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# On a Least-Squares Collocation Method for Linear Differential-Algebraic Equations

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**Summary.** The aim of this note is to extend some results on least-squares collocation methods and to prove the convergence of a least-squares collocation method applied to linear differential-algebraic equations. Some numerical examples are presented.

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#### 1. Introduction

In recent time, great interest has been spent on differential algebraic equations (DAE's). Such equations arise in many fields of applications. Whereas in former times equations describing electrical networks were mainly stimulating the investigation, nowadays the development of numerical methods is focused on equations describing constraint mechanical motions or kinetic reactions subject to balance invariants and on singular optimal control problems (cf. [2, 3, 9, 17]. In the present paper we deal with linear DAE's only, that is, with equations

$$A(t) x'(t) + B(t) x(t) = f(t), \quad t \in (0, 1)$$
 (1.1)

$$D_0 x(0) + D_1 x(1) = \gamma (1.2)$$

where  $x(t) \in \mathbb{R}^n$ ,  $D_0$ ,  $D_1 \in \mathbf{B}(\mathbb{R}^n, \mathbb{R}^r)$ . Let A(t),  $B(t) \in \mathbf{B}(\mathbb{R}^n)$  be sufficiently smooth. Here, for Banach spaces X and Y,  $\mathbf{B}(X, Y)$  denotes the Banach space of all bounded linear operators defined on X and mapping into Y. The analytic solution of (1.1), if A and B are constant, is reasonably well understood (cf. [2, 3, 8]). It turns out that so-called higher index DAE's include differentiation problems. This behaviour is preserved in the case of time-dependent coefficients (cf. [18, 19]). Therefore, they are essentially ill-posed [13, 16]. In this context it is not very surprising that usual finite-difference methods become unstable when applied to higher index equations [13]. To solve them numerically we should use some regularization procedure. Least-squares collocation methods are known to be regularizing algorithms [5-7, 15]. For (1.1)-(1.2) this method can be formulated as follows: For  $v \in \mathbb{N}$ , let  $\pi_v \subset [0, 1]$  be a finite set of collocation

points. A function  $x_v: [0, 1] \to \mathbb{R}^n$  is called least-squares collocation solution of (1.1)–(1.2) on  $\pi_v$  if  $x_v$  satisfies (1.2) and

$$A(t) x_{\nu}'(t) + B(t) x_{\nu}(t) = f(t), \quad t \in \pi_{\nu}$$
 (1.3)

and  $x_v$  minimizes some scalar product norm among all solutions of (1.3)–(1.2). The feasibility and convergence of this method will be proved. In Sect. 2 we develop the functional-analytic background. An essential tool of our proof is the theory of reproducing kernel Hilbert spaces (RKHS) [1]. Section 3 contains the application of these results to DAE's. Our construction follows some ideas of [20, 21]. Finally, some numerical are presented in Sect. 4.

# 2. Theoretical Background

Let X be a Hilbert space with the scalar product  $(\cdot, \cdot)$ . For  $l \in X^*$ , denote  $\langle l, x \rangle = l(x)$  for  $x \in X$ . For a linear operator T, we denote the kernel and the range by  $\mathbf{N}(T)$  and  $\mathbf{R}(T)$ , respectively. On X let the linear mappings  $T_j$  be defined:

$$T_j \in L(X, (I_j, \mathbb{R})), \quad j = 1, \ldots, m$$

with  $\phi \neq I_j \subset \mathbb{R}$ ,  $(I_j, \mathbb{R}) = \{f \mid f: I_j \to \mathbb{R}\}$ . Denote Dirac's  $\delta$ -functional by  $\delta_t$ :

$$\langle \delta_t, x \rangle = x(t), \quad t \in \mathbb{R}.$$

Throughout this section we assume the following supposition to be fulfilled.

Supposition. For j=1,...,m and  $t \in I_j$ ,  $\delta_t \circ T_j$  is a continuous linear functional on X, i.e.,

$$\delta_t \circ T_j \in X^*, \quad j = 1, \dots, m, \ t \in I_j. \tag{2.1}$$

According to the Riesz representation theorem there are  $\eta_t^j \in X$  such that

$$\langle \delta_t \circ T_j, x \rangle = T_j x(t) = (\eta_t^j, x), \quad x \in X$$
 (2.2)

for all  $t \in I_j$ , j = 1, ..., m. Denote  $T = (T_1, ..., T_m)$ . The equation to be solved is denoted by

$$Tx = f (2.3)$$

for some vector function  $f = (f_1, ..., f_m)$  and  $f_j \in (I_j, \mathbb{R})$ .

