

METHODS FOR COMPUTATION OF SOLUTIONS TO REGULAR DISCRETE-TIME LINEAR SYSTEMS

TADEUSZ KACZOREK*

Two new methods for the computation of solutions to regular discrete-time linear systems are presented. The first of them is an extension of the Dias-Mesquista method for regular discrete-time linear systems. The other is based on an expansion in a series of the inverse matrix $[Ez - A]^{-1}$. The methods are compared with the Weierstrass-Kronecker decomposition method and the Drazin inverse method. Relationships between the coefficient matrices of the four methods are established. A new mixed method is presented.

1. Introduction

Generalised (descriptor, singular) continuous-time and discrete-time linear systems have been considered in many papers and books (see References). An interesting survey of regular (singular) linear systems has been given by Lewis (1986).

Consider a discrete-time linear system described by the equation

$$Ex_{i+1} = Ax_i + Bu_i, \quad i = 0, 1, \dots \quad (1)$$

where $x_i \in \mathbb{R}^n$ is the local semistate vector, $u_i \in \mathbb{R}^m$ is the input vector and E, A, B are real matrices of appropriate dimensions. It is assumed that $\det E = 0$ and

$$\det[Ez - A] \neq 0 \quad (2)$$

for some $z \in \mathbb{C}$ (the field of complex numbers).

System (1) is called regular if (2) holds and it is called standard if E is equal to the identity matrix. It is well-known (Aplevich, 1991; Campbell, 1976; Dai, 1989; Gantmacher, 1959; Kaczorek, 1993; Lewis, 1986; Wonham, 1979) that if (2) holds, then eqn. (1) has a unique solution for any input sequence $\{u_i\}$ and admissible initial conditions x_0 .

Four different methods of finding the solution x_i to (1) will be presented and a comparative study of them will be given. First, the method based on the Weierstrass-Kronecker decomposition of the regular pencil $[Ez - A]$ will be presented. Next, the method based on the Drazin inverse (Campbell *et al.*, 1976; Campbell, 1980; Gantmacher, 1959; Kaczorek, 1993) will be considered. The third method will be an

* Institute of Control and Industrial Electronics, Warsaw University of Technology,
ul. Koszykowa 75, 00-662 Warsaw, Poland

extension for discrete-time linear systems of the method given by Dias and Mesquita (1990). The fourth method will be based on expansion in a series of the inverse matrix $[Ez - A]^{-1}$.

2. Weierstrass-Kronecker Decomposition Method

It is well-known (Aplevich, 1991; Gantmacher, 1959; Kaczorek, 1993) that if (2) holds, then there exist non-singular matrices $P, Q \in \mathbb{R}^{n \times n}$ such that

$$P[Ez - A]Q = \begin{bmatrix} I_{n_1}z - A_1 & 0 \\ 0 & Nz - I_{n_2} \end{bmatrix} \tag{3}$$

where I_k is the $k \times k$ identity matrix, n_1 is the degree of $\det [Ez - A]$, $A_1 \in \mathbb{R}^{n_1 \times n_1}$ and $N \in \mathbb{R}^{n_2 \times n_2}$ ($n_2 = n - n_1$) is a nilpotent matrix with index q , i.e. $N^{q-1} \neq 0$ and $N^q = 0$.

There exist several methods for computing the matrices P and Q (Aplevich, 1991; Dai, 1989; Gantmacher, 1959; Kaczorek, 1993; Lewis, 1986). Among them one which is worth recommending is as follows. Let z_i be the i -th root of $\det [Ez - A] = 0$ and

$$m_i := \dim \text{Ker}[Ez_i - A] \tag{4}$$

where Ker denotes the kernel (null space).

Compute the eigenvectors v_{ij}^1 defined by

$$[Ez_i - A]v_{ij}^1 = 0 \quad \text{for } j = 1, \dots, m_i \tag{5}$$

and next v_{ij}^{k+1} from

$$[Ez_i - A]v_{ij}^{k+1} = -Ev_{ij}^k \quad \text{for } k \geq 1 \tag{6}$$

Let $m_\infty := \dim \ker E = n - \text{rank } E$. Compute the infinite eigenvectors $v_{\infty j}^1$ defined by

$$Ev_{\infty j}^1 = 0 \quad \text{for } j = 1, \dots, m_\infty \tag{7}$$

and next $v_{\infty j}^{k+1}$ from

$$Ev_{\infty j}^{k+1} = Av_{\infty j}^k \quad \text{for } k \geq 1 \tag{8}$$

Arrange the eigenvectors as the columns of the matrices

$$Q = [v_{ij}^k \quad v_{\infty j}^k], \quad P^{-1} = [Ev_{ij}^k \quad Av_{\infty j}^k] \tag{9}$$

Using (5)–(8) it is easy to check that

$$[Ez - A] \begin{bmatrix} v_{ij}^k & v_{\infty j}^k \end{bmatrix} = \begin{bmatrix} Ev_{ij}^k & Av_{\infty j}^k \end{bmatrix} \begin{bmatrix} I_{n_1}z - A_1 & 0 \\ 0 & Nz - I_{n_2} \end{bmatrix} \tag{10}$$

Premultiplying (10) by $P = [E \ v_j^k]$ we obtain (3) with P, Q defined by (9). Premultiplying (1) by P , defining

$$\begin{bmatrix} x_1^i \\ x_2^i \end{bmatrix} := Q^{-1}x_i, \quad \dim x_1^i = n_1, \quad \dim x_2^i = n_2$$

and using (3) we obtain

$$x_{i+1}^i = A_1 x_1^i + B_1 u_i, \quad i = 0, 1, \dots \tag{11}$$

and

$$N x_{i+1}^i = x_2^i + B_2 u_i, \quad i = 0, 1, \dots \tag{12}$$

where

$$\begin{bmatrix} B_1 \\ B_2 \end{bmatrix} := PB, \quad B_1 \in \mathbb{R}^{n_1 \times m}, \quad B_2 \in \mathbb{R}^{n_2 \times m}$$

The solutions x_1^1, x_2^2 to (11) and (12) are respectively given by

$$x_1^i = A_1^i x_0^1 + \sum_{k=0}^{i-1} A_1^{i-k-1} B_1 u_k \tag{13}$$

and

$$x_2^i = - \sum_{k=0}^{q-1} N^k B_2 u_{i+k} \tag{14}$$

Therefore the solution x_i to (1) is given by

$$x_i = Q \begin{bmatrix} A_1^i Q_1 x_0 + \sum_{k=0}^{i-1} A_1^{i-k-1} B_1 u_k \\ - \sum_{k=0}^{q-1} N^k B_2 u_{i+k} \end{bmatrix} \tag{15}$$

where $Q^{-1} = \begin{bmatrix} Q_1 \\ Q_2 \end{bmatrix}$, $Q_1 \in \mathbb{R}^{n_1 \times n}$.

