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Symbolic algorithm for solving SLAEs with multi-diagonal coefficient matrices

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Abstract. Systems of linear algebraic equations with multi-diagonal coefficient matrices may arise after many different scientific and engineering problems, as well as problems of the computational linear algebra where finding the solution of such a system of linear algebraic equations is considered to be one of the most important problems. This paper presents a generalised symbolic algorithm for solving systems of linear algebraic equations with multi-diagonal coefficient matrices. The algorithm is given in a pseudocode. A theorem which gives the condition for correctness of the algorithm is formulated and proven. Formula for the complexity of the multi-diagonal numerical algorithm is obtained.

Key words and phrases: numerical analysis, computational methods for sparse matrices, numerical mathematical programming methods, complexity of numerical algorithms

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1. Introduction

Systems of linear algebraic equations (SLAEs) with multi-diagonal coefficient matrices may arise after many different scientific and engineering problems, as well as problems of the computational linear algebra where finding the solution of a SLAE is considered to be one of the most important problems. For instance, the resultant SLAE after discretization of partial differential equations (PDEs), using finite difference methods (FDM) or finite element methods (FEM) has a banded coefficient matrix. The methods for solving such SLAEs known in the literature usually require the matrix to possess special characteristics so as the method to be numerically correct and stable, e. g. diagonal dominance, positive definiteness, etc. Another possible approach which ensures numerically correct formulae without adding special additional requirements or using pivoting is the symbolic algorithms.

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By definition, a band SLAE is a SLAE with band coefficient matrix. The lower band width p is the number of sub-diagonals, the upper band width q is the number of super-diagonals, and the band width of the matrix is defined as $p + q + 1$ (we should add 1 because of the main diagonal), that is, the total number of non-zero diagonals in the matrix [1]. Here, we are going to focus on SLAEs with matrices for which $p = q = M$. The author of [2] presents a generalised numerical algorithm for solving multi-diagonal SLAEs with pivoting (implemented in Fortran) in the case $p \neq q$, and has applied it for solving boundary value problems discretized by finite difference approximations.

A whole branch of symbolic algorithms for solving systems of linear algebraic equations with different coefficient matrices exists in the literature. For instance, in [3] the author considers a tridiagonal matrix and a symbolic version of the Thomas method [4–7] is formulated. The authors of [8] build an algorithm in the case of a general bordered tridiagonal SLAE, while in [9] the coefficient matrix taken into consideration is a general opposite-bordered tridiagonal one.

A pentadiagonal coefficient matrix is of interest in [10], while a cyclic pentadiagonal coefficient matrix is considered in [11]. The latter algorithm can be applied to periodic tridiagonal and periodic pentadiagonal SLAE either by setting the corresponding matrix terms to be zero.

In [12] a symbolic method for the case of a cyclic heptadiagonal SLAEs is presented.

What is common for all these symbolic algorithms, is that they are implemented using Computer Algebra Systems (CASs) such as Maple [13], Mathematica [14], and Matlab [15].

Finally, [16] presents a symbolic method for the case of a pure heptadiagonal SLAE.

A performance analysis of symbolic methods (and numerical as well) for solving band matrix SLAEs (with three and five diagonals) being implemented in C++ and run on modern (as of 2018) computer systems is made in [17]. Different strategies (symbolic included) for solving band matrix SLAEs (with three and five diagonals) are explored in [18]. A performance analysis of effective symbolic algorithms for solving band matrix SLAEs with coefficient matrices with three, five and seven diagonals being implemented in both C++ and Python and run on modern (as of 2018) computer systems is made in [19].

Having in mind all these introductory notes, it is clear that a generalised multi-diagonal symbolic algorithm is the novelty that addresses the need of a direct method which solves multi-diagonal systems of linear algebraic equations without putting any requirements for the characteristics of the coefficient matrix. Thus, the aim of this paper, which is a logical continuation of [16–19], is to present such a generalised symbolic algorithm for solving SLAEs with multi-diagonal coefficient matrices. The symbolic algorithms investigated in [16–19] are actually particular cases of the generalised multi-diagonal symbolic method when $p = q = M = 1, 2$, and 3 .

The layout of the paper is as follows: in the next section, we outline the multi-diagonal numerical algorithm, and introduce the multi-diagonal symbolic algorithm in pseudocode. Afterwards, we make some correctness remarks for the symbolic method, and present a generalised formula for the complexity of the multi-diagonal numerical algorithm. Finally, some conclusions are drawn.

The novelties of this work are as follows: suggested multi-diagonal symbolic algorithm for solving SLAEs, formulation and proof of a correctness theorem, and an additionally obtained formula for the complexity of the multi-diagonal numerical method.