**Lemma 2.1.** For all j,  $N(T_j) \subset X$  is closed. Moreover,  $N(T) = N(T_1) \cap ... \cap N(T_m)$ .

*Proof.* Let  $1 \le j \le m$ . Let  $\{x_l\} \subset \mathbf{N}(T_j)$  with  $x_l \to x \in X$ . Hence,  $0 = T_j x_l(t)$   $= \langle \delta_t \circ T_j, x_l \rangle \to \langle \delta_t \circ T_j, x \rangle = T_j x(t)$  for  $t \in I_j$ . Thus,  $x \in \mathbf{N}(T_j)$ . The identity  $\mathbf{N}(T)$   $= \mathbf{N}(T_1) \cap \ldots \cap \mathbf{N}(T_m)$  is obvious.  $\square$ 

**Corollary.** The Moore-Penrose generalized inverses  $T_j^+$  and  $T^+$  of  $T_j$  and  $T_j^+$  respectively, are well-defined on  $\mathbf{R}(T_j)$  and  $\mathbf{R}(T)$ , respectively.

For a definition of the Moore-Penrose generalized inverse see, e.g., [14]. The next lemma characterizes  $\mathbf{R}(T_i)$ .

**Lemma 2.2.** Let  $Y_i := \mathbb{R}(T_i) \subset (I_i, \mathbb{R}), j = 1, ..., m$ . Then:

(i) Every space Y<sub>j</sub> is a RKHS with the reproducing kernel

$$Q^{j}(t, t') = (\eta_{t}^{j}, \eta_{t'}^{j}), \quad t, t' \in I_{j}.$$
(2.4)

The associated scalar product is given by

$$[f,g]_{i} = (T_{i}^{+}f, T_{i}^{+}g), \quad f \in \mathbf{R}(T_{i}).$$
 (2.5)

(ii) Let  $Y = Y_1 \times ... \times Y_m$ . Then  $T \in \mathbf{B}(X, Y)$ , and  $\mathbf{R}(T) \subset Y$  is closed.

*Proof.* (i) Since  $T_j^+$  is linear, (2.5) defines a positive semi-definite bilinear form  $[\cdot,\cdot]_j$ . If  $f \in \mathbf{R}(T_j)$  and  $[f,f]_j = 0$ ,  $(T_j^+f,T_j^+f) = 0$ . Hence,  $T_j^+f = 0$ . Because of  $T_j^+f \in \mathbf{N}(T_j)^\perp$  we have  $T_j^-T_j^+f = f = 0$ . Thus,  $[\cdot,\cdot]_j$  is a scalar product.

Now, let  $\{f_l\}\subset \mathbf{R}(T_j)$  be a Cauchy sequence with respect to  $[\cdot,\cdot]_j$ . Choose  $x_l\in \mathbb{N}(T_j)^\perp$  such that  $Tx_l=f_l$ . Because of  $[f_l-f_k,f_l-f_k]=(x_l-x_k,x_l-x_k),$   $\{x_l\}$  is a Cauchy sequence in  $\mathbf{N}(T_j)^\perp$ . Since  $\mathbf{N}(T_j)$  is closed, there exists an  $x\in \mathbf{N}(T_j)^\perp$  with  $x_l\to x$ . Now,  $[f_l-T_jx,f_l-T_jx]_j=(x_l-x,x_l-x)=\|x_l-x\|^2$ . Hence,  $f_l\to T_ix\in \mathbf{R}(T_i)$ .

Next we show that  $Q^j$  is the reproducing kernel of  $Y_j$ . Denote the orthogonal projection of X onto  $\mathbf{N}(T_i)^{\perp}$  by  $P_i$ . Due to (2.2) and (2.4),

$$Q^{j}(t, t') = (\eta_{t}^{j}, \eta_{t'}^{j}) = T_{i} \eta_{t'}^{j}(t).$$

Let  $f \in \mathbf{R}(T_i)$  and set  $x = T_i^+ f \in \mathbf{N}(T_i)^{\perp}$ . Compute

$$[f, Q^{j}(\cdot, t')]_{j} = (T_{j}^{+}f, T_{j}^{+}Q^{j}(\cdot, t')) = (x, T_{j}^{+}T_{j}\eta_{t'}^{j}) = (x, P_{j}\eta_{t'}^{j}) = (P_{j}x, \eta_{t'}^{j}) = (x, \eta_{t'}^{j})$$
$$= T_{j}x(t') = f(t').$$

(ii) For all  $x \in X$ ,

$$[Tx, Tx] = \sum_{j=1}^{m} [T_j x, T_j x]_j = \sum_{j=1}^{m} (T_j^+ T_j x, T_j^+ T_j x) = \sum_{j=1}^{m} (P_j x, P_j x) \leq m(x, x).$$

