will make use of the shifted equation (2.7). Analogously, if the pair  $\left( A_{n,1}, \begin{bmatrix} B_{n+1,2} \\ C_{n+2,3} \end{bmatrix} \right)$  has hidden redundancy then we will make use of the shifted equation

$$B_{n+1,2}x(n+2) + C_{n+1,2}x(n+1) = f_2(n+1), \quad (2.8)$$

and may be also the double shifted equation

$$C_{n+2,3}x(n+2) = f_3(n+2). \quad (2.9)$$

**Lemma 8** Consider the SiDE (1.2) and the equivalent system (2.6). Then, (1.2) has an identical solution set as the extended system

$$\begin{array}{l} r_2 \\ r_1 \\ r_0 \\ v \\ r_0 \\ r_1 \\ r_0 \end{array} \begin{array}{c} \left[ \begin{array}{ccc} A_{n,1} & B_{n,1} & C_{n,1} \\ 0 & B_{n,2} & C_{n,2} \\ 0 & 0 & C_{n,3} \\ 0 & 0 & 0 \\ 0 & C_{n+1,3} & 0 \\ B_{n+1,2} & C_{n+1,2} & 0 \\ C_{n+2,3} & 0 & 0 \end{array} \right] \begin{array}{c} x(n+2) \\ x(n+1) \\ x(n) \end{array} \end{array} = \begin{array}{c} \left[ \begin{array}{c} f_1(n) \\ f_2(n) \\ f_3(n) \\ f_4(n) \\ f_3(n+1) \\ f_2(n+1) \\ f_3(n+2) \end{array} \right], \quad (2.10)$$

162 for all  $n \geq n_0$ .

163 *Proof.* Since all equations in the lower part of (2.10) at any time point  $n$  is the  
 164 consequence of the upper part (which is exactly (2.6)) at the time instants  $n+1$   
 165 and  $n+2$ , the proof is directly followed.  $\square$

166 **Lemma 9** Consider the behavior matrix sequence  $\{M_n\}_{n \geq n_0}$  in (2.3). Assume  
 167 that Assumption 1 is satisfied. Then, there exist matrix sequences  $\{S_n^{(i)}\}_{n \geq n_0}$ ,  
 168  $i = 1, 2$ , and  $\{Z_n^{(j)}\}_{n \geq n_0}$ ,  $j = 1, \dots, 5$ , of appropriate sizes such that for all  
 169  $n \geq n_0$ , the following conditions hold true.

170 i) For  $i = 1, 2$ , the matrices  $\begin{bmatrix} S_n^{(i)} \\ Z_n^{(i)} \end{bmatrix} \in \mathbb{R}^{r_i, r_i}$  are orthogonal.

ii) The following identities hold true.

$$Z_n^{(1)} B_{n,2} + Z_n^{(3)} C_{n+1,3} = 0, \quad (2.11a)$$

$$Z_n^{(2)} A_{n,1} + Z_n^{(4)} B_{n+1,2} + Z_n^{(5)} C_{n+2,3} = 0. \quad (2.11b)$$

171 iii) Both matrix pairs  $\left( S_n^{(2)} A_n, \begin{bmatrix} B_{n+1,2} \\ C_{n+2,3} \end{bmatrix} \right)$ ,  $\left( S_n^{(1)} B_{n,2}, C_{n+1,3} \right)$  have no hidden  
 172 redundancy.

173 *Proof.* The proof can be directly obtained by applying Lemma 3 to two matrix  
 174 pairs  $(B_{n,2}, C_{n+1,3})$  and  $\left( A_{n,1}, \begin{bmatrix} B_{n+1,2} \\ C_{n+2,3} \end{bmatrix} \right)$ .  $\square$

**Lemma 10** Under the condition of Lemma 9, the SiDE (1.2) has exactly the same solution set as the transformed system

$$\begin{array}{c} d_2 \\ s_2 \\ d_1 \\ s_1 \\ r_0 \\ v \end{array} \begin{bmatrix} S_n^{(2)} A_{n,1} & S_n^{(2)} B_{n,1} & S_n^{(2)} C_{n,1} \\ 0 & Z_n^{(2)} B_{n,1} + Z_n^{(4)} C_{n+1,2} & Z_n^{(2)} C_{n,1} \\ 0 & S_n^{(1)} B_{n,2} & S_n^{(1)} C_{n,2} \\ 0 & 0 & Z_n^{(1)} C_{n,2} \\ 0 & 0 & C_{n,3} \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix} = \\ = \begin{bmatrix} S_n^{(2)} f_1(n) \\ Z_n^{(2)} f_1(n) + Z_n^{(4)} f_2(n+1) + Z_n^{(5)} f_3(n+2) \\ S_n^{(1)} f_2(n) \\ Z_n^{(1)} f_2(n) + Z_n^{(3)} f_3(n+1) \\ f_3(n) \\ f_4(n) \end{bmatrix}, \quad \text{for all } n \geq n_0. \quad (2.12)$$

175 Furthermore, both matrix pairs  $\left( S_n^{(2)} A_n, \begin{bmatrix} S_n^{(1)} B_{n+1,2} \\ C_{n+2,3} \end{bmatrix} \right)$ ,  $\left( S_n^{(1)} B_{n,2}, C_{n+1,3} \right)$  have  
 176 no hidden redundancy.

177 *Proof.* The proof is simple but quite long and technical, so we leave it to Ap-  
 178 pendix A.  $\square$

179 Consider system (2.12), we see that the upper rank of the behavior matrix  
180 is

$$\begin{aligned} r_u^{new} &\leq 3d_2 + 2(s_2 + d_1) + (s_1 + r_0) \\ &= 3(r_2 - s_2) + 2(s_2 + r_1 - s_1) + (s_1 + r_0) \\ &= r - (s_2 + s_1) \leq r. \end{aligned}$$

181 In conclusion, after performing a so-called *index reduction step*, which passes  
182 from (2.6) to (2.12), we have reduced the upper rank  $r_u$  at least by  $s_2 + s_1$ .  
183 Continue in this fashion until  $s_1 = s_2 = 0$ , we obtain the following algorithm.

---

**Algorithm 1** Index reduction steps for SiDEs at the time point  $n$

---

- 1: **Input:** The SiDE (1.2) and its behavior form (2.1). Set  $i = 0$ ,  $\mu = 0$ .
- 2: **Return:** A strangeness-free SiDE of the form (2.4).
- 3: Transform the behavior matrix  $[A_n \ B_n \ C_n]$  to the block upper triangular form

$$\tilde{M}_n := \begin{bmatrix} A_{n,1} & B_{n,1} & C_{n,1} \\ 0 & B_{n,2} & C_{n,2} \\ 0 & 0 & C_{n,3} \\ 0 & 0 & 0 \end{bmatrix},$$

where all the matrices  $A_{n,1}$ ,  $B_{n,2}$ ,  $C_{n,3}$  on the main diagonal have full row rank.