2. Multi-diagonal symbolic algorithm

Let us consider a SLAE $Ax = y$, where $A = \text{diag}(\mathbf{b}^0, \mathbf{b}^1, \mathbf{b}^2, \dots, \mathbf{b}^M, \mathbf{b}^{M+1}, \mathbf{b}^{M+2}, \mathbf{b}^{M+3}, \dots, \mathbf{b}^{2 \times M})$, A is a real $N \times N$ multi-diagonal matrix with M sub-diagonals, and M super-diagonals, and $2 \times M + 1 < N$, that is, the number of diagonals should be smaller than the number of equations within the SLAE; x

and y are real column vectors with N elements:

$$\begin{bmatrix}
 b_0^M & b_0^{M+1} & b_0^{M+2} & b_0^{M+3} & \dots & b_0^{2M} & 0 & \dots & \dots & \dots & \dots & 0 \\
 b_1^{M-1} & b_1^M & b_1^{M+1} & b_1^{M+2} & \dots & b_1^{2M-1} & b_1^{2M} & 0 & \dots & \dots & \dots & 0 \\
 b_2^{M-2} & b_2^{M-1} & b_2^M & b_2^{M+1} & \dots & b_2^{2M-2} & b_2^{2M-1} & b_2^{2M} & 0 & \dots & \dots & 0 \\
 b_3^{M-3} & b_3^{M-2} & b_3^{M-1} & b_3^M & \dots & b_3^{2M-3} & b_3^{2M-2} & b_3^{2M-1} & b_3^{2M} & 0 & \dots & 0 \\
 b_4^{M-4} & b_4^{M-3} & b_4^{M-2} & b_4^{M-1} & \dots & b_4^{2M-4} & b_4^{2M-3} & b_4^{2M-2} & b_4^{2M-1} & b_4^{2M} & \dots & 0 \\
 \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\
 0 & \dots & 0 & b_{N-4}^0 & b_{N-4}^1 & \dots & \dots & b_{N-4}^{M-1} & b_{N-4}^M & b_{N-4}^{M+1} & b_{N-4}^{M+2} & b_{N-4}^{M+3} \\
 0 & \dots & \dots & 0 & b_{N-3}^0 & b_{N-3}^1 & \dots & \dots & b_{N-3}^{M-1} & b_{N-3}^M & b_{N-3}^{M+1} & b_{N-3}^{M+2} \\
 0 & \dots & \dots & \dots & 0 & b_{N-2}^0 & b_{N-2}^1 & \dots & \dots & b_{N-2}^{M-1} & b_{N-2}^M & b_{N-2}^{M+1} \\
 0 & \dots & \dots & \dots & \dots & 0 & b_{N-1}^0 & b_{N-1}^1 & \dots & \dots & b_{N-1}^{M-1} & b_{N-1}^M
 \end{bmatrix}
 \begin{bmatrix}
 x_0 \\
 x_1 \\
 x_2 \\
 x_3 \\
 x_4 \\
 \vdots \\
 x_{N-4} \\
 x_{N-3} \\
 x_{N-2} \\
 x_{N-1}
 \end{bmatrix}
 =
 \begin{bmatrix}
 y_0 \\
 y_1 \\
 y_2 \\
 y_3 \\
 y_4 \\
 \vdots \\
 y_{N-4} \\
 y_{N-3} \\
 y_{N-2} \\
 y_{N-1}
 \end{bmatrix}.$$

The multi-diagonal numerical solver which we are going to formulate below is a generalization of the Thomas method for multi-diagonal SLAEs. The algorithm is based on LU decomposition, and requires forward reduction for reducing the initial matrix into a lower triangular one:

$$\begin{array}{l|l}
 \mu_0 = b_0^M & \alpha_1^j = 0, \quad i = 1, 2, \dots, M-1 \\
 \alpha_0^{M+1} = \frac{b_0^{M+1}}{\mu_0} & \alpha_1^M = b_1^{M-1} \\
 \alpha_0^{M+2} = \frac{b_0^{M+2}}{\mu_0} & \mu_1 = b_1^M - \alpha_0^{M+1} \times \alpha_1^M \\
 \dots & \alpha_1^{M+1} = \frac{b_1^{M+1} - \alpha_0^{M+2} \times \alpha_1^M}{\mu_1} \\
 \alpha_0^{2M} = \frac{b_0^{2M}}{\mu_0} & \alpha_1^{M+2} = \frac{b_1^{M+2} - \alpha_0^{M+3} \times \alpha_1^M}{\mu_1} \\
 z_0 = \frac{y_0}{\mu_0} & \dots \\
 & \alpha_1^{2M} = \frac{b_1^{2M}}{\mu_1} \\
 & z_1 = \frac{y_1 - z_0 \times \alpha_1^M}{\mu_1}
 \end{array}$$