From (13) and (14) it follows that the set of admissible initial conditions x_0^1 and x_0^2 is given by

$$S_{x_0^1, x_0^2} := \left\{ \mathbb{R}^{n_1} \oplus \text{Im} [B_2, N B_2, \dots, N^{q-1} B_2] \right\} \tag{16}$$

where Im denotes the image (range). The method will be illustrated by two examples with a non-singular and singular matrix A .

Example 1. Consider eqn. (1) with

$$E = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \quad (17)$$

In this case

$$\det[Ez - A] = \begin{vmatrix} z-1 & 0 & -1 \\ 0 & z-1 & 0 \\ -1 & 0 & 0 \end{vmatrix} = 1 - z = 0$$

and $n_1 = 1$, $n_2 = 2$, $z_1 = 1$, $m_1 = \dim \ker[Ez_1 - A] = 1$.

Using (5), (7) and (8) we obtain

$$[Ez_1 - A]v_1 = \begin{bmatrix} 0 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 0 \end{bmatrix} v_1 = 0, \quad v_1 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad Ev_1 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

$$Ev_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} v_2 = 0, \quad v_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \quad Av_2 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

and

$$Ev_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} v_3 = Av_2 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad v_3 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad Av_3 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

Then from (9) and (3) we get

$$Q = [v_1 \ v_2 \ v_3] = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}, \quad P^{-1} = [Ev_1, Av_2, Av_3] = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

and

$$\begin{aligned}
 P[Ex - A]Q &= \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & -1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} z-1 & 0 & -1 \\ 0 & z-1 & 0 \\ -1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \\
 &= \begin{bmatrix} z-1 & 0 & 0 \\ 0 & -1 & z \\ 0 & 0 & -1 \end{bmatrix}
 \end{aligned}$$

Therefore

$$A_1 = 1, \quad N = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \quad q = 2, \quad \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} = PB = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

and from (15) we have

$$x_i = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_0^1 \\ -u_{i+1} \\ -u_i \end{bmatrix} = \begin{bmatrix} -u_i \\ x_0^1 \\ -u_{i+1} \end{bmatrix} \quad i = 0, 1, \dots \tag{18}$$

Example 2. Consider eqn. (1) with

$$E = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & -1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \tag{19}$$

In this case

$$\det[Ex - A] = \begin{vmatrix} z-1 & 0 & -1 \\ 0 & z-1 & 0 \\ 1 & 0 & 0 \end{vmatrix} = z(z-1) = 0$$

and $n_1 = 2, n_2 = 1, z_1 = 1, z_2 = 0, m_1 = \dim \ker[Ex_1 - A] = 1$.

In a similar way as in Example 1 we compute

$$v_1 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad E^{v_1} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad [Ex_2 - A]v_2 = \begin{bmatrix} -1 & 0 & -1 \\ 0 & -1 & 0 \\ 1 & 0 & 1 \end{bmatrix} v_2 = 0$$

$$v_2 = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \quad Ev_2 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

$$Ev_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} v_3 = 0, \quad v_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \quad Av_3 = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

and

$$Q = [v_1 \ v_2 \ v_3] = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & -1 & 1 \end{bmatrix}, \quad P^{-1} = [Ev_1, Ev_2, Av_3] = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

From (9) and (3) we have

$$\begin{aligned} P[Es - A]Q &= \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} s-1 & 0 & -1 \\ 0 & s-1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & -1 & 1 \end{bmatrix} \\ &= \begin{bmatrix} s-1 & 0 & 0 \\ 0 & s & 0 \\ 0 & 0 & -1 \end{bmatrix} \end{aligned}$$

Therefore

$$A_1 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \quad N = 0, \quad q = 1, \quad \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} = PB = \begin{bmatrix} 0 \\ -2 \\ -1 \end{bmatrix}$$

and from (15) we obtain

$$x_i = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} a \\ 2u_{i-1} \\ u_i \end{bmatrix} = \begin{bmatrix} 2u_{i-1} \\ a \\ u_i - 2u_{i-1} \end{bmatrix} \quad (20)$$

where a is any real number (a is equal to the first component of $x_0^1 = [x_{20} \ x_{10}]^T$, T denotes the transposition).

Using (16) it is easy to check that the set of admissible initial conditions x_0 in this case is given by

$$S_{x_0} = Q \left\{ \mathbb{R}^{n_1} \oplus \text{Im}[B_2, NB_2, \dots, N^{q-1}B_2] \right\} = \mathbb{R}^3 \quad (21)$$

3. Drazin Inverse Method

The smallest non-negative integer q is called the index of A if $\text{rank } A^q = \text{rank } A^{q+1}$. A matrix A^D is called the Drazin inverse of a square matrix if: i) $AA^D = A^D A$, ii) $A^D A A^D = A^D$, iii) $A^D A^{q+1} = A^q$, where q is the index of A (Campbell *et al.*, 1976; Kaczorek, 1993). The Drazin inverse A^D of a square matrix A always exists and is unique. If $\det A \neq 0$, then $A^D = A^{-1}$, where A^{-1} is the classical inverse of A . There exist several methods for computing A^D of A (Campbell *et al.*, 1976; Kaczorek, 1993). Two of them will be presented here.