- 4: **if** both matrix pairs  $\left(A_{n,1}, \begin{bmatrix} B_{n+1,2} \\ C_{n+2,3} \end{bmatrix}\right)$  and  $(B_{n,2}, C_{n+1,3})$  have no hidden redundancy **then** STOP.
  - 5: **else** set  $i := i + 1$  and go to 6
  - 6: Find the matrices  $S_n^{(j)}$ ,  $j = 1, 2$ , and  $Z_n^{(j)}$ ,  $j = 1, \dots, 5$  as in Lemma 9.
  - 7: **if**  $Z_n^{(5)} \neq []$  **then** set  $\mu := \mu + 2$ .
  - 8: **else** set  $\mu := \mu + 1$
  - 9: **end if**
  - 10: **end if**
  - 11: Go back to 3.
- 

184 After each index reduction step the upper rank  $r_u^i$  has been decreased at  
185 least by  $s_2^i + s_1^i$ , so Algorithm 1 terminates after a finite number  $\mu$  of iterations,  
186 which will be called the *strangeness-index* of the SiDE (1.2).

**Theorem 2** Consider the SiDE (2.1) and assume that Assumption 1 is satisfied for any  $n$  and any  $i$  considered within the loop, such that the strangeness-index  $\mu$  is well-defined by Algorithm 1. Then, the SiDE (1.2) has the same solution set as the strangeness-free SiDE

$$\begin{bmatrix} r_2^\mu \\ r_1^\mu \\ r_0^\mu \\ v^\mu \end{bmatrix} \begin{bmatrix} \hat{A}_{n,1} & \hat{B}_{n,1} & \hat{C}_{n,1} \\ 0 & \hat{B}_{n,2} & \hat{C}_{n,2} \\ 0 & 0 & \hat{C}_{n,3} \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix} = \begin{bmatrix} \hat{g}_1(n) \\ \hat{g}_2(n) \\ \hat{g}_3(n) \\ \hat{g}_4(n) \end{bmatrix}, \text{ for all } n \geq n_0, \quad (2.13)$$

187 where the matrix  $\begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix}$  has full row rank for all  $n \geq n_0$ . Here  $\hat{g}_2$  and  $\hat{g}_3$   
188 are functions of  $f(n+1), \dots, f(n+\mu)$ .

189 *Proof.* The proof is a direct consequence of Algorithm 1, where the matrix  
 190  $\begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix}$  has full row rank due to Lemma 2.  $\square$

191 To illustrate Algorithm 1, we consider the following example.

*Example 2* Given a parameter  $\alpha \in \mathbb{R}$ , we consider the second order SiDE

$$\begin{bmatrix} 1 & n+1 & n+4 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} x(n+2) + \begin{bmatrix} 0 & \alpha & 2n+3 \\ 1 & n & 1 \\ 0 & 0 & 0 \end{bmatrix} x(n+1) + \begin{bmatrix} 0 & n+1 & 0 \\ 0 & 0 & n \\ 0 & 0 & n+1 \end{bmatrix} x(n) = \begin{bmatrix} f_1(n) \\ f_2(n) \\ f_3(n) \end{bmatrix}, \quad (2.14)$$

for all  $n \geq 0$ . Fortunately, the behavior matrix

$$M = \left[ \begin{array}{ccc|ccc|ccc} 1 & n+1 & n+4 & 0 & \alpha & 2n+3 & 0 & n+1 & 0 \\ 0 & 0 & 0 & 1 & n & 1 & 0 & 0 & n \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & n+1 \end{array} \right] = \begin{bmatrix} A_{n,1} & B_{n,1} & C_{n,1} \\ 0 & B_{n,2} & C_{n,2} \\ 0 & 0 & C_{n,3} \end{bmatrix}$$

is already in the block diagonal form, so we do not need to perform Step 3 in Algorithm 1. Furthermore, all constant rank conditions required in Assumption 1 are satisfied. We observe that

$$\begin{aligned} B_{n+1,2} &= [1 \quad n+1 \quad 1], & C_{n+1,2} &= [0 \quad 0 \quad n+1], \\ C_{n+1,3} &= [0 \quad 0 \quad n+2], & C_{n+2,3} &= [0 \quad 0 \quad n+3]. \end{aligned}$$

By directly verifying, we see that the matrix pair  $\left( A_{n,1}, \begin{bmatrix} B_{n+1,2} \\ C_{n+2,3} \end{bmatrix} \right)$  has hidden redundancy, while the pair  $(B_{n,2}, C_{n+1,3})$  does not. Due to Lemma 9 we choose  $S_n^{(2)} = [ ]$ ,  $Z_n^{(2)} = 1$ ,  $Z_n^{(4)} = -1$ ,  $Z_n^{(5)} = -1$ . Notice that the fact  $Z_n^{(5)}$  is non-empty leads to the appearance of  $f_3(n+2)$ . Furthermore, the resulting system (2.12) reads

$$\begin{bmatrix} 0 & \alpha & n+2 \\ 1 & n & 1 \\ 0 & 0 & 0 \end{bmatrix} x(n+1) + \begin{bmatrix} 0 & n+1 & 0 \\ 0 & 0 & n \\ 0 & 0 & n+1 \end{bmatrix} x(n) = \begin{bmatrix} f_1(n) - f_2(n+1) - f_3(n+2) \\ f_2(n) \\ f_3(n) \end{bmatrix}. \quad (2.15)$$

Here the matrix coefficient associated with  $x(n+2)$  becomes zero, so for notational convenience we do not write this term. Go back to Step 3, we see that two following cases may happen.

- i) If  $\alpha \neq 0$ , then Algorithm 1 terminates here, and the strangeness-index is  $\mu = 2$ , which is exactly the number of time-shift appear in the inhomogeneity  $f$  in the strangeness-free formulation (2.15).
- ii) If  $\alpha = 0$ , then the matrix pair  $\left( \begin{bmatrix} 0 & \alpha & n+2 \\ 1 & n & 1 \end{bmatrix}, [0 \quad 0 \quad n+2] \right)$  have hidden redundancy. Due to Lemma 9 we choose  $S_n^{(1)} = [1 \ 0]$ ,  $Z_n^{(1)} = [0 \ 1]$ ,  $Z_n^{(2)} = -[0 \ 1]$ .

The resulting system (2.12) now reads

$$\begin{aligned} & \begin{bmatrix} 1 & n & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} x(n+1) + \begin{bmatrix} 0 & 0 & n \\ 0 & n+1 & 0 \\ 0 & 0 & n+1 \end{bmatrix} x(n) \\ &= \begin{bmatrix} f_2(n) \\ f_1(n) - f_2(n+1) - f_3(n+2) - f_3(n+1) \\ f_3(n) \end{bmatrix}. \end{aligned} \quad (2.16)$$

192 Algorithm 1 terminates here, and the strangeness-index is  $\mu = 3$ , which is bigger  
193 than the number of time-shift appear in the inhomogeneity  $f$  in the strangeness-  
194 free formulation (2.16).