For $i = 2, 3, \dots, M-1$:

counter = $M - i$

$$\alpha_i^{k-\text{counter}} = b_i^{k-1}, \quad k = M, M-1, \dots, 1, \quad k - \text{counter} \geq 1$$

$$\begin{aligned}
 \alpha_i^{k-\text{counter}} &= \alpha_i^{k-\text{counter}} - \alpha_0^{M+k-\text{counter}-1} \times \alpha_i^1 - \alpha_1^{M+k-\text{counter}-2} \times \alpha_i^2 - \dots \\
 &\quad - \alpha_{k-\text{counter}-2}^{M+1} \times \alpha_i^{k-\text{counter}-1}, \quad k = 2, 3, \dots, M
 \end{aligned}$$

$$\mu_i = b_i^M - \alpha_0^{M+i} \times \alpha_i^1 - \alpha_1^{M+i-1} \times \alpha_i^2 - \dots - \alpha_{i-1}^{M+1} \times \alpha_i^i$$

$$\alpha_i^{M+1} = \frac{b_i^{M+1} - \alpha_0^{M+i+1} \times \alpha_i^1 - \alpha_1^{M+i} \times \alpha_i^2 - \dots - \alpha_{i-1}^{M+2} \times \alpha_i^i}{\mu_i}$$

num_sub = $\min(i, M-2)$ number of subtractions for α_i^{M+2}

$$\alpha_i^{M+2} = \frac{1}{\mu_i} \left(b_i^{M+2} - \alpha_{i-\text{num_sub}}^{M+2+\text{num_sub}} \times \alpha_i^{i-\text{num_sub}+1} \right)$$

$$\begin{aligned}
& -\alpha_{i-\text{num_sub}+1}^{M+2+\text{num_sub}-1} \times \alpha_i^{i-\text{num_sub}+2} - \dots \alpha_{i-1}^{M+3} \times \alpha_i^i) \\
& \dots \\
& \text{num_sub} = \min(i, M-1-(k-1)) \quad \text{number of subtractions for } \alpha_i^{M+k} \\
& \dots \\
& \alpha_i^{2M} = \frac{b_i^{2M}}{\mu_i} \\
& z_i = \frac{y_i - z_0 \times \alpha_i^1 - z_1 \times \alpha_i^2 - \dots - z_{i-1} \times \alpha_i^i}{\mu_i}
\end{aligned}$$

For $i = M, M+1, \dots, N-1$:

$$\begin{aligned}
& \alpha_i^j = b_i^{j-1}, \quad j = 1, 2, \dots, M-1 \\
& \alpha_i^k = \alpha_i^k - \alpha_{i-M}^{M+k-1} \times \alpha_i^1 - \alpha_{i-M+1}^{M+k-2} \times \alpha_i^2 - \dots - \alpha_{i-M+\text{iter}}^{M+k-1-\text{iter}} \times \alpha_i^{1+\text{iter}}, \\
& \quad k = 2, 3, \dots, M, \quad \text{iter} = 0, 1, \dots, k-2, \quad i = M, M+1, \dots, N-1 \\
& \mu_i = b_i^M - \alpha_{i-M}^{2M} \times \alpha_i^1 - \alpha_{i-M+1}^{2M-1} \times \alpha_i^2 - \dots - \alpha_{i-1}^{M+1} \times \alpha_i^M, \\
& \quad i = M, M+1, \dots, N-1 \\
& \alpha_i^{M+1} = \frac{b_i^{M+1} - \alpha_{i-1}^{M+2} \times \alpha_i^M - \alpha_{i-2}^{M+3} \times \alpha_i^{M-1} - \alpha_{i-3}^{M+4} \times \alpha_i^{M-2} - \dots - \alpha_{i-M+1}^{2M} \times \alpha_i^2}{\mu_i}, \\
& \quad i = M, M+1, \dots, N-2 \\
& \alpha_i^{M+2} = \frac{b_i^{M+2} - \alpha_{i-1}^{M+3} \times \alpha_i^M - \alpha_{i-2}^{M+4} \times \alpha_i^{M-1} - \alpha_{i-3}^{M+5} \times \alpha_i^{M-2} - \dots - \alpha_{i-M+2}^{2M} \times \alpha_i^3}{\mu_i}, \\
& \quad i = M, M+1, \dots, N-3 \\
& \dots \\
& \alpha_i^{2M} = \frac{b_i^{2M}}{\mu_i}, \quad i = M, M+1, \dots, N-1-M \\
& z_i = \frac{y_i - z_{i-M} \times \alpha_i^1 - z_{i-M+1} \times \alpha_i^2 - \dots - z_{i-1} \times \alpha_i^M}{\mu_i}, \quad i = M, M+1, \dots, N-1
\end{aligned}$$

and a backward substitution for finding the unknowns x in a reverse order:

$$\begin{aligned}
& x_{N-1} = z_{N-1} \\
& x_{N-k} = z_{N-k} - \alpha_{N-k}^{M+1} \times x_{N-k+1} - \dots - \alpha_{N-k}^{M+1+k-2} \times x_{N-1}, \quad k = 2, 3, \dots, M \\
& x_i = z_i - \alpha_i^{M+1} \times x_{i+1} - \alpha_i^{M+2} \times x_{i+2} - \alpha_i^{M+3} \times x_{i+3} - \dots - \alpha_i^{2M} \times x_{i+M}, \\
& \quad i = N-(M+1), N-(M+2), \dots, 0.
\end{aligned}$$