The first method is based on the factorisation VM of A^q (Kaczorek, 1993). Let $A \in \mathbb{R}^n \times m$ and

$$A^q = VM^T, \quad V \in \mathbb{R}^n \times r, \quad M^T \in \mathbb{R}^r \times n \tag{22}$$

where $\ker V = \{0\}$ and $M^T M = I_r$, $r = \text{rank } A^q$. The Drazin inverse of A is given by (Kaczorek, 1993)

$$A^D = V[M^T AV]^{-1}M^T \tag{23}$$

Example 3. Find the Drazin inverse of the matrix

$$E = \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 \\ -\frac{1}{2} & 0 & 0 \end{bmatrix} \tag{24}$$

The index of (24) is equal to $q = 1$ since $r = \text{rank } E^2 = 2$ and by (22) we have

$$E = VM^T \quad \text{for} \quad V = \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & 1 \\ -\frac{1}{2} & 0 \end{bmatrix}, \quad M^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

Using (23) we obtain

$$\begin{aligned} E^D &= V[M^T EV]^{-1}M^T \\ &= \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & 1 \\ -\frac{1}{2} & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{4} & 0 \\ 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & 0 & 0 \end{bmatrix} \end{aligned} \tag{25}$$

The second method is based on the following algorithm (Campbell and Meyer, 1979):

Step 1. Set $S_0 = I_n$ and compute recursively $S_j = AS_{j-1} + \beta_{n-j}I_n$, $\beta_{n-j} = -\frac{1}{j}\text{tr}[AS_{j-1}]$ (tr denotes the trace) until some $S_i = 0$ but $S_{i-1} \neq 0$.

Step 2. Let k be a number such that $\beta_{n-k} \neq 0$ and $\beta_{n-k-1} = \beta_{n-k-2} = \dots = \beta_{n-i-1} = 0$.

Step 3. Let $l := n - k$ and compute S_{k-1}^{l+1} .

Step 4. Compute

$$A^D = \frac{(-1)^{l+1}}{\beta_l^{l+1}} A^l S_{k-1}^{l+1} \quad (26)$$

Example 4. Using the above algorithm compute E^D for (24).

$$\begin{aligned} \text{Step 1. } \beta_2 &= -\text{tr } E = -\frac{3}{2}, \quad S_1 = ES_0 + I_n \beta_2 = E + I_3 \left(-\frac{3}{2}\right) = \begin{bmatrix} -1 & 0 & 0 \\ 0 & -\frac{1}{2} & 0 \\ -\frac{1}{2} & 0 & -\frac{3}{2} \end{bmatrix}, \\ \beta_1 &= -\frac{1}{2} \text{tr } [ES_1] = \frac{1}{2}, \quad S_2 = ES_1 + I_n \beta_1 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix}, \\ \beta_0 &= -\frac{1}{3} \text{tr } [ES_2] = 0. \end{aligned}$$

Step 2. $k = 2$.

$$\text{Step 3. } l = n - k = 1 \quad \text{and} \quad S_{k-1}^{l+1} = S_1^2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{4} & 0 \\ \frac{5}{4} & 0 & \frac{9}{4} \end{bmatrix} \quad (27)$$

Step 4. Using (26) and (27) we obtain

$$\begin{aligned} E^D &= \frac{(-1)^{l+1}}{\beta_1^{l+1}} E^l S_{k-1}^{l+1} = \frac{(-1)^2}{\beta_1^2} ES_1^2 \\ &= \frac{1}{\left(\frac{1}{2}\right)^2} \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 \\ -\frac{1}{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{4} & 0 \\ \frac{5}{4} & 0 & \frac{9}{4} \end{bmatrix} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & 0 & 0 \end{bmatrix} \end{aligned}$$

If (2) holds, then there exists a scalar $c \in \mathbb{C}$ such that

$$\det [Ec - A] \neq 0$$

and we may find

$$\bar{A} := [Ec - A]^{-1} E \quad \text{and} \quad \bar{A} := [Ec - A]^{-1} A \quad (28)$$

It is easy to show that $\bar{E}\bar{A} = \bar{A}\bar{E}$ and $\ker \bar{A} \cap \ker \bar{E} = \{0\}$ (Campbell *et al.*, 1976; Kaczorek, 1993).

Premultiplying (1) by $[Ec - A]^{-1}$ and using (28) we obtain

$$\bar{E}x_{i+1} = \bar{A}x_i + \bar{B}u_i, \quad i = 0, 1, \dots \tag{29}$$

where

$$\bar{B} := [Ec - A]^{-1}B \tag{30}$$

The solution x_i to (29) (and also to (1)) is given by (Campbell *et al.*, 1976; Kaczorek, 1993)

$$\begin{aligned} x_i = & (\bar{E}^D \bar{A})^i \bar{E}^D \bar{E} x_0 + \sum_{k=0}^{i-1} \bar{E}^D (\bar{E}^D \bar{A})^{i-k-1} \bar{B} u_k \\ & + (\bar{E} \bar{E}^D - I_n) \sum_{k=0}^{q-1} (\bar{E} \bar{A})^k \bar{A}^D \bar{B} u_{i+k} \end{aligned} \tag{31}$$

where q is the index of $\bar{E}(E)$.