195 A direct consequence of Theorem 2 is, that we can deduce the theoretical  
196 solvability for (1.2) as follows.

197 **Corollary 1** *Under the assumption of Theorem 2, the following statements hold*  
198 *true.*

199 *i) The corresponding IVP for the SiDE (1.2) is solvable if and only if either*  
200  *$v^\mu = 0$  or  $\hat{g}_4(n) = 0$  for all  $n \geq n_0$ . Furthermore, it is uniquely solvable if, in*  
201 *addition, we have  $r_2^\mu + r_1^\mu + r_0^\mu = d$ .*

202 *ii) The initial condition (1.3) is consistent if and only if the following equalities*  
203 *hold.*

$$\begin{aligned} \hat{B}_{n_0,2}x_1 + \hat{C}_{n_0,2}x_0 &= \hat{g}_2(n_0), \\ \hat{C}_{n_0,3}x_0 &= \hat{g}_3(n_0). \end{aligned}$$

204 Another direct consequence of Theorem 2 is, that we can obtain an under-  
205 lying difference equation as follows.

**Corollary 2** *Assume that the IVP (1.2)-(1.3) is uniquely solvable for any con-*  
*sistent initial condition. Under the assumption of Theorem 2, the solution  $x$  to*  
*this IVP is also a solution to the (implicit) underlying difference equation*

$$\begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix} x(n+2) + \begin{bmatrix} \hat{B}_{n,1} \\ \hat{C}_{n+1,2} \\ 0 \end{bmatrix} x(n+1) + \begin{bmatrix} \hat{C}_{n,1} \\ 0 \\ 0 \end{bmatrix} x(n) = \begin{bmatrix} \hat{g}_1(n) \\ \hat{g}_2(n+1) \\ \hat{g}_3(n+2) \end{bmatrix}, \quad (2.17)$$

206 where the matrix  $\begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix}$  is invertible for all  $n \geq n_0$ .

207 *Remark 6* Unlike in [14, 18], we do not change the variable  $x$ . This trick permits  
208 us to simplify significantly the condensed forms in [2, 18]. We emphasize that  
209 as in (2.5), we only require five constant rank conditions within one step of  
210 index reduction, instead of seven as in [18]. Therefore, this trick will enlarge  
211 the domain of application for SiDEs (and also for DAEs, in the continuous time  
212 case). This trick is also useful for the control analysis of the descriptor system  
213 (1.1), as will be seen later.

214 *Remark 7* i) Within one loop of Algorithm 1, for each  $n$ , we have used 4 SVDs  
 215 to remove the hidden redundancies in two matrix pairs. The total cost depends  
 216 on the problems itself, i. e., depending on sizes of the matrix pairs which applied  
 217 SVDs. Nevertheless, it would not exceed  $\mathcal{O}(m^2 d^2)$ .  
 218 ii) Unfortunately, since  $Z_n^{(3)}, Z_n^{(4)}, Z_n^{(5)}$  are not orthogonal, in general Algorithm  
 219 1 could not be stably implemented. For the numerical solution to the IVP (1.2)-  
 220 (1.3), we will consider a suitable numerical scheme in Section 4.  
 221 iii) Furthermore, similar to the case of continuous time systems, the strangeness  
 222 index  $\mu$  constructed here only gives an upper bound for the number of shift-  
 223 forward operator that have been used, in order to achieve the strangeness-free  
 224 form (2.13). For further details, see Remark 17, [18]. To overcome this obstacle,  
 225 another approach will be presented in Section 4.

### 226 3 Strangeness-index of second order descriptor systems

227 Based on the index reduction procedure for SiDEs in Section 2, in this section  
 228 we construct the strangeness-index concept for the descriptor system (1.1). The  
 229 solvability analysis for first order descriptor systems with variable coefficients  
 230 have been carefully discussed in [3, 12, 20]. Nevertheless, for second order de-  
 231 scriptor systems, this problem has been rarely considered. We refer the interested  
 232 readers to [14, 23] for continuous time systems.

233 It is well known, that in regularization procedures of continuous time sys-  
 234 tems, one should avoid differentiating equations that involve an input function,  
 235 due to the fact that it may not be differentiable. Here, we will also keep this  
 236 spirit, and hence, will not shift any equation that involve an input function,  
 237 since it may destroy the causality of the considered system. In the following  
 238 lemma, we give the condensed form for system (1.1).

**Lemma 11** *Consider the descriptor system (1.1). Then, there exist two point-  
 wise nonsingular matrix sequences  $\{U_n\}_{n \geq n_0}, \{V_n\}_{n \geq n_0}$  such that the following  
 identities hold.*

$$\begin{aligned}
 & (U_n [A_n \ B_n \ C_n], U_n D_n V_n) \\
 &= \left( \begin{array}{c} \left[ \begin{array}{ccc} A_{n,1} & B_{n,1} & C_{n,1} \\ 0 & B_{n,2} & C_{n,2} \\ 0 & 0 & C_{n,3} \\ 0 & B_{n,4} & C_{n,4} \\ 0 & 0 & C_{n,5} \\ 0 & 0 & 0 \end{array} \right], \left[ \begin{array}{ccc} D_{n,1} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & \Sigma_{\varphi,1} & 0 \\ 0 & 0 & \Sigma_{\varphi,0} \\ 0 & 0 & 0 \end{array} \right] \end{array} \right), \begin{array}{l} r_{2,n} \\ r_{1,n} \\ r_{0,n} \\ \varphi_{1,n} \\ \varphi_{0,n} \\ v_n \end{array} \quad \text{for all } n \geq n_0. \quad (3.1)
 \end{aligned}$$

239 Here sizes of the block rows are  $r_{2,n}, r_{1,n}, r_{0,n}, \varphi_{1,n}, \varphi_{0,n}, v_n$ , the matrices  
 240  $A_{n,1}, B_{n,2}, B_{n,4}, C_{n,3}$  are of full row rank and the matrices  $\Sigma_{\varphi,1}, \Sigma_{\varphi,0}$  are  
 241 nonsingular and diagonal.

242 *Proof.* First we apply Lemma 5 to four matrices  $A_n, B_n, C_n$  and  $D_n$  to obtain  
 243 the matrix  $U_n$  that satisfies (1.6). Decompose the matrix  $\begin{bmatrix} \check{D}_4 \\ \check{D}_5 \end{bmatrix}$  via one SVD,  
 244 we then obtain the block  $\begin{bmatrix} 0 & \Sigma_{\varphi,1} & 0 \\ 0 & 0 & \Sigma_{\varphi,0} \end{bmatrix}$ . Finally, we use Gaussian elimination

245 to cancel out all matrices on the two columns of  $\check{D}$  that contain  $\Sigma_{\varphi,1}$  and  $\Sigma_{\varphi,0}$ ,  
 246 and hence, we obtain the desired form (3.1).  $\square$

247 In order to build an index reduction procedure for (1.1), we also need the  
 248 following assumption.

249 **Assumption 3.** *Assume that the local characteristic invariants  $r_{2,n}$ ,  $r_{1,n}$ ,  $r_{0,n}$ ,  
 250  $\varphi_{1,n}$ ,  $\varphi_{0,n}$ ,  $v_n$ , become global, i.e., they are constant for all  $n \geq n_0$ .*

Make use of Lemma 11, we can transform the descriptor system (1.1) to the following system

$$\begin{array}{l} r_2 \\ r_1 \\ r_0 \\ \varphi_1 \\ \varphi_0 \\ v \end{array} \begin{array}{c} \left[ \begin{array}{ccc} A_{n,1} & B_{n,1} & C_{n,1} \\ 0 & B_{n,2} & C_{n,2} \\ 0 & 0 & C_{n,3} \\ 0 & B_{n,4} & C_{n,4} \\ 0 & 0 & C_{n,5} \\ 0 & 0 & 0 \end{array} \right] \\ \left[ \begin{array}{c} x(n+2) \\ x(n+1) \\ x(n) \end{array} \right] + \begin{array}{c} \left[ \begin{array}{ccc} D_{n,1} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & \Sigma_{\varphi,1} & 0 \\ 0 & 0 & \Sigma_{\varphi,0} \\ 0 & 0 & 0 \end{array} \right] \\ \left[ \begin{array}{c} v_1(n) \\ v_2(n) \\ v_3(n) \end{array} \right] = \tilde{f}(n), \quad (3.2) \end{array}$$

251 where  $u(n) = V_n v(n)$ ,  $v(n) := \begin{bmatrix} v_1(n) \\ v_2(n) \\ v_3(n) \end{bmatrix}$ ,  $\tilde{f}(n) := U_n f(n)$ , for all  $n \geq n_0$ .  
 252

253 Moreover, we notice that the third and fourth block rows, whose sizes are  
 254  $\varphi_1$  and  $\varphi_0$ , are related to the feedback regularization of (1.1), as shown in the  
 255 following proposition.