Hence,  $T \in \mathbf{B}(X, Y)$ . We compute  $T^*$ . Let  $x \in X$  and  $f \in Y$ .  $[Tx, f] = \sum_{j=1}^{m} [T_j x, f_j]_j$   $= \sum_{j=1}^{m} (x, T_j^* f_j) = \left(x, \sum_{j=1}^{m} T_j^* f_j\right)$ , thus,  $T^* f = \sum_{j=1}^{m} T_j^* f_j$ . This gives  $\mathbf{R}(T^*) = \mathbf{R}(T_1^*)$  $+ \dots + \mathbf{R}(T_m^*)$ . Since  $\mathbf{R}(T_j) = Y_j$  is closed  $\mathbf{R}(T_j^*)$  is so. Hence,  $\mathbf{R}(T^*)$  is closed and, consequently,  $\mathbf{R}(T)$ , too.  $\square$ 

Remark. (i) From

$$[T_j x, f_j]_j = (T_j^+ T_j x, T_j^+ f_j) = (P_j x, T_j^+ f_j) = (x, P_j T_j^+ f_j) = (x, P_j T_j^+ f_j) = (x, T_j^+ f_j),$$
  
all  $x \in X$ ,  $f_j \in Y_j$ , it follows that  $T_j^+ = T_j^*$ .

(ii) Using the introduced norms, the problem Tx = f,  $f \in Y$ , is (conditionally) correctly posed since  $T^+$  is continuous. Certainly, the Y-norm is very sharp and not adequate in practical cases.

For later convenience we introduce the notation  $Q_{t'}^j(t) = Q^j(t, t')$ . In order to approximate (2.3) we introduce a least-squares collocation method. Let  $\pi_{j,\nu} \subset I_j$  be a finite set of collocation points for  $\nu \in \mathbb{N}$ ,  $1 \leq j \leq m$ . Without loss of generality we assume that  $\pi_{j,\nu-1} \subset \pi_{j,\nu}$ . For  $f = (f_1, \ldots, f_m) \in Y$ , let  $x_{\nu} \in X$  be the least-squares solution of

$$T_i x(t) = f_i(t), \quad t \in \pi_{i, v}, \ 1 \le j \le m.$$
 (2.6)

By Lemma 2.2, (2.6) is equivalent to

$$[T_i x, Q_i^j]_i = [f_i, Q_i^j]_i, \quad t \in \pi_{i,v}, \ 1 \le j \le m.$$
 (2.7)

**Lemma 2.3.** Let  $Q^j$  be continuous and  $\bigcup_{\nu=1}^{\infty} \pi_{j,\nu}$  dense in  $I_j$ . Then:

- (i)  $T_i x \in C(I_i)$  for all j.
- (ii) Let  $W_{j,\nu} = \lim \{Q_t^j | t \in \pi_{j,\nu}\} \subset Y_j$ . Then  $W_{j,\nu-1} \subset W_{j,\nu}$  and  $\bigcup_{\nu=1}^{\infty} W_{j,\nu}$  is dense in  $Y_i$ .
- (iii) Let  $W_v = \lim \{e_j Q_t^j | t \in \pi_{j,v}, 1 \le j \le m\} \subset Y$ .  $e_j$  be the j-th unit vector in  $\mathbb{R}^m$ . Then  $W_{v-1} \subset W_v$  and  $\bigcup_{v=1}^{\infty} W_v$  is dense in Y.

*Proof.* Let  $x \in X$ ,  $1 \le j \le m$ . Then, for all  $s, t \in I_j$ ,

$$|T_j x(s) - T_j x(t)| = |(x, \eta_s^j) - (x, \eta_t^j)| \le ||x|| ||\eta_s^j - \eta_t^j||,$$
  
and 
$$||\eta_s^j - \eta_t^j||^2 = (\eta_s^j, \eta_s^j) - 2(\eta_s^j, \eta_t^j) + (\eta_t^j, \eta_t^j) = Q^j(s, s) - 2Q^j(s, t) + Q^j(t, t).$$

Since  $Q^j$  is continuous, (i) follows.

In order to prove (ii) we show that  $\left(\bigcup_{v=1}^{\infty} W_{j,v}\right)^{\perp} = \{0\}$ . Let  $f_j \in Y_j$  such that

$$f_j \perp W_{j,v}$$
 for all  $v \in \mathbb{N}$ . Hence,  $0 = [f_j, Q_t^j]_j = f_j(t)$ ,  $t \in \pi_{j,v}$ . Since  $\bigcup_{v=1}^{\infty} \pi_{j,v}$  is dense

in  $I_j$  and  $f_j \in \mathbf{R}(T_j)$ , we obtain  $f_j = 0$  by (i). (iii) follows immediately from (ii).  $\square$  The next theorem states the convergence of the method (2.6). If  $W_v \subset \mathbf{R}(T)$ , the convergence follows from well-known theorems (cf. [6, 7]). In our setting  $W_v \subset \mathbf{R}(T)$  is not true in general.