The set of admissible initial conditions x_0 is given by

$$Sx_0 := \text{Im}[H_0, H_1, \dots, H_q] \tag{32}$$

where

$$H_k := \begin{cases} (I_n - \bar{E} \bar{E}^D)(\bar{E} \bar{A})^k \bar{A}^D \bar{B} & \text{for } k = 0, 1, \dots, q-1 \\ \bar{E} \bar{E}^D & \text{for } k = q \end{cases} \tag{33}$$

Example 5. Find the solution x_i to (1) with (19). Choosing $c = 2$ and using (28) and (30) we obtain

$$[Ec - A] = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}, \quad [Ec - A]^{-1} = \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 1 & 0 \\ -\frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix}$$

and

$$\bar{E} := [Ec - A]^{-1}E = \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 \\ -\frac{1}{2} & 0 & 0 \end{bmatrix} \quad \cdot$$

$$\bar{A} = [Ec - A]^{-1}A = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & -1 \end{bmatrix}, \quad \bar{B} = [Ec - A]^{-1}B = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

Taking into account the result of Example 3 and that

$$\overline{A}^D = \overline{A} \overline{A}^D \overline{B} = \begin{bmatrix} 0 \\ 0 \\ -1 \end{bmatrix}, \quad \overline{E}^D \overline{A} = \overline{E} \overline{A}^D = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\overline{E} \overline{E}^D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 0 \end{bmatrix}$$

from (31) we obtain

$$\begin{aligned} x_i &= \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}^i \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 0 \end{bmatrix} x_0 \\ &+ \sum_{k=0}^{i-1} \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}^{i-k-1} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} u_k \\ &+ \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ -1 & 0 & -1 \end{bmatrix} \left(\sum_{k=0}^i \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}^k \begin{bmatrix} 0 \\ 0 \\ -1 \end{bmatrix} u_{i+k} \right) \\ &= \begin{bmatrix} 2u_{i-1} \\ a \\ u_i - 2u_{i-1} \end{bmatrix} \end{aligned} \quad (34)$$

where a is any real number (a is the second component of x_0).

Through (22) and (33) the set of admissible initial conditions is given by

$$S_{x_0} = \text{Im}[H_0, H_1]$$

$$= \text{Im} \left[(I_n - \overline{E} \overline{E}^D) \overline{A}^D \overline{B}, \overline{E} \overline{E}^D \right] = \text{Im} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & -1 & 0 \end{bmatrix} = \mathbb{R}^3 \quad (35)$$

Note that (34) and (35) are the same as (20) and (21), respectively.

4. Extension of the Dias and Mesquista Method for Discrete-Time Systems

Without loss of generality it is assumed that

$$E = \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix}, \quad A = \begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix}, \quad B = \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} \quad (36)$$

where $r = \text{rank } E$.

Consider the family of (A, E) -invariant subspaces defined by (Armentano, 1984; Dias and Mesquista, 1990; Kaczorek, 1993; Wonham, 1979)

$$\mathcal{Z} := \left\{ \mathcal{Z} : A\mathcal{Z} \subset E\mathcal{Z}; \mathcal{Z} \subset \mathbb{R}^n \right\} \quad (37)$$

The supremal element $\mathcal{Z}^* := \sup \mathcal{Z}$ can be computed in a finite number of steps $\mu \leq n$ by the algorithm

$$\mathcal{Z}_k = A^{-1}(E\mathcal{Z}_{k-1}), \quad k = 1, \dots, n \quad \mathcal{Z}_0 = \mathbb{R}^n$$

$$\mathcal{Z}^* = \mathcal{Z}_\mu = \mathcal{Z}_{\mu+1}$$

$$\text{where } A^{-1}\mathcal{Z}_k := \{x \in \mathbb{R}^n : Ax \in \mathcal{Z}_k\}. \quad (38)$$

Let the columns of V form a basis for \mathcal{Z}^* , $\mathcal{Z}^* = \text{Im}V$ and

$$x_i = Vz_i + \sum_{k=-i}^p L_k u_{i+k} \quad (39)$$

where z_i, L_k and p will be defined later. Substituting (39) and (36) into (1) we obtain

$$\begin{aligned} V_1 z_{i+1} &= [A_1 A_2]Vz_i + \left([A_1 A_2]L_0 + B_1 - [I_r 0]L_{-1} \right) u_i \\ &\quad + \sum_{k=1}^p \left([A_1 A_2]L_k - [I_r 0]L_{k-1} \right) u_{i+k} - [I_r 0]L_p u_{i+p+1} \\ &\quad + \sum_{k=1}^{i-1} \left([A_1 A_2]L_{-k} - [I_r 0]L_{-k-1} \right) u_{i-k} + [A_1 A_2]L_{-i} u_0 \end{aligned} \quad (40)$$

$$\begin{aligned} 0 &= [A_3 A_4]Vz_i + \left([A_3 A_4]L_0 + B_2 \right) u_i \\ &\quad + \sum_{k=1}^p \left([A_3 A_4] \right) L_k u_{i+k} + \sum_{k=1}^i \left([A_3 A_4]L_{-k} u_{i-k} \right) \end{aligned} \quad (41)$$

where

$$V := \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}, \quad V_1 \in \mathbb{R}^{r \times l}, \quad l = \dim \mathcal{Z}^* \quad (42)$$

It can be easily shown that $l = \deg \det[Ez - A] \leq r$.

$$\text{Define } \bar{Q} := \begin{bmatrix} \bar{Q}_1 \\ \bar{Q}_1 \end{bmatrix}, \quad \bar{Q}_1 \in \mathbb{R}^{l \times r}, \quad \bar{Q}_2 \in \mathbb{R}^{(r-l) \times r} \text{ such that } \det \bar{Q} \neq 0$$

and

$$\bar{Q}V_1 = \begin{bmatrix} I_l \\ 0 \end{bmatrix} \quad (43)$$

Premultiplying (40) by \bar{Q} and using (43) we obtain

$$z_{i+1} = \bar{A}_1 z_i, \quad \bar{A}_1 := \bar{Q}_1[A_1 \ A_2]V \quad (44)$$

if

$$\bar{Q}_2[A_1 \ A_2]V z_i = 0 \quad \text{for an arbitrary } z_i \quad (45)$$

$$[A_1 \ A_2]L_0 + B_1 - [I_r \ 0]L_{-1} = 0 \quad (46)$$

$$[A_1 \ A_2]L_k - [I_r \ 0]L_{k-1} = 0 \quad \text{for } k = 1, \dots, p \quad (47)$$

$$[A_1 \ A_2]L_{-k} - [I_r \ 0]L_{-k-1} = 0 \quad \text{for } k = 1, \dots, i-1 \quad (48)$$

$$[I_r \ 0]L_p = 0, \quad [A_1 \ A_2]L_{-i} = 0 \quad (49)$$

since $\det \bar{Q} \neq 0$.