**Proposition 1** *i) Assume that for each  $n \geq n_0$ , the matrix  $\begin{bmatrix} A_{n,1} \\ B_{n+1,2} \\ C_{n+2,3} \end{bmatrix}$  is of full row rank. Then, there exist two matrices sequences  $\{F_{n,1}\}_{n \geq n_0}$ ,  $\{F_{n,0}\}_{n \geq n_0}$  which take values  $\mathbb{R}^{m,d}$  such that the following matrix has full row rank*

$$\begin{array}{c} \left[ \begin{array}{c} A_{n,1} \\ B_{n+1,2} \\ C_{n+2,3} \\ \hline B_{n+1,4} + [0 \ \Sigma_{\varphi,1} \ 0] F_{n+1,1} \\ C_{n+2,5} + [0 \ 0 \ \Sigma_{\varphi,0}] F_{n+2,0} \end{array} \right]. \end{array}$$

*ii) Consequently, if the upper part of (3.2) is strangeness-free then there exists a first order feedback of the form*

$$v(n) = F_{n,1}x(n+1) + F_{n,0}x(n), \text{ for all } n \geq n_0, \quad (3.3)$$

such that the closed loop system

$$A_n x(n+2) + (B_n + D_n F_{n,1}) x(n+1) + (C_n + D_n F_{n,0}) x(n) = f(n),$$

256 is strangeness-free.

257 *Proof.* Since the part ii) is a direct consequence of part i), we only need to prove

258 i). The part i) is directly followed by applying Lemma 6 for  $P = \begin{bmatrix} A_{n,1} \\ B_{n+1,2} \\ C_{n+2,3} \end{bmatrix}$ ,

259  $Q = \begin{bmatrix} 0 & \Sigma_{\varphi,1} & 0 \\ 0 & 0 & \Sigma_{\varphi,0} \end{bmatrix}$  and  $G = \begin{bmatrix} B_{n+1,4} \\ C_{n+2,5} \end{bmatrix}$ .  $\square$

260 From Proposition 1, we see that we only need to remove the hidden redun-  
 261 dancies in the upper part of (3.2) as follows. By performing one index reduction  
 262 step for the upper part of (3.2), as in Section 2, we obtain the following lemma.

**Lemma 12** *Assume that the upper part of the descriptor system (3.2) is not strangeness-free. Then, for each input sequence  $\{v(n)\}_{n \geq n_0}$ , it has exactly the same solution set as the following system*

$$\begin{array}{l} \tilde{r}_2 \\ \tilde{r}_1 \\ \tilde{r}_0 \\ \varphi_1 \\ \varphi_0 \\ \tilde{v} \end{array} \begin{array}{c} \left[ \begin{array}{ccc} \tilde{A}_{n,1} & \tilde{B}_{n,1} & \tilde{C}_{n,1} \\ 0 & \tilde{B}_{n,2} & \tilde{C}_{n,2} \\ 0 & 0 & \tilde{C}_{n,3} \\ 0 & B_{n,4} & C_{n,4} \\ 0 & 0 & C_{n,5} \\ 0 & 0 & 0 \end{array} \right] \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix} + \begin{array}{c} \left[ \begin{array}{ccc} \tilde{D}_{n,1} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & \Sigma_{\varphi,1} & 0 \\ 0 & 0 & \Sigma_{\varphi,0} \\ 0 & 0 & 0 \end{array} \right] \begin{bmatrix} v_1(n) \\ v_2(n) \\ v_3(n) \end{bmatrix} = \tilde{f}(n), \quad (3.4) \end{array}$$

263 where  $\tilde{r}_2 = r_2 - s_2$ ,  $\tilde{r}_0 = r_0 + s_0$ ,  $\tilde{v} \geq v$ , for some  $s_2 > 0$ ,  $s_1 > 0$ .

264 *Proof.* System (3.4) is directly obtained by applying Lemma 10 to the upper  
 265 part of (3.2). To keep the brevity of this paper, we will omit the details here.  $\square$

266 Similar to the observation made in Section 2, here we also see, that an index  
 267 reduction step, which passes system (3.2) to the new form (3.4) has reduced the  
 268 upper rank  $r^u$  by at least  $s_2 + s_1$ . Continue in this way, finally we obtain the  
 269 strangeness-free descriptor system in the next theorem.

**Theorem 4** *Consider the descriptor system (1.1). Furthermore, assume that Assumption 3 is fulfilled whenever needed. Then, for each fixed input sequence  $\{u(n)\}_{n \geq n_0}$ , system (1.1) has the same solution set as the so-called strangeness-free descriptor system*

$$\begin{array}{l} \hat{r}_2 \\ \hat{r}_1 \\ \hat{r}_0 \\ \hat{\varphi}_1 \\ \hat{\varphi}_0 \\ \hat{v} \end{array} \begin{array}{c} \left[ \begin{array}{ccc} \hat{A}_{n,1} & \hat{B}_{n,1} & \hat{C}_{n,1} \\ 0 & \hat{B}_{n,2} & \hat{C}_{n,2} \\ 0 & 0 & \hat{C}_{n,3} \\ 0 & \hat{B}_{n,5} & \hat{C}_{n,5} \\ 0 & 0 & \hat{C}_{n,6} \\ 0 & 0 & 0 \end{array} \right] \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix} + \begin{array}{c} \left[ \begin{array}{c} \hat{D}_{n,1} \\ 0 \\ 0 \\ \hat{D}_{n,4} \\ \hat{D}_{n,5} \\ 0 \end{array} \right] u(n) = \begin{array}{c} \left[ \begin{array}{c} \hat{f}_1(n) \\ \hat{f}_2(n) \\ \hat{f}_3(n) \\ \hat{f}_4(n) \\ \hat{f}_5(n) \\ \hat{f}_6(n) \end{array} \right], \text{ for all } n \geq n_0, \end{array} \end{array} \quad (3.5)$$

270 where the matrices  $\begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix}$ ,  $\begin{bmatrix} \hat{D}_{n,4} \\ \hat{D}_{n,5} \end{bmatrix}$  have full row rank for all  $n \geq n_0$ .