**Theorem 2.1.** Let  $Q^j$  be continuous and  $\bigcup_{\nu=1}^{\infty} \pi_{j,\nu}$  dense in  $I_j$ . Let  $x_{\nu}$  be defined by (2.6).

- (i) If  $f \in \mathbf{R}(T)$ ,  $x_v \to T^+ f$  for  $v \to \infty$ .
- (ii) If  $f \notin \mathbf{R}(T)$ ,  $\{x_v\}$  does not contain a weakly convergent subsequence, i.e.,  $||x_v|| \to \infty$  for  $v \to \infty$ .

*Proof.* This proof is a slight modification of Groetsch [15, Chapter 4.3]. Let  $\sigma_v: \{(j,t) | t \in \pi_{j,v}, 1 \le j \le m\} \to \{1, ..., \mu(v)\}$  be a one-to-one mapping and  $q_i = e_j Q_t^j$  with  $l = \sigma_v(j,t)$ .

(i) Denote the mapping  $f \to ([f, q_1], \ldots, [f, q_{\mu(\nu)}])$  by  $r_{\nu}$ . Setting  $T_{\nu} = r_{\nu} T$  we have  $x_{\nu} = T_{\nu}^{+} r_{\nu} f$ . Set  $V_{\nu} = \text{lin} \{ T^{*} q_{1}, \ldots, T^{*} q_{\mu(\nu)} \}$ . Let  $f \in \mathbf{R}(T)$  be fixed. Because of  $T^{+} f \in \mathbf{N}(T)^{\perp} = \mathbf{R}(T^{*})$  there is a  $z \in \mathbf{N}(T^{*})^{\perp} = \mathbf{R}(T)$  such that  $T^{*} z = T^{+} f$ . Since  $W_{\nu} = \text{lin} \{ q_{1}, \ldots, q_{\mu(\nu)} \}$ , there are a  $\kappa \in \mathbb{N}$  and a  $u \in W_{\kappa} \subset W_{\nu} \ (\nu \ge \kappa)$  such that  $\|u - z\| < \varepsilon / \|T\|$  for given  $\varepsilon > 0$ . Therefore,  $\|T^{*} u - T^{+} f\| = \|T^{*} u - T^{*} z\| < \varepsilon$ . Since  $T^{*} u \in V_{\nu} \ (\nu \ge \kappa)$ ,

$$P_{\nu} T^{+} f \to T^{+} f \quad (\nu \to \infty)$$
 (2.8)

where  $P_{\nu}$  denotes the orthogonal projection onto  $V_{\nu}$ . Furthermore,

$$T_{\nu}T^{+}f = r_{\nu}TT^{+}f = r_{\nu}Pf = r_{\nu}f$$
 (2.9)

with the orthogonal projection P onto  $\mathbf{R}(T)$ . But

$$x \in \mathbf{N}(T_{v})$$
 iff  $T_{v} x = 0$   
iff  $[Tx, q_{l}] = 0, l = 1, ..., \mu(v)$   
iff  $(x, T * q_{l}) = 0, l = 1, ..., \mu(v)$   
iff  $x \in V_{v}^{\perp}$ .

Hence, by (2.8)–(2.9),

$$x_{\nu} = P_{\nu} T^{+} f \to T^{+} f.$$
 (2.10)

(ii) By (2.7),

$$[Tx_v - f, q_l] = 0, \quad l = 1, ..., \mu(v).$$
 (2.11)

Let  $\{x_{\nu_k}\}$  be a weakly convergent subsequence of  $\{x_{\nu_k}\}$ ,  $x_{\nu_k} \rightarrow z$  for  $k \rightarrow \infty$ . Since T is continuous,  $Tx_{\nu_k} \rightarrow Tz$ . By (2.11),  $[Tz - f, q_i] = 0$  for  $i = 1, ..., \mu(\nu)$ . Hence, Tz - f = 0 because of Lemma 2.3. Thus,  $f \in \mathbf{R}(T)$ . This contradiction proves the assertion.  $\square$ 

By (2.10),  $x_v = P_v T^+ f$  where  $P_v$  is the orthogonal projection of X onto  $V_v = T^*(W_v)$ . If  $e_i Q_i^j \in W_v$  and  $x \in X$ ,

$$(T^* e_j Q_t^j, x) = [e_j Q_t^j, Tx] = [Q_t^j, T_j x]_j = T_j x(t) = (\eta_t^j, x).$$