Note that (41) is satisfied if

$$[A_3 \ A_4]V z_i = 0 \quad \text{for an arbitrary } z_i \quad (50)$$

$$[A_3 \ A_4]L_0 + B_2 = 0 \quad (51)$$

$$[A_3 \ A_4]L_k = 0 \quad \text{for } k = 1, \dots, p \quad (52)$$

$$[A_3 \ A_4]L_{-k} = 0 \quad \text{for } k = 1, \dots, i \quad (53)$$

In a similar way as in (Dias and Mesquista, 1990) it can be shown that

$$\begin{bmatrix} Q_2[A_1 \ A_2] \\ A_3 A_4 \end{bmatrix} V = 0 \quad (54)$$

since $Vz_i \in \mathcal{Z}^*$. Therefore (45) and (50) are satisfied for an arbitrary z_i .

Two cases will be considered for $\det A \neq 0$ and $\det A = 0$.

Case $\det A \neq 0$.

It will be shown that if $\det A \neq 0$, then it can be assumed in (39) that $k = -i = 0$. Assuming $k = -i = 0$ and $L_{-1} = 0$ from (46) and (51) we obtain

$$L_0 = -A^{-1}B \tag{55}$$

Next from (47), (49), and (52) we have

$$L_k = A^{-1}EL_{k-1}, \quad k = 1, \dots, p \tag{56}$$

and

$$EL_p = 0 \tag{57}$$

Using (55) and (56) we may compute L_0, L_1, \dots, L_p and the procedure stops when (57) is satisfied.

The solution z_i of (44) has the form

$$z_i = \overline{A}_1^i z_0 \tag{58}$$

Substitution of (58) into (39) for $k = -i = 0$ yields

$$x_i = V\overline{A}_1^i z_0 + \sum_{k=0}^p L_k u_{i+k} \tag{59}$$

Therefore if $\det A \neq 0$, then the solution x_i to (1) with (36) is given by (59).

Remark. Note that if $\text{rank } A = \text{rank}[A, B]$ and $\text{rank}[A_1, A_2] = r$, it can be also assumed that $k = -i = 0$ in (39), since the equations $AL_0 = -B$ and $AL_k = EL_{k-1}$ have solutions L_0 and L_k for $k = 1, \dots, p$.

Example 6. Find the solution x_i to (1) with (17). Using (38) we obtain

$$Z_1 = \overline{A}^{-1}(E\mathbb{R}^3) = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}^{-1} \left(\text{Im} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \right) = \text{Im} \begin{bmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$Z_2 = \overline{A}^{-1}(EZ_1) = \text{Im} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad Z_3 = \overline{A}^{-1}(EZ_2) = \text{Im} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

and

$$Z^* = \text{Im} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, V = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, V_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \bar{Q} = \begin{bmatrix} \bar{Q}_1 \\ \bar{Q}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

From (55), (56), and (57) we have

$$L_0 = -\bar{A}^{-1}B = - \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix}$$

$$L_1 = \bar{A}^{-1}EL_0 = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}^{-1} \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ -1 \end{bmatrix}$$

and $EL_1 = 0, p = 1$.

Using (59) we obtain $\bar{A}_1 = Q_1[A_1 A_2]V = [0 \ 1] \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = 1$ and

$$x_i = V\bar{A}_1^i z_0 + \sum_{k=0}^1 L_k u_{i+k} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} z_0 + \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix} u_i + \begin{bmatrix} 0 \\ 0 \\ -1 \end{bmatrix} u_{i+1} = \begin{bmatrix} -u_i \\ z_0 \\ -u_{i+1} \end{bmatrix} \quad (60)$$

Note that (60) agrees with (18).

Case $\det A = 0$.

Note that if (2) holds, then for (36) the matrix $[A_3 A_4]$ has a full row rank and from (51) we have

$$L_0 = - \begin{bmatrix} A_3^T \\ A_4^T \end{bmatrix} [A_3 A_3^T + A_4 A_4^T]^{-1} B_2 + \left(I_n - \begin{bmatrix} A_3^T \\ A_4^T \end{bmatrix} [A_3 A_3^T + A_4 A_4^T]^{-1} [A_3 A_4] \right) K_1 \quad (61)$$

where K_1 is an arbitrary matrix.

The matrix K_1 is chosen so that $[I_r \ 0]L_0 = 0$. If there exists K_1 such that $[I_r \ 0]L_0 = 0$, then $p = 0$. Next, from (46) we may compute

$$L_{-1} = \begin{bmatrix} I_r \\ 0 \end{bmatrix} ([A_1 A_2]L_0 + B_1) + \begin{bmatrix} 0 & 0 \\ 0 & I_{n-r} \end{bmatrix} K_2 \quad (62)$$

The matrix K_2 is chosen so that $[A_1 A_2]L_{-1} = 0$. If there exists K_2 such that $[A_1 A_2]L_{-1} = 0$, then $i = 1$.

Combining the equations

$$\begin{aligned} \bar{Q}_2([A_1 A_2]L_k - [I_r 0]L_{k-1}) &= 0, & k = 1, \dots, p \\ \bar{Q}_2([A_1 A_2]L_{-k} - [I_r 0]L_{-k-1}) &= 0, & k = 1, \dots, i-1 \end{aligned}$$

with (52) and (53), respectively we obtain

$$HL_k = \bar{E}L_{k-1}, \quad k = 1, \dots, p \tag{63}$$

and

$$HL_{-k} = \bar{E}L_{-k-1}, \quad k = 1, \dots, i-1 \tag{64}$$

where

$$\begin{aligned} H &:= \begin{bmatrix} \bar{Q}_2[A_1 A_2] \\ A_3 A_4 \end{bmatrix} = \begin{bmatrix} \bar{Q}_2 & 0 \\ 0 & I_{n-r} \end{bmatrix} A \\ \bar{E} &:= \begin{bmatrix} \bar{Q}_2 & 0 \\ 0 & I_{n-r} \end{bmatrix} E = \begin{bmatrix} \bar{Q}_2[I_r 0] \\ 0 & 0 \end{bmatrix} \end{aligned} \tag{65}$$

In a similar way as in (Dias and Mesquista, 1990) it can be shown that matrix (65) has a full row rank. Therefore from (63),(64) we have

$$L_k = H_R \bar{E}L_{k-1}, \quad k = 1, \dots, p \tag{66}$$

and

$$L_{-k} = H_R \bar{E}L_{-k-1}, \quad k = 1, \dots, i-1 \tag{67}$$

where

$$H_R := H^T [HH^T]^{-1}$$

The Moore-Penrose generalized inverse H^g can also be used (Dias and Mesquista, 1990). Using (66), (67) we may compute $L_k(L_{-k})$ for $k = 1, \dots, p$ ($k = 1, \dots, i-1$). The algorithm stops when (49) is satisfied.