*Proof.* By repeating index reduction steps until the upper rank  $r^u$  stop decreasing, we obtain the system

$$\begin{array}{l} \hat{r}_2 \\ \hat{r}_1 \\ \hat{r}_0 \\ \hat{\varphi}_1 \\ \hat{\varphi}_0 \\ \hat{v} \end{array} \begin{array}{c} \left[ \begin{array}{ccc} \hat{A}_{n,1} & \hat{B}_{n,1} & \hat{C}_{n,1} \\ 0 & \hat{B}_{n,2} & \hat{C}_{n,2} \\ 0 & 0 & \hat{C}_{n,3} \\ 0 & \hat{B}_{n,5} & \hat{C}_{n,5} \\ 0 & 0 & \hat{C}_{n,6} \\ 0 & 0 & 0 \end{array} \right] \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix} + \begin{array}{c} \left[ \begin{array}{ccc} \hat{D}_{n,1} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & \Sigma_{\hat{\varphi}_1} & 0 \\ 0 & 0 & \Sigma_{\hat{\varphi}_0} \\ 0 & 0 & 0 \end{array} \right] v(n) = \begin{array}{c} \left[ \begin{array}{c} \hat{f}_1(n) \\ \hat{f}_2(n) \\ \hat{f}_3(n) \\ \hat{f}_4(n) \\ \hat{f}_5(n) \\ \hat{f}_6(n) \end{array} \right], \end{array}$$



for all  $n \geq n_0$ , where the matrix  $\begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix}$  has full row rank for all  $n \geq n_0$ . Here the new input sequence  $\{v(n)\}_{n \geq n_0}$  satisfies  $u(n) = V_n v(n)$ ,  $V_n$  is nonsingular for all  $n \geq n_0$ . Transform back  $v(n) = V_n^{-1} u(n)$ , and set

$$\begin{bmatrix} \hat{D}_{n,1} \\ 0 \\ 0 \\ \hat{D}_{n,4} \\ \hat{D}_{n,5} \\ 0 \end{bmatrix} := \begin{bmatrix} \hat{D}_{n,1} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & \Sigma_{\hat{\varphi}_1} & 0 \\ 0 & 0 & \Sigma_{\hat{\varphi}_0} \\ 0 & 0 & 0 \end{bmatrix} V^{-1},$$

we obtain exactly the strangeness-free descriptor system (3.5).  $\square$

As a direct corollary of Theorem 4, we obtain the existence and uniqueness of a solution to the closed-loop system via feedback as follows.

**Corollary 3** *Under the conditions of Theorem 4, the following statements hold true.*

- i) There exists a first order feedback of the form (3.3) such that the closed-loop system is solvable if and only if either  $\hat{v} = 0$  or  $\hat{f}_6(n) = 0$  for all  $n \geq n_0$ .*
- ii) Furthermore, the solution to the corresponding IVP (of the closed-loop system) is unique if and only if in addition,  $d = \sum_{i=0}^2 \hat{r}_i + \sum_{i=0}^1 \hat{\varphi}_i$ .*

*Remark 8* It should be noted that, in analogous to SiDEs, each index reduction step of the descriptor system (1.1) also makes use of Lemma 10, where the matrices  $Z_n^{(i)}$ ,  $i = 3, 4, 5$ , may not be orthogonal. Furthermore, in Lemma 11, two matrices  $U_n$ ,  $V_n$  are only nonsingular but not orthogonal. Therefore, in general, the strangeness-free formulation (3.5) could not be stably computed. For the numerical treatment of (continuous time) second order DAEs, in [23] a different approach was developed. We will modify it for SiDEs/descriptor systems in the next section.

*Remark 9* Another interesting method while considering descriptor systems is the *behavior approach*, where we combine both the state  $x$  and input  $u$  in one behavior vector. Then, (1.1) will become a SiDE of this behavior variable, and hence, we can apply the results in Section 2 for this system. However, to keep the brevity of this research, we will not present the details here. For the interested readers, we refer to [12, 20, 21] for the case of first order DAEs, and [23] for the case of second order DAEs.

#### 4 Difference arrays of second-order SiDEs/descriptor systems

As have shown in two previous sections, to analyze the theoretical solvability of the SiDE (1.2) or of the descriptor system (1.1), first one needs to bring it to a strangeness-free formulation. Nevertheless, this task is not always doable, for example when Assumptions 1, 3 are violated at some index reduction steps. These difficulties have also been observed for continuous time systems of both first and higher orders, and they have been addressed in [12, 23]. The basic

302 idea, thanks to Campbell [4], while considering DAEs, is to differentiate a given  
 303 system a number of times and put every one of them, including the original one,  
 304 into a so-called *inflated system*. Then, the strangeness-free formulation will be  
 305 determined by appropriate selection of equations inside this inflated system. In  
 306 this section we will examine this approach to the descriptor system (1.1). The  
 307 analysis for SiDEs of the form (1.2) can be obtained by simply setting  $D_n$  to be  
 308  $0^{m,p}$  for all  $n$ . We further assume the following condition.

309 **Assumption 5.** *Consider the descriptor system (1.1). Assume that there exists*  
 310 *a first order feedback of the form (3.3) such that the corresponding IVP of the*  
 311 *closed-loop system is uniquely solvable.*

312 Notice that, in case of the SiDE (1.2), Assumption 5 means that the IVP  
 313 (1.2)-(1.3) is uniquely solvable. Now let us introduce the *difference-inflated*  
 314 *system of level  $\ell \in \mathbb{N}$*  as follows.

$$\begin{aligned} A_n x(n+2) + B_n x(n+1) + C_n x(n) + D_n u(n) &= f(n), \\ A_{n+1} x(n+3) + B_{n+1} x(n+2) + C_{n+1} x(n+1) + D_{n+1} u(n+1) &= f(n+1), \\ &\dots \\ A_{n+\ell} x(n+\ell+2) + B_{n+\ell} x(n+\ell+1) + C_{n+\ell} x(n+\ell) + D_{n+\ell} u(n+\ell) &= f(n+\ell). \end{aligned}$$

We rewrite this system as

$$\begin{aligned} &\underbrace{\begin{bmatrix} C_n & B_n & A_n & & & \\ & C_{n+1} & B_{n+1} & A_{n+1} & & \\ & & \ddots & \ddots & \ddots & \\ & & & \ddots & \ddots & \ddots \\ & & & & C_{n+\ell} & B_{n+\ell} & A_{n+\ell} \end{bmatrix}}_{=: \mathcal{M}} \underbrace{\begin{bmatrix} x(n) \\ x(n+1) \\ x(n+2) \\ \vdots \\ x(n+\ell) \end{bmatrix}}_{=: \mathcal{X}} + \\ &+ \underbrace{\begin{bmatrix} D_n & & & & \\ & D_{n+1} & & & \\ & & \ddots & & \\ & & & & D_{n+\ell} \end{bmatrix}}_{=: \mathcal{N}} \underbrace{\begin{bmatrix} u(n) \\ u(n+1) \\ \vdots \\ u(n+\ell) \end{bmatrix}}_{=: \mathcal{U}} = \underbrace{\begin{bmatrix} f(n) \\ f(n+1) \\ \vdots \\ f(n+\ell) \end{bmatrix}}_{=: \mathcal{G}}, \quad \text{for all } n \geq n_0. \quad (4.1) \end{aligned}$$

315

316 **Definition 3** Suppose that the descriptor system (1.1) satisfies Assumption 5.  
 317 The minimum number  $\ell$  such that by using elementary matrix's row operations,  
 318 a strangeness-free descriptor system of the form (3.5) can be extracted from (4.1)  
 319 is called the *shift-index* of (1.1), and be denoted by  $\nu$ .

320 We give the relation between this shift-index  $\nu$  and the strangeness-index  $\mu$   
 321 in the following proposition.