Hence,  $T^*e_j Q_t^j = \eta_t^j$  and  $V_v = \lim \{\eta_t^j | t \in \pi_{j,v}, 1 \le j \le m\}$ . From now on we suppose that the set  $\{\eta_t^j | t \in \pi_{j,v}, 1 \le j \le m\}$  forms a basis of  $V_v$ . This assumption must be checked in actual problems since  $\eta_t^j$  depends essentially on T. For integral equations,  $\eta_t^j$  is connected with the kernel of the integral operator. Integral kernels which have this property for every  $\pi_{j,v}$  are called "quasi-Čebyšev" in [12]. Using some index mapping  $\sigma_v$  we define  $p_l = \eta_t^j$ ,  $l = \sigma_v(j, t)$ , as above. Now,

$$x_{v} = \sum_{k=1}^{\mu(v)} c_{k} p_{k}, \quad c_{k} \in \mathbb{R}.$$
 (2.12)

Multiplying (2.12) by  $p_l$ ,  $l = 1, ..., \mu(v)$ , we obtain  $\sum_{k=1}^{\mu(v)} c_k(p_k, p_l) = (x_v, p_l)$ = $(x_v, \eta_t^j) = [f, e_i Q_t^j]$ , i.e.,

$$\sum_{k=1}^{\mu(v)} c_k(p_k, p_l) = f_j(t)$$
 (2.13)

where  $l = \sigma_v(j, t)$ . (2.13) is a linear system of equations for determing the coefficients  $c_k$ . The matrix  $A_v = (p_k, p_l)_{k,l=1,...,\mu(v)}$  is a Grammian matrix of linearly independent elements, therefore,  $A_v$  is positive definite. We investigate the influence of perturbations of the right-hand side on the least-squares solution. Denote by  $d_l = f_i(t)$  the exact data in (2.13) and  $\tilde{d}_l \in \mathbb{R}$  erroneous data with

$$\|(\tilde{d}_1, \dots, \tilde{d}_{u(v)}) - (d_1, \dots, d_{u(v)})\| \le \delta_v.$$
 (2.14)

The following theorem is due to Engl [6].

**Theorem 2.2.** Let the suppositions of Theorem 2.1 be true and  $f \in \mathbf{R}(T)$ . Denote by  $\lambda_{\mathbf{v}}$  the smallest eigenvalue of  $A_{\mathbf{v}}$ . If  $\limsup \delta_{\mathbf{v}}^2/\lambda_{\mathbf{v}} = 0$ , then  $\tilde{x}_{\mathbf{v}} \to T^+ f$  where  $\tilde{x}_{\mathbf{v}}$  be the solution of (2.13) with  $d_I$  replaced by  $\tilde{d}_I$ .

Remark. A generalization of an analogous result on weak convergence is possible by introducing further suppositions: If  $\limsup \delta_{\nu}^2/\lambda_{\nu} < \infty$  and there exist  $f_{\nu} \in \mathbf{R}(T)$  such that  $f_{\nu,j}(t) = d_i$  for every  $\nu$ , l, then  $x_{\nu} \to T^+ f$ . The additional assumption seems to be very restrictive.

Using the notations of Theorem 2.2 we obtain the error estimate

$$\|\tilde{x}_{v} - T^{+}f\| = \|x_{v} - T^{+}f\| + \delta_{v} \lambda_{v}^{-1/2}.$$
 (2.15)

The first term of the right-hand side is the discretization error. Regarding the fact that  $x_v \in \mathbf{N}(T)^{\perp}$  and Remark (ii) of Lemma 2.2 we obtain

$$\begin{split} \|x_{v} - T^{+} f\| &= \|T^{+} (Tx_{v} - f)\| \leq \|T^{+}\| \|Tx_{v} - f\| \\ &\leq \|T^{+}\| \sum_{j=1}^{m} \|(\widetilde{P}_{v} f)_{j} - f_{j}\|_{j} \\ &\leq T^{+}\| \sum_{j=1}^{m} \|\widetilde{P}_{j, v} f_{j} - f_{j}\|_{j} \end{split}$$

where  $\tilde{P}_{\nu}$  and  $\tilde{P}_{j,\nu}$  are the orthogonal projections onto  $W_{\nu}$  and  $W_{j,\nu}$ , respectively. Estimates of the convergence order of the terms of the right-hand side in terms of the mesh size of  $\pi_{j,\nu}$  and the smoothness of the reproducing kernels  $Q^{j}$  are well-known (cf. [20, 21]).

The second term of (2.15) represents the ill-posedness of the problem. Generally,  $\lambda_{\nu} \to 0$  for  $\nu \to \infty$ . But until now, no estimates of  $\lambda_{\nu}$  in terms of X,  $Q^{j}$ ,  $\pi_{j,\nu}$ , and T have been known.