Example 7. Find the solution x_i to (1) with(19).

Using (38) we obtain

$$Z_1 = \bar{A}^{-1} (E \mathbb{R}^3) = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & -1 \end{bmatrix}^{-1} \left(\text{Im} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \right) = \text{Im} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$Z_2 = \bar{A}^{-1}(EZ_1) = \text{Im} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \end{bmatrix}$$

and

$$Z^* = \text{Im} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \end{bmatrix}, \quad V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \end{bmatrix}, \quad V_1 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad \bar{Q} = \bar{Q}_1 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

From (61) we have

$$L_0 = \frac{1}{2} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + \begin{bmatrix} 0.5 & 0 & -0.5 \\ 0 & 1 & 0 \\ -0.5 & 0 & 0.5 \end{bmatrix} \begin{bmatrix} k_{11} \\ k_{12} \\ k_{13} \end{bmatrix} = \begin{bmatrix} 0.5(1 + k_{11} - k_{13}) \\ k_{12} \\ 0.5(1 - k_{11} + k_{13}) \end{bmatrix}$$

and

$$[I_r \ 0]L_0 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0.5(1 + k_1 - k_3) \\ k_2 \\ 0.5(1 - k_1 + k_3) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

gives $k_1 = -1$, $k_2 = k_3 = 0$. Hence

$$L_0 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad \text{and} \quad p = 0$$

Using (62) we obtain

$$L_{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \left(\begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right) + \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} k_{21} \\ k_{22} \\ k_{23} \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ k_{23} \end{bmatrix}$$

and

$$[A_1 \ A_2]L_{-1} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 0 \\ k_{23} \end{bmatrix} = 0$$

gives $k_{23} = -2$.

Hence

$$L_{-1} = \begin{bmatrix} 2 & & \\ & 0 & \\ & & -2 \end{bmatrix} \quad \text{and} \quad t = 1$$

Thus from (39) we obtain

$$x_i = V\bar{A}_1^i z_0 + \sum_{k=-1}^0 L_k u_{i+k} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \end{bmatrix}^i \begin{bmatrix} z_{10} \\ 0 & 1 \\ z_{20} \end{bmatrix} + \begin{bmatrix} 2 & & \\ & u_{i-1} & \\ & & 0 \end{bmatrix} + \begin{bmatrix} 0 & & \\ & u_i & \\ & & 1 \end{bmatrix} = \begin{bmatrix} 2u_{i-1} & & \\ & z_{20} & \\ u_i - 2u_{i-1} & & \end{bmatrix} \tag{68}$$

Note that (68) agrees with (20).

5. Method of Expansion in a Series

Let $X(z)$ be the z -transform of x_i defined by

$$X(z) := \sum_{i=0}^{\infty} x_i z^{-i} \tag{69}$$

Using the z -transformation eqn. (1) can be written in the form

$$[Ez - A]X(z) = zEx_0 + BU(z) \tag{70}$$

where $U(z)$ is the z -transform of u_i .

From (70) we have

$$X(z) = [Ez - A]^{-1} zEx_0 + [Ez - A]^{-1} BU(z)z \tag{71}$$

Note that if the degree of $\det[Ez - A]$ is less than rank E , then the matrix $[Ez - A]^{-1}$ may be improper and it can be decomposed into a polynomial part $P(z)$ and a strictly proper part $T_{sp}(z)$

$$[Ez - A]^{-1} = P(z) + T_{sp}(z) \tag{72}$$

where

$$P(z) = \sum_{i=0}^p P_i z^i \tag{73}$$

$$T_{sp}(z) = \sum_{i=0}^{\infty} T_i z^{-(i+1)} \tag{74}$$

$$p \leq \text{rank } E - \text{deg det}[Ez - A]$$

Substituting (72)–(74) into (71) we obtain

$$X(z) = \sum_{i=0}^p P_i E x_0 z^{i+1} + \sum_{i=0}^{\infty} T_i E x_0 z^{-1} + \sum_{i=0}^p P_i B z^i U(z) + \sum_{i=0}^{\infty} T_i B z^{-(i+1)} U(z) \tag{75}$$

The inverse z -transformation of (75) yields

$$x_i = T_i E x_0 + \sum_{k=0}^p P_k B u_{i+k} + \sum_{k=0}^{i-1} T_k B u_{i-k-1} \quad i \geq 0 \tag{76}$$

Note that all terms with positive powers z of (75) have been neglected since we are interested in the solution for $i \geq 0$. Therefore the solution x_i to (1) is given by (76).

Example 8. Find the solution x_i to eqn. (1) with (19).