322 **Proposition 2** *Suppose that the descriptor system (1.1) satisfies Assumption*  
 323 *5. If the strangeness-index  $\mu$  is well-defined, then so is the shift-index  $\nu$ . Fur-*  
 324 *thermore, we have that  $\nu \leq \mu$ .*

325 *Proof.* The first claim is straight forward, since every reformulation step per-  
 326 formed in Algorithm 1 is a consequence of an inflated system (4.1) with  $\ell =$   
 327  $\mu$ .  $\square$

328 *Remark 10* As will be seen later in Example 3, for second order SiDEs, the shift  
329 index can be strictly smaller than the strangeness index.

Assume that  $\nu$  is already known, now we construct an algorithm to select the strangeness-free descriptor system (3.5) from the inflated system (4.1). For notational convenience, we will follow the Matlab language, [16]. Consider the following spaces and matrices

$$\mathcal{W} := [\mathcal{M}(:, 3n+1 : \text{end}) \quad \mathcal{N}(:, n+1 : \text{end})], \quad (4.2)$$

$U_1$  basis of  $\text{kernel}(\mathcal{W}^T)$ , and  $U_{1,\perp}$  basis of  $\text{range}(\mathcal{W})$ ,

due to Lemma 4 we have that  $U_1^T \mathcal{W} = 0$  and  $U_{1,\perp}^T \mathcal{W}$  has full row rank. Furthermore, the matrix  $\begin{bmatrix} U_1^T \\ U_{1,\perp}^T \end{bmatrix}$  is nonsingular, and hence system (4.1) is equivalent to the system below.

$$U_1^T \mathcal{M}(:, 1 : 3n) \begin{bmatrix} x(n) \\ x(n+1) \\ x(n+2) \end{bmatrix} + U_1^T \mathcal{N}(:, 1 : n) u(n) = U_1^T \mathcal{G}, \quad (4.3)$$

$$U_{1,\perp}^T \mathcal{W} \begin{bmatrix} x(n+3) \\ \vdots \\ x(n+\nu) \\ u(n+1) \\ \vdots \\ u(n+\nu) \end{bmatrix} + U_{1,\perp}^T [\mathcal{M}(:, 1 : 3n) \quad \mathcal{N}(:, 1 : n)] \begin{bmatrix} x(n) \\ x(n+1) \\ x(n+2) \\ u(n) \end{bmatrix} = U_{1,\perp}^T \mathcal{G}. \quad (4.4)$$

330 Notice that due to the full row rank property of  $U_{1,\perp}^T \mathcal{W}$ , (4.4) plays no role in  
331 the determination of the strangeness-free descriptor system (3.5). Thus, (3.5) is  
332 a consequence of (4.3). In the following proposition we show that system (4.3)  
333 is not affected by left equivalence transformation.

334 **Proposition 3** Consider two left equivalent systems. Then, at the same level  
335  $\ell$ , their difference-inflated systems of the form (4.1) are also left equivalent.  
336 Consequently, system (4.3) is not affected by left equivalence transformation.

*Proof.* Let us assume that (1.1) is left equivalent to the SiDE

$$\tilde{A}_n x(n+2) + \tilde{B}_n x(n+1) + \tilde{C}_n x(n) + \tilde{D}_n u(n) = \tilde{f}(n), \quad \text{for all } n \geq n_0. \quad (4.5)$$

Thus, there exists a pointwise nonsingular matrix sequence  $\{P_n\}_{n \geq n_0}$  such that

$$[\tilde{A}_n \quad \tilde{B}_n \quad \tilde{C}_n \quad \tilde{D}_n] = P_n [A_n \quad B_n \quad C_n \quad D_n] \quad \text{and} \quad \tilde{f}(n) = P_n f(n), \quad \text{for all } n \geq n_0.$$

Therefore, the difference-inflated system of level  $\ell$  for system (4.5) takes the form

$$\tilde{\mathcal{M}} \mathcal{X} + \tilde{\mathcal{N}} \mathcal{U} = \tilde{\mathcal{G}}, \quad (4.6)$$

where the matrix coefficients are

$$\tilde{\mathcal{M}} = \text{diag}(P_n, \dots, P_{n+\ell}) \mathcal{M}, \quad \tilde{\mathcal{N}} = \text{diag}(P_n, \dots, P_{n+\ell}) \mathcal{N}, \quad \tilde{\mathcal{G}} = \text{diag}(P_n, \dots, P_{n+\ell}) \mathcal{G}.$$

337 This follows that two systems (4.1) and (4.6) are left equivalent, which finishes  
338 the proof.  $\square$

For notational convenience, let us rewrite system (4.3) as

$$\left[ \check{A} \quad \check{B} \quad \check{C} \mid \check{D} \right] \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \\ u(n) \end{bmatrix} = \check{G}.$$

Scale this system with the matrix  $\check{U}$  obtained in Lemma 5, we have

$$\left[ \begin{array}{ccc|c} \check{A}_1 & \check{B}_1 & \check{C}_1 & \check{D}_1 \\ 0 & \check{B}_2 & \check{C}_2 & 0 \\ 0 & 0 & \check{C}_3 & 0 \\ 0 & 0 & 0 & 0 \\ \hline 0 & \check{B}_4 & \check{C}_4 & \check{D}_4 \\ 0 & 0 & \check{C}_5 & \check{D}_5 \end{array} \right] \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \\ u(n) \end{bmatrix} = \begin{bmatrix} \check{G}_1 \\ \check{G}_2 \\ \check{G}_3 \\ 0 \\ \check{G}_4 \\ \check{G}_5 \end{bmatrix}. \quad (4.7)$$

339 Here the matrices  $\check{A}_1$ ,  $\check{B}_2$ ,  $\check{B}_4$ ,  $\check{C}_3$ , and  $\begin{bmatrix} \check{D}_4 \\ \check{D}_5 \end{bmatrix}$  have full row rank. Notice that the  
 340 presence of the 0 block on the right hand side vector is due to Assumption 5. In  
 341 the following theorem we answer the question how to derive the strangeness-free  
 342 formulation (3.5) from (4.7).

**Theorem 6** *Assume that the shift index  $\nu$  of the descriptor system (1.1) is well-defined. Furthermore, suppose that (1.1) satisfies Assumption 5. Then, any solution to the descriptor system (1.1) is also a solution to the following system*

$$\begin{array}{l} \hat{r}_2 \\ \hat{r}_1 \\ \hat{r}_0 \\ \hat{\varphi}_1 \\ \hat{\varphi}_0 \end{array} \begin{bmatrix} \hat{A}_{n,1} & \hat{B}_{n,1} & \hat{C}_{n,1} \\ 0 & \hat{B}_{n,2} & \hat{C}_{n,2} \\ 0 & 0 & \hat{C}_{n,3} \\ \hline 0 & \hat{B}_{n,5} & \hat{C}_{n,5} \\ 0 & 0 & \hat{C}_{n,6} \end{bmatrix} \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix} + \begin{bmatrix} \hat{D}_{n,1} \\ 0 \\ 0 \\ \hline \hat{D}_{n,4} \\ \hat{D}_{n,5} \end{bmatrix} u(n) = \begin{bmatrix} \hat{G}_{n,1} \\ \hat{G}_{n,2} \\ \hat{G}_{n,3} \\ \hline \hat{G}_{n,4} \\ \hat{G}_{n,5} \end{bmatrix}, \text{ for all } n \geq n_0, \quad (4.8)$$

where the matrices  $\begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix}$ ,  $\begin{bmatrix} \hat{D}_{n,4} \\ \hat{D}_{n,5} \end{bmatrix}$  have full row rank for all  $n \geq n_0$ . Furthermore,  $\sum_{i=0}^2 \hat{r}_i + \sum_{i=0}^1 \hat{\varphi}_i = d$ , or equivalently,

$$\text{rank} \left( \begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix} \right) + \text{rank} \left( \begin{bmatrix} \hat{D}_{n,4} \\ \hat{D}_{n,5} \end{bmatrix} \right) = d. \quad (4.9)$$