## 3. Least-Squares Collocation for DAE's

Now we return to the DAE (1.1)–(1.2). In this section we show how the method (1.3) fits into the abstract setting introduced in the previous section. Let m=n+1. With  $I_i=[0, 1]$ ,  $1 \le j \le n$ , and  $I_m=\{1, ..., r\}$  we define

$$T_{j}x(t) = \sum_{i=1}^{n} (a_{ji}(t) x_{i}'(t) + b_{ji}(t) x_{i}(t)), \quad 1 \le j \le n$$
(3.1)

$$T_m x(t) = D_0 x(0) + D_1 x(1). (3.2)$$

The underlying Hilbert space X be some Sobolev space  $[H^k(0, 1)]^n$ . For (2.1) to hold,  $k \ge 2$  should be chosen. In order to reduce the computational expense we select the minimal value for k and set

$$X = \lceil H^2(0, 1) \rceil^n \tag{3.3}$$

endowed with the scalar product

$$(x, y) = \sum_{i=1}^{n} \left\{ \int_{0}^{1} x_{i}''(t) y_{i}''(t) dt + x_{i}(0) y_{i}(0) + x_{i}(1) y_{i}(1) \right\}.$$
(3.4)

The construction of the functions  $\eta_t^j$  is based on the following lemma. It is an extension of a procedure given in [21].

**Lemma 3.1.** Let  $X = X_1 \times ... \times X_n$ , where  $X_i$  is an RKHS on sets  $I^i \subset \mathbb{R}$  with the reproducing kernels  $R^i$ , i.e.,  $(x, R^i_s)_i = x(s)$   $(x \in X_i, s \in I^i)$ . Let  $T_j$  be given such that (2.1) holds. Then (2.2) holds for  $\eta^i_j = (\eta^{ij}_t, ..., \eta^{nj}_t)$  where  $\eta^{ij}_t(s) = T_j e_i R^i_s(t) \in X_i$ .

*Proof.* Let  $\delta_{it}$  be defined by  $x_i \in X_i \mapsto T_j e_i x_i(t)$ . Because of (2.1),  $\delta_{it} \in X_i^*$ . Hence, there are  $\eta_i^{ij} \in X_i$  such that

$$T_i e_i x_i(t) = (x_i, \eta_t^{ij})_i$$
 (3.5)

for all  $x_i \in X_i$ . Denote  $\eta^j = (\eta_t^{1j}, \dots, \eta_t^{nj}) \in X$ . Hence, for all  $x \in X$ ,

$$T_j x(t) = T_j \left( \sum_{i=1}^n e_i x_i(t) \right) = \sum_{i=1}^n (x_i, \eta_t^{ij})_i = (x_i, \eta_t^j),$$

i.e. (2.2). Moreover, by (3.5),

$$\eta_t^{ij}(s) = (\eta_t^{ij}, R_s^i)_i = (R_s^i, \eta_t^{ij})_i = T_i e_i R_s^i(t).$$

In our setting we have  $X_i = H^2(0, 1)$  with the scalar product

$$(x, y)_i = \int_0^1 x''(t) y''(t) dt + x(0) y(0) + x(1) y(1).$$

The reproducing kernel is given by

$$R_s(t) = \frac{1}{6}(s-1)t^3 + (s+\frac{5}{6}(s-1) - \frac{1}{6}(1-s)^3)t + 1 - s + \begin{cases} 0, & t \le s \\ \frac{1}{6}(t-s)^3, & t > s \end{cases}$$
(3.6)

Lemma 3.1 gives

$$\eta_t^j(s) = a_j(t) R_s'(t) + b_j(t) R_s'(t), \quad j = 1, ..., n, \ t \in I$$

$$\eta_t^m(s) = d_t^0 R_s(0) + d_t^1 R_s(1), \quad \iota \in I_m$$

with  $a_j(t) = (a_{j1}(t), \ldots, a_{jn}(t)), b_j(t) = (b_{j1}(t), \ldots, b_{jn}(t)), d_i^k = (d_{i1}^k, \ldots, d_{in}^k)$  and  $D_k = (d_{i\kappa}^k)$  for k = 0, 1. Now, by (2.4) for  $1 \le j \le n$ ,