In this case

$$\begin{aligned} [Ez - A]^{-1} &= \begin{bmatrix} z-1 & 0 & -1 \\ 0 & z-1 & 0 \\ 1 & 0 & 1 \end{bmatrix}^{-1} \\ &= \begin{bmatrix} z^{-1} & 0 & z^{-1} \\ 0 & (z-1)^{-1} & 0 \\ -z^{-1} & 0 & 1-z^{-1} \end{bmatrix} = P_0 + T_{sp}(z) \end{aligned}$$

where

$$P_0 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad T_{sp}(z) = \begin{bmatrix} z^{-1} & 0 & z^{-1} \\ 0 & (z-1)^{-1} & 0 \\ -z^{-1} & 0 & -z^{-1} \end{bmatrix} = \sum_{k=0}^{\infty} T_k z^{-(k+1)}$$

$$T_0 = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}, \quad T_k = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \text{for } k = 1, 2, \dots$$

Using (76) we obtain

$$\begin{aligned}
 x_i &= T_i E x_0 + \sum_{k=0}^0 P_k B u_{k+1} + \sum_{k=0}^{i-1} T_k B u_{i-k-1} \\
 &= \begin{bmatrix} 0 \\ x_{20} \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u_i + \begin{bmatrix} 2 \\ 0 \\ -2 \end{bmatrix} u_{i-1} = \begin{bmatrix} 2u_{i-1} \\ x_{20} \\ u_i - 2u_{i-1} \end{bmatrix}
 \end{aligned}$$

This result agrees with the previous ones.

6. Comparison of Methods

Solution (15) may be written in the form

$$x_i = Q \begin{bmatrix} A_1^i Q_1 \\ 0 \end{bmatrix} x_0 + \sum_{k=0}^{i-1} Q \begin{bmatrix} A_1^{i-k-1} B_1 \\ 0 \end{bmatrix} u_k - \sum_{k=0}^{q-1} Q \begin{bmatrix} 0 \\ N^k B_2 \end{bmatrix} u_{i+k} \tag{77}$$

A comparison of (77) with (31) yields

$$Q \begin{bmatrix} A_1^i Q_1 \\ 0 \end{bmatrix} = (\overline{E}^D \overline{A})^k \overline{E}^D \overline{E}, \quad k = 0, 1, \dots \tag{78}$$

$$Q \begin{bmatrix} A_1^k B_1 \\ 0 \end{bmatrix} = \overline{E}^D (\overline{E}^D \overline{A})^k \overline{B}, \quad k = 0, 1, \dots \tag{79}$$

$$Q \begin{bmatrix} 0 \\ N^k B_2 \end{bmatrix} = (I - \overline{E} \overline{E}^D) (\overline{E} \overline{A})^k \overline{A}^D \overline{B}, \quad k = 0, 1, \dots \tag{80}$$

From (78)-(80) for $k = 0$ we have

$$\begin{aligned}
 Q \begin{bmatrix} Q_1 \\ 0 \end{bmatrix} &= \overline{E}^D \overline{E}, & Q \begin{bmatrix} B_1 \\ 0 \end{bmatrix} &= \overline{E}^D \overline{B} \\
 Q \begin{bmatrix} 0 \\ B_2 \end{bmatrix} &= (I - \overline{E} \overline{E}^D) \overline{A}^D \overline{B}
 \end{aligned} \tag{81}$$

Knowing $\overline{E}, \overline{E}^D, \overline{B}, \overline{A}^D$ and Q we may find from (81) Q_1, B_1 and B_2 .

From a comparison of (77) and (39) with (58) we obtain

$$Q \begin{bmatrix} A_1^k Q_1 \\ 0 \end{bmatrix} x_0 = V \overline{A}_1^k z_0, \quad k = 0, 1, \dots \tag{82}$$

$$Q \begin{bmatrix} A_1^{k-1} Q_1 \\ 0 \end{bmatrix} = L_k, \quad k = 0, 1, \dots, -i \tag{83}$$

$$Q \begin{bmatrix} 0 \\ N^k B_2 \end{bmatrix} = L_k, \quad k = 0, 1, \dots, p = q - 1 \tag{84}$$

From (82) for $k = 0$ we have

$$\text{Im} Q \begin{bmatrix} Q_1 \\ 0 \end{bmatrix} = \text{Im} V \tag{85}$$

A comparison of (31) and (39) with (58) yields

$$(\overline{E^D A})^k \overline{E^D} \overline{E} x_0 = V \overline{A_1^k} z_0, \quad k = 0, 1, \dots \tag{86}$$

$$\overline{E^D} (\overline{E^D A})^{k-1} \overline{B} = L_k, \quad k = -1, \dots, -i \tag{87}$$

$$(I_n - \overline{E E^D}) (\overline{E E^D})^k \overline{A^D} \overline{B} = L_k, \quad k = 0, 1, \dots, p = q - 1 \tag{88}$$

From a comparison of (77) and (76) it follows that

$$Q \begin{bmatrix} A_1^k Q_1 \\ 0 \end{bmatrix} = T_k E, \quad k = 0, 1, \dots \tag{89}$$

$$Q \begin{bmatrix} A_1^k B_1 \\ 0 \end{bmatrix} = T_k B, \quad k = -1, \dots, -i \tag{90}$$

$$Q \begin{bmatrix} 0 \\ N^k B_2 \end{bmatrix} = -P_k B, \quad k = 0, 1, \dots, p = q - 1 \tag{91}$$

Similarly, from a comparison of (76) and (39) with (58) we obtain

$$T_k E x_0 = V \overline{A_1^k} z_0 \tag{92}$$

$$L_k = T_k B, \quad k = -1, \dots, -i \tag{93}$$

$$L_k = -P_k B, \quad k = 0, 1, \dots, p = q - 1 \tag{94}$$

Note that from the above comparisons we may find new formulae for the solution x_i to (1). For example, using (92) and (76) we may obtain the solution in the form

$$x_i = V \overline{A_1^i} z_0 + \sum_{k=0}^{q-1} P_k B u_{k+1} + \sum_{k=0}^{i-1} T_k B u_{i-k-1} \tag{95}$$

where $V, \overline{A_1}, P_k$ and T_k are defined by (38), (44), and (73)–(74), respectively.