*Proof.* First we will extract the first two block row equations of system (4.8) from (4.7), by suitably removing the existence hidden redundancy. Applying Lemma 4 consecutively for two following matrix pairs  $(\check{B}_2, \check{C}_3)$ ,  $(\check{A}_1, \begin{bmatrix} \check{B}_2 \\ \check{C}_3 \end{bmatrix})$ ,

we obtain two orthogonal matrices  $\begin{bmatrix} S_n^{(i)} \\ Z_n^{(i)} \end{bmatrix} \in \mathbb{R}^{r_i, r_i}$ ,  $i = 1, 2$  such that both pairs

$(S_n^{(1)} \check{B}_2, \check{C}_3)$ ,  $(S_n^{(2)} \check{A}_1, \begin{bmatrix} \check{B}_2 \\ \check{C}_3 \end{bmatrix})$  have no hidden redundancy. Scale the first and second block row equations of (4.7) with  $S_n^{(2)}$  and  $S_n^{(1)}$  respectively, we obtain

$$\left[ \begin{array}{ccc|c} S_n^{(2)} \check{A}_1 & S_n^{(2)} \check{B}_1 & S_n^{(2)} \check{C}_1 & S_n^{(2)} \check{D}_1 \\ 0 & S_n^{(1)} \check{B}_2 & S_n^{(1)} \check{C}_2 & 0 \end{array} \right] \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \\ u(n) \end{bmatrix} = \begin{bmatrix} S_n^{(2)} \check{G}_1 \\ S_n^{(1)} \check{G}_2 \end{bmatrix}.$$

Combining these equations with the third, fifth and sixth block equations of (4.7), we obtain the system

$$\left[ \begin{array}{ccc|c} S_n^{(2)} \check{A}_1 & S_n^{(2)} \check{B}_1 & S_n^{(2)} \check{C}_1 & S_n^{(2)} \check{D}_1 \\ 0 & S_n^{(1)} \check{B}_2 & S_n^{(1)} \check{C}_2 & 0 \\ 0 & 0 & \check{C}_3 & 0 \\ \hline 0 & \check{B}_4 & \check{C}_4 & \check{D}_4 \\ 0 & 0 & \check{C}_5 & \check{D}_5 \end{array} \right] \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \\ u(n) \end{bmatrix} = \begin{bmatrix} S_n^{(2)} \check{G}_1 \\ S_n^{(1)} \check{G}_2 \\ \check{G}_3 \\ \check{G}_4 \\ \check{G}_5 \end{bmatrix}. \quad (4.10)$$

343 which is exactly our desired system (4.8). Moreover, due to Lemma 2, the ma-  
 344 trix  $\begin{bmatrix} S_n^{(2)} \check{A}_1 \\ S_n^{(1)} \check{B}_2 \\ \check{C}_3 \end{bmatrix}$  has full row rank. Finally, the identity (4.9) holds true due to  
 345 Assumption 5. □

346 We summarize our result in the following algorithm.

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**Algorithm 2** Strangeness-free formulation for SiDEs using difference arrays

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- 1: **Input:** The SiDE (1.1).
  - 2: **Return:** The strangeness-free descriptor system (4.8).
  - 3: Set  $\ell := 0$ .
  - 4: Construct the difference-inflated system of level  $\ell$ , and rewrite it in the form (4.1).
  - 5: Find  $U_1$  as in (4.2) and construct system (4.3).
  - 6: Find  $\check{U}$  as in Lemma 5 and construct system (4.7).
  - 7: Find the matrices  $S_n^{(1)}$ ,  $S_n^{(2)}$  in the process used to remove the hidden redundancies in two matrix pairs  $(\check{B}_2, \check{C}_3)$ ,  $(\check{A}_1, \begin{bmatrix} \check{B}_2 \\ \check{C}_3 \end{bmatrix})$ , respectively.
  - 8: Construct the system (4.10).
  - 9: **if**  $\text{rank} \begin{bmatrix} \hat{A}_{n,1} \\ \hat{B}_{n+1,2} \\ \hat{C}_{n+2,3} \end{bmatrix} + \text{rank} \begin{bmatrix} \hat{D}_{n,4} \\ \hat{D}_{n,5} \end{bmatrix} = d$  **then** STOP.
  - 10: **else** set  $\ell := \ell + 1$  and go to 4
  - 11: **end if**
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347 In order to illustrate Algorithm 2, we consider two following examples.

*Example 3* Let us revisit system (2.14) for the case  $\alpha = 0$ . In this system,  $D_n = 0$  for all  $n \geq 0$ . For  $\ell = 2$ , the inflated system (4.1) reads

$$\left[ \begin{array}{ccc|cc} C_n & B_n & A_n & 0 & 0 \\ 0 & C_{n+1} & B_{n+1} & A_{n+1} & 0 \\ 0 & 0 & C_{n+2} & B_{n+2} & A_{n+2} \end{array} \right] \begin{bmatrix} x(n) \\ x(n+1) \\ x(n+2) \\ x(n+3) \\ x(n+4) \end{bmatrix} = \begin{bmatrix} f(n) \\ f(n+1) \\ f(n+2) \end{bmatrix} \quad (4.11)$$

Let  $U_1$  be the basis of  $\text{kernel}(\mathcal{W}^T)$ , where  $\mathcal{W} = \begin{bmatrix} 0 & 0 \\ A_{n+1} & 0 \\ B_{n+2} & A_{n+2} \end{bmatrix}$ . We then compute system (4.3) by scaling (4.11) with  $U_1^T$ . The resulting system reads

$$U_1^T \begin{bmatrix} C_n & B_n & A_n \\ 0 & C_{n+1} & B_{n+1} \\ 0 & 0 & C_{n+2} \end{bmatrix} \begin{bmatrix} x(n) \\ x(n+1) \\ x(n+2) \end{bmatrix} = U_1^T \begin{bmatrix} f(n) \\ f(n+1) \\ f(n+2) \end{bmatrix}. \quad (4.12)$$

348 Finally, by performing Steps 6 to 10 we can extract the strangeness-free form  
349 (2.16) from (4.12). Thus, we conclude that the shift index is  $\nu = 2$ .