$$\begin{split} Q^{j}(t,\,t') &= \sum_{i=1}^{n} \left\{ a_{ji}(t)\,a_{ji}(t') \left[ \frac{1}{2}\,t^{2} + \frac{1}{2}\,t'^{2} - t' + \frac{7}{3} \right] \right. \\ &+ a_{ji}(t)\,b_{ji}(t') \left[ \frac{1}{6}\,t'^{3} - \frac{1}{2}\,t'^{2} + \frac{1}{2}\,t^{2}(t'-1) + \frac{7}{3}\,t' - 1 \right] \\ &+ b_{ji}(t)\,a_{ji}(t') \left[ \frac{1}{6}\,t^{3} + \frac{1}{2}\,t'^{2}\,t - t\,t' + \frac{7}{3}\,t - 1 \right] \\ &+ b_{ji}(t)\,b_{ji}(t') \left[ \frac{1}{6}\,t^{3}(t'-1) + t\left( \frac{1}{6}\,t'^{3} - \frac{1}{2}\,t'^{2} + \frac{1}{3}\,t' \right) + 1 - t' - t + 2\,t\,t' \right] \right\}, \ t \leq t'. \\ Q^{m}(\iota,\,\kappa) &= \sum_{i=1}^{n} \left( d_{ii}^{0}\,d_{\kappa i}^{0} + d_{ii}^{1}\,d_{\kappa i}^{1} \right), \quad \iota,\,\kappa = 1,\,\ldots,\,r. \end{split}$$

Obviously, the reproducing kernels  $Q^{j}$  are continuous. For a properly chosen set of collocation points, Theorem 2.1 ensures the convergence of the method.

Let 
$$\pi_{j,\nu} \subset I = [0, 1], j = 1, ..., n$$
, be sets of collocation points. Let  $\bigcup_{\nu=1}^{\infty} \pi_{j,\nu}$  be dense

in I. Without loss of generality we may assume that  $\pi_{m,v} = I_m$ ,  $v \in \mathbb{N}$ . For Theorem 2.2 to hold we have to check the assumption on the linear independence of the set  $\{\eta_t^j | t \in \pi_{j,v}, 1 \le j \le m\}$ . It turns out that very weak suppositions on the matrices A and B are sufficient for this property to hold.

**Lemma 3.2.** Let  $0, 1 \notin \pi_{j, \nu}$  for j = 1, ..., n. Then  $\{\eta_t^j | t \in \pi_{j, \nu}, 1 \le j \le m\}$  is linearly independent if the matrix pencil (A(t), B(t)) is regular for all  $t \in (0, 1)$  and rank  $(D_0, D_1) = r$ .

*Proof.* Without loss of generality we assume that  $\pi_{j,\nu} \equiv \pi_{\nu} = \{t_1, \ldots, t_{\kappa}\}$  for  $j=1,\ldots,n$ . It is known that the set  $\{R_s(0), R_s(t_1), \ldots, R_s(t_{\kappa}), R_s(1), R_s'(t_1), \ldots, R_s'(t_{\kappa})\}$  is linearly independent (functions with respect to s!). Consider

$$0 = \sum_{k=1}^{\kappa} \sum_{j=1}^{n} \alpha_{jk} \, \eta_{t_k}^{j}(s) + \sum_{i=1}^{r} \alpha_{m_i} \, \eta_{i}^{m}(s)$$

$$= \sum_{k=1}^{\kappa} \left\{ R'_{s}(t_k) \left[ \sum_{j=1}^{n} \alpha_{jk} \, a_{j}(t_k) \right] + R'_{s}(t_k) \left[ \sum_{j=1}^{n} \alpha_{jk} \, b_{j}(t_k) \right] \right\}$$

$$+ R_{s}(0) \sum_{i=1}^{r} \alpha_{m_i} \, d_{i}^{0} + R_{s}(1) \sum_{i=1}^{r} \alpha_{m_i} \, d_{i}^{1}.$$

Hence, for all  $k = 1, ..., \kappa$ 

$$\sum_{j=1}^{n} \alpha_{jk} a_{j}(t_{k}) = \sum_{j=1}^{n} \alpha_{jk} b_{j}(t_{k}) = 0.$$

If one of the  $\alpha_{jk}$ 's is nonzero, the vectors  $\{\lambda a_j(t_k) + b_j(t_k) | 1 \le j \le n\}$  are linearly dependent for all  $\lambda \in \mathbb{R}$  in contradiction to the regularity of the matrix pencil  $(A(t_k), B(t_k))$ . Analogously, from

$$\sum_{i=1}^{r} \alpha_{mi} d_{i}^{0} = \sum_{i=1}^{r} \alpha_{mi} d_{i}^{1} = 0$$

we obtain  $\alpha_{m1} = ... = \alpha_{mr} = 0$  because of rank  $(D_0, D_1) = r$ .  $\square$ 

### 4. Examples

Finally, I add some examples for illustrating the behaviour of the method developed above. In all examples, the problem is uniquely solvable. For every example, the number of equispaced gridpoints  $\nu$ , the minimal resp. maximal eigenvalue  $\lambda_{\min}$  ( $\lambda_{\max}$ ) of the matrix  $A_{\nu}$ , and the error of the approximation in the norms of  $C[0, 1](e_c)$  and  $L^2(0, 1)(e_0)$  are recorded. The examples were run in double precision (about 16 decimal digits) on an EC 1055 M computer.