7. Concluding Remarks

An extension for regular discrete-time linear systems of the Dias and Mesquista method (Dias and Mesquista, 1990) has been given. In (Kaczorek, 1995) it has been shown that the Dias and Mesquista method should be modified by adding an additional term containing an integral for regular continuous-time linear systems. A method based on the expansion in a series of the inverse matrix $[Ez - A]^{-1}$ has been proposed. These two new methods have been compared with two well-known methods based on the Weierstrass-Kronecker decomposition and on the Drazin inverse. The methods have been illustrated by numerical examples with a non-singular and singular matrix A . Relationships between the coefficient matrices of the four methods have been established. Moreover, a new mixed method of finding the solution (95) to eqn. (1) has been presented.

References

- Aplevich J.D. (1991): *Implicit Linear Systems*. — Lecture Notes in Control and Information Sciences, Berlin: Springer-Verlag.
- Armentano V.A. (1979): *Eigenvalue placement for generalized linear systems*. — Systems and Control Letters, v.17, No.4, pp.509–522.
- Armentano V.A. (1984): *The pencil $(sE - A)$ and controllability-observability for generalized linear systems: A geometric approach*. — Proc. 23rd IEEE Conf. Dec. and Control, Las Vegas, pp.1507–1510.
- Bernhard P. (1982): *On singular implicit linear dynamical systems*. — SIAM J. Control and Optim., v.20, No.5, pp.612–633.
- Campbell S.L. and Meyer C.D. (1979): *Generalized Inverses of Linear Transformations*. — San Francisco: Pitman.
- Campbell S.L., Meyer C.D. Jr and Rose N.J. (1976): *Applications of the Drazin inverse to linear systems of differential equations with singular coefficients*. — SIAM J. Appl. Math., v.10, No.3, pp.542–551.
- Campbell S.L. (1980): *Singular Systems of Differential Equations*. — San Francisco: Pitman.
- Cobb D. (1981): *Feedback and pole-placement in descriptor variable systems*. — Int. J. Control, v.33, No.6, pp.1135–1146.
- Cobb D. (1982): *On the solution of linear differential equations with singular coefficients*. — J. Diff. Eq., v.46, No.3, pp.310–323.
- Cobb D. (1983): *Descriptor variable systems and optimal state regulation*. — IEEE Trans. Automat. Contr., v.AC-28, No.5, pp.601–611.
- Cobb D. (1984): *Controllability, observability and duality in singular systems*. — IEEE Trans. Automat. Contr., v.AC-29, No.12, pp.1076–1082.
- Dai L. (1989): *Singular Control Systems*. — Lecture Notes in Control and Information Sciences, Berlin: Springer-Verlag.
- Dias R.J. and Mesquista A. (1990): *A closed form solution for regular descriptor systems using the Moore-Penrose generalized inverse*. — Automatica, v.26, No.2, 417–420.

- Dziurla B. and Newcomb R. (1979): *The Drazin inverse and semi-state equations* — Proc. 4th Int. Symp. *Math. Theory of Networks and Systems* Delft, The Netherlands, pp.283–289.
- Fettweis A. (1969): *On the algebraic derivation of state equations*. — IEEE Trans. Circuit Theory, v.CT-16, No.2, pp.171–175.
- Gantmacher F.R. (1959): *Theory of Matrices*. — New York, Chelsea: Pub. Co., v.1 and v.2.
- Kaczorek T. (1993): *Linear Control Systems*. — New York: Research Studies Press and J. Wiley.
- Kaczorek T. (1995): *A closed form solution for regular descriptor systems*. — Proc. SPETO'96 (in press).
- Karcanias N. and Hayton G.E. (1981): *State-space and transfer function invariant infinite zeros. A unified approach*. — Charlottesville, VA, Proc. JACC, paper TA-4C.
- Karcanias N. and Hayton G.E. (1981): *Generalized autonomous dynamical systems. Algebraic duality and geometric theory*. — Proc. IFAC VIII Triennial World Congress, Kyoto, Japan.
- Kronecker L. (1980): *Algebraische Reduction der Schaaren Bilinearer Formen*. — S.-B. Akad. Berlin, pp.763–776.
- Lewis F.L. (1986): *A survey of linear singular systems*.— Circuits Systems Signal Process, v.5, No.1, pp.1–36.
- Lewis F.L. (1984): *Descriptor systems: decomposition into forward and backward subsystems*.— IEEE Trans. Automat. Contr., v.AC-29, No.2, pp.167–170.
- Lewis F.L. and Ozcaldiran K. (1983): *The relative eigenstructure problem and descriptor systems*.— Proc. SIAM National Meeting, Denver, CO, pp.83–96.
- Lewis F.L. and Ozcaldiran K. (1985): *On the eigenstructure assignment of singular systems*. — Proc. 24th IEEE Conf. Dec. and Control, Ft. Lauderdale, FL, pp.179–182.
- Luenberger D.G. (1977): *Dynamical equations in descriptor form*.— IEEE Trans. Automat. Contr., v.AC-22, No.3, pp.312–321.
- Luenberger D.G. (1978): *Time-invariant descriptor systems*. — Automatica, v.14, pp.473–480.
- Mertzios B.G. (1984): *Leverrier's algorithm for singular systems*.— IEEE Trans. Automat. Contr., v.AC-29, No.7, pp.652–653.
- Molinari B.P. (1979): *Structural invariants of linear multivariable systems*. — Int. J. Contr., v.28, No.2, pp.525–535.
- Sincovec R.F., Erisman A.M., Yip E.L. and Epton M.A (1981): *Analysis of descriptor systems using numerical algorithms*.— IEEE Trans. Automat. Contr., v.AC-26, No.1, pp.139–147.
- Van Dooren P. (1979): *The computation of Kronecker's canonical form of a singular pencil*. — Linear Algebra and Its Applications, v.27, pp.103–140.
- Verghese G.P., Van Dooren P. and Kailath T. (1979): *Properties of the system matrix of a generalized state-space system*. — Int. J. Contr., v.30, No.2, pp.235–243.
- Wonham W.M. (1979): *Linear Multivariable Control: A Geometric Approach*.— New York: Springer-Verlag, (2nd ed.).
- Yip E.L. and Sincovec R.F. (1981): *Solvability, controllability and observability of continuous descriptor systems*. — IEEE Trans. Automat. Contr., v.AC-26, No.3, pp.702–707.

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