*Example 4* Our consider system, which describes a three link robot arm [8], is of the form

$$\begin{bmatrix} M_0 & 0 \\ 0 & 0 \end{bmatrix} \ddot{x}(t) + \begin{bmatrix} G_0 & 0 \\ 0 & 0 \end{bmatrix} \dot{x}(t) + \begin{bmatrix} K_0 & H_0^T \\ H_0 & 0 \end{bmatrix} x(t) = \begin{bmatrix} B_0 \\ 0 \end{bmatrix} u(t).$$

Here  $M_0$  represents the nonsingular mass matrix,  $G_0$  the coefficient matrix associated with damping, centrifugal, gravity, and Coriolis forces,  $K_0$  the stiffness matrix, and  $H_0$  the constraint. A simple discretized version of this system takes the form

$$\begin{bmatrix} M_0 & 0 \\ 0 & 0 \end{bmatrix} \frac{x(n+2) - 2x(n+1) + x(n)}{h^2} + \begin{bmatrix} G_0 & 0 \\ 0 & 0 \end{bmatrix} \frac{x(n+2) - x(n+1)}{h} + \begin{bmatrix} K_0 & H_0^T \\ H_0 & 0 \end{bmatrix} x(n) = \begin{bmatrix} B_0 \\ 0 \end{bmatrix} u(n).$$

350 where  $h$  is the discretized stepsize.

351 As a simple example, let us take  $M_0 = G_0 = K_0 = H_0 = B_0 = 1$ ,  $h = 0.01$ .  
352 Then, Algorithm 2 terminates after two steps and hence, the shift index is  $\nu = 2$   
353 for all  $n \geq n_0$ . Furthermore, we notice that no matter forward or backward  
354 approximations has been chosen for discretizing the derivative  $\dot{x}(t)$ , the shift  
355 index remains unchanged  $\nu = 2$ . Nevertheless, the resulting strangeness-free  
356 descriptor systems are different.

## 357 5 Conclusion

358 By using the algebraic approach, we have analyzed the solvability analysis of  
359 second order SiDEs/descriptor systems, based on derived condensed forms constructed  
360 under certain constant rank assumptions. In comparison to well-known  
361 results [18, 22], we have reduce the number of constant rank conditions in every  
362 index reduction step from seven to five. This would enlarge the domain of

363 application for SiDEs (and also for DAEs). However, requiring constant rank  
364 assumptions in the discrete-time case seems less nature than in the continuous-  
365 time case. To overcome this limitation, we also consider the difference-array  
366 method, which is numerically applicable. We also notice that the backward time  
367 case ( $n \leq n_0$ ) can be directly extended from the forward time case, as it has  
368 been done in [2]. The analysis of two way case, which happens while considering  
369 boundary value problems for DAEs, have presented many difficulties, is under  
370 our research.

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## 429 A Proof of Lemma 10

430 First we prove that any solution to (2.10) is also a solution to (2.12). Notice that, due  
431 to Lemma 8, two systems (2.6) and (2.10) have identical solution set. Thus, we only  
432 need to prove that (2.10) and (2.12) are equivalent.

433 **Necessity:** The main idea here is to apply elementary row transformations to system  
434 (2.10) to obtain (2.12). Notice that we use only two elementary block row operations:

- 435 i) scaling a block row equation with a nonsingular matrix,  
436 ii) add to one row a linear combinations of another rows.



Firstly, by scaling the first (resp., second) block row equation of (2.10) with a unitary matrix  $\begin{bmatrix} S_n^{(2)} \\ Z_n^{(2)} \end{bmatrix}$  (resp.,  $\begin{bmatrix} S_n^{(1)} \\ Z_n^{(1)} \end{bmatrix}$ ), we obtain an equivalent system to (2.6), as follows

$$\begin{bmatrix} S_n^{(2)} A_{n,1} & S_n^{(2)} B_{n,1} & S_n^{(2)} C_{n,1} \\ Z_n^{(2)} A_{n,1} & Z_n^{(2)} B_{n,1} & Z_n^{(2)} C_{n,1} \\ \hline 0 & S_n^{(1)} B_{n,2} & S_n^{(1)} C_{n,2} \\ 0 & Z_n^{(1)} B_{n,2} & Z_n^{(1)} C_{n,2} \\ \hline 0 & 0 & C_{n,3} \\ 0 & 0 & 0 \\ \hline 0 & C_{n+1,3} & 0 \\ B_{n+1,2} & C_{n+1,2} & 0 \\ C_{n+2,3} & 0 & 0 \end{bmatrix} \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix} = \begin{bmatrix} S_n^{(2)} f_1(n) \\ Z_n^{(2)} f_1(n) \\ \hline S_n^{(1)} f_2(n) \\ Z_n^{(1)} f_2(n) \\ \hline f_3(n) \\ f_4(n) \\ \hline f_3(n+1) \\ f_2(n+1) \\ f_3(n+2) \end{bmatrix}, \quad \begin{matrix} d_2 \\ s_2 \\ d_1 \\ s_1 \\ r_0 \\ v \\ r_0 \\ r_0 \end{matrix}. \quad (\text{A.1})$$

By adding the seventh row scaled with  $Z_n^{(3)}$  to the fourth row of (A.1) and making use of (2.11a) we obtain the first hidden constraint

$$Z_n^{(1)} C_{n,2} x(n) = Z_n^{(1)} f_2(n) + Z_n^{(3)} f_3(n+1),$$

437 which is exactly the fourth row of (2.12).

We continue by adding the seventh row scaled with  $Z_n^{(4)}$  and the eighth row scaled with  $Z_n^{(5)}$  to the second row of (A.1) and making use of (2.11b) to obtain

$$\begin{aligned} & (Z_n^{(2)} B_{n,1} + Z_n^{(4)} C_{n+1,2}) x(n+1) + Z_n^{(2)} C_{n,1} x(n) \\ & = Z_n^{(2)} f_1(n) + Z_n^{(4)} f_2(n+1) + Z_n^{(5)} f_3(n+2). \end{aligned}$$

This is exactly the second row of (2.12). Therefore, any solution to (2.6) is also a solution to (2.12).

**Sufficiency:** Let  $x$  be an arbitrary solution to (2.12). Thus,  $x$  is also a solution to the shifted system

$$\begin{matrix} d_2 \\ s_2 \\ d_1 \\ s_1 \\ r_0 \\ v \\ r_0 \\ r_0 \end{matrix} \begin{bmatrix} S_n^{(2)} A_{n,1} & S_n^{(2)} B_{n,1} & S_n^{(2)} C_{n,1} \\ 0 & Z_n^{(2)} B_{n,1} + Z_n^{(4)} C_{n+1,2} & Z_n^{(2)} C_{n,1} \\ \hline 0 & S_n^{(1)} B_{n,2} & S_n^{(1)} C_{n,2} \\ 0 & 0 & Z_n^{(1)} C_{n,2} \\ \hline 0 & 0 & C_{n,3} \\ 0 & 0 & 0 \\ \hline 0 & C_{n+1,3} & 0 \\ C_{n+2,3} & 0 & 0 \end{bmatrix} \begin{bmatrix} x(n+2) \\ x(n+1) \\ x(n) \end{bmatrix} \\ = \begin{bmatrix} S_n^{(2)} f_1(n) \\ Z_n^{(2)} f_1(n) + Z_n^{(4)} f_2(n+1) + Z_n^{(5)} f_3(n+2) \\ \hline S_n^{(1)} f_2(n) \\ Z_n^{(1)} f_2(n) + Z_n^{(3)} f_3(n+1) \\ \hline f_3(n) \\ f_4(n) \\ \hline f_3(n+1) \\ f_3(n+2) \end{bmatrix}, \quad \text{for all } n \geq n_0. \quad (\text{A.2})$$

438 Since elementary matrix row operations are reversible, we can reverse the transforma-  
439 tions performed in the necessity part. Consequently, we see that any solution to (A.2)  
440 is also a solution to (A.1), and hence, this completes the proof.