Example 4.1.

$$A(t) = \begin{pmatrix} 0 & 0 \\ 1 & \eta t \end{pmatrix}, \quad B(t) = \begin{pmatrix} 1 & \eta t \\ 0 & 1 + \eta \end{pmatrix}. \tag{4.1}$$

This example is extensively used in the literatur (e.g., [4, 10, 11]). It has several interesting features. The implicit Euler method (i.e., the simplest BDF) is unstable if  $\eta < -1/2$  and convergent and weak unstable if  $\eta > -1/2$ . For  $\eta = -1$ , the pencil (A(t), B(t)) is singular, hence, the implicit Euler method is not applicable. The tables show the results given by our method for  $\eta = -3, -1, 0$  with the right-hand side  $f(t) = (e^t, t^2)$ . For  $\eta = -1$ , Lemma 3.2 does not apply.

 $\eta = -3$ 

ν	$\lambda_{\min}$	$\lambda_{max}$	$e_c$	$e_0$
3	1.8(-2)	5.3(+1)	1.8(+0)	9.6(-1)
5	9.7(-4)	7.6(+1)	1.2(+0)	5.8(-1)
9	7.0(-5)	1.2(+2)	7.9(-1)	3.5(-1)
17	6.6(-6)	2.3(+2)	4.0(-1)	1.5(-1)
33	7.3(-7)	4.2(+2)	2.0(-1)	6.0(-2)
65	5.2(-8)	8.3(+2)	1.0(-1)	2.5(-2)

 $\eta = -1$ 

v	$\lambda_{\min}$	$\lambda_{\max}$	$e_c$	$e_0$
3	1.2(-1)	9.3(+0)	1.1(-1)	9.0(-2)
5	4.9(-4)	1.4(+1)	3.9(-2)	2.7(-2)
9	4.9(-5)	2.6(+1)	2.4(-2)	1.2(-2)
17	5.6(-6)	4.8(+1)	1.2(-2)	5.1(-3)
33	6.8(-7)	9.2(+1)	6.4(-3)	1.8(-3)
65	8.2(-8)	1.8(+2)	3.6(-3)	7.1(-4)

$$\eta = 0$$

ν	$\lambda_{\min}$	$\lambda_{max}$	$e_{c}$	$e_0$	
3	1.0(-2)	8.4(+0)	3.1(-1)	1.6(-1)	
5	3.7(-4)	1.4(+1)	1.0(-1)	4.3(-2)	
9	4.2(-5)	2.4(+1)	4.0(-2)	1.3(-2)	
17	5.1(-6)	4.5(+1)	1.8(-2)	5.2(-3)	
33	6.3(-7)	8.7(+1)	8.5(-3)	2.2(-3)	
65	7.6(-8)	1.7(+2)	3.9(-3)	8.8(-4)	

*Remark.* a.b(c) should be read  $a.b \times 10^{c}$ .

## Example 4.2

$$D_0 = (1, 0, 0, 0), \quad D_1 = (0, 0, 0, 0), \quad \gamma = 1.$$

This equation describes a simple electrical network:

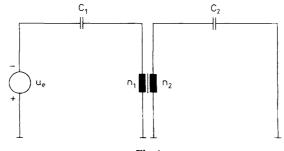


Fig. 1

The index of the system is 2. For simplicity we choose  $C_1 = C_2 = n_1 = 1$  and  $n_2 = 2$ . In a first run we assumed that there is no voltage source, i.e.,  $u_e = 0$ . A second run was made assuming  $u_e = \sin 2\pi t$ . The results are given in the following table.

v	$\hat{\lambda}_{ ext{min}}$	$\lambda_{ ext{max}}$	$u_e = 0$		$u_e = \sin 2\pi t$	
			$e_c$	$e_0$	$e_c$	$e_0$
3	5.0(-2)	1.2(+1)	4.5(-2)	2.7(-2)	5.5(+0)	4.0(+0)
5	1.8(-3)	1.8(+1)	1.1(-2)	6.7(-3)	4.5(+0)	3.2(+0)
9	2.1(-4)	3.3(+1)	2.9(-3)	1.7(-3)	2.8(+0)	1.8(+0)
17	2.6(-5)	6.2(+1)	7.3(-4)	4.2(-4)	1.4(+0)	8.7(-1)
33	3.2(-6)	1.2(+2)	1.8(-4)	1.0(-4)	1.0(+0)	4.7(-1)

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