In order to cope with the stability issue of the Thomas method in the case of non-diagonally dominant matrices, in the case of a zero (or numerically zero) quotient of two subsequent leading principal minors within the symbolic method a symbolic variable is assigned instead and the calculations are continued. At the end of the algorithm, this symbolic variable is substituted with zero. The same approach is suggested in [3].

The full multi-diagonal symbolic method in pseudocode is given in Algorithm 1. There, ε plays the role of a numerical zero, and was set to $1.0e-20$ in our code.

Algorithm 1: Multi-diagonal symbolic algorithm for solving a SLAE $Ax = y$.**Input:** $N, \mathbf{b}^0, \mathbf{b}^1, \dots, \mathbf{b}^M, \mathbf{b}^{M+1}, \dots, \mathbf{b}^{2M}, \mathbf{y}, \varepsilon$ **Output:** \mathbf{x}

```

1: if  $\det(A) == 0$  then
2:   exit
3: end if
4: bool flag = False
5:  $\mu_0 = b_0^M$ 
6: if  $|\mu_0| < \varepsilon$  then
7:    $\mu_0 = \text{symb}$ ; flag = True
8: end if
9: for  $k = \overline{M+1, \dots, 2M}$  do
10:    $\alpha_0^k = \frac{b_0^k}{\mu_0}$ 
11: end for
12:  $z_0 = \frac{y_0}{\mu_0}$ 
13: for  $k = \overline{1, 2, \dots, M}$  do
14:    $\alpha_1^k = b_1^{k-1}$ 
15: end for
16:  $\mu_1 = b_1^M - \alpha_0^{M+1} \times \alpha_1^M$ 
17: if !flag then
18:   if  $|\mu_1| < \varepsilon$  then
19:      $\mu_1 = \text{symb}$ ; flag = True
20:   end if
21: end if
22: for  $k = \overline{M+1, \dots, 2M}$  do
23:    $\alpha_1^k = b_1^k$ 
24:   if  $M > 1$  and  $k < 2M$  then
25:      $\alpha_1^k = \alpha_1^k - \alpha_0^{k+1} \times \alpha_1^M$ 
26:   end if
27:    $\alpha_1^k = \frac{\alpha_1^k}{\mu_1}$ 
28: end for
29:  $z_1 = \frac{y_1 - z_0 \times \alpha_1^M}{\mu_1}$ 
30: for  $i = \overline{2, \dots, N-1}$  do
31:   counter = 0  $\triangleright$  number of non-zero helping  $\alpha_i^k$ ,
32:    $\triangleright$  where  $k = \overline{1, 2, \dots, M}$ 
33:   if  $i < M$  then
34:     counter =  $M - i$ 
35:   end if
36:   for  $k = \overline{1, \dots, M}$  do
37:     if  $k - \text{counter} \geq 1$  then
38:        $\alpha_i^{k-\text{counter}} = b_i^{k-1}$ 
39:     end if
40:   end for
41:    $\triangleright$  above we shift the non-zero  $\alpha_i^j$ ,  $j \leq M$  in order
42:    $\triangleright$  to have them in the interval  $j \in [0; \dots]$ 
43:   for  $k = \overline{M+1, \dots, 2M}$  do
44:      $\alpha_i^k = b_i^k$ 
45:   end for
46:    $\mu_i = b_i^M$ 
47:    $z_i = y_i$ 
48:   iter = 0
49:    $\triangleright$  number of iterations for  $\alpha_i^k$ , where  $k \leq M$ 
50:   coeff = 0
51:    $\triangleright$  the biggest distance between the lower coeff of
52:    $\triangleright \alpha_i^k$  and  $\alpha_{\text{coeff}+\text{iter}}^{M+\dots}$ 
53:   if  $i \geq M$  then
54:     coeff =  $i - M$ 
55:   end if
56:   for  $k = \overline{2, \dots, M}$  do
57:     iter = 0
58:     for  $l = \overline{2, \dots, k - \text{counter}}$  do
59:        $\alpha_i^{k-\text{counter}} = \alpha_i^{k-\text{counter}}$ 
60:        $\quad - \alpha_{\text{coeff}+\text{iter}}^{M+k-1-\text{counter}-\text{iter}} \times \alpha_i^{1+\text{iter}}$ 
61:       iter =  $1 + \text{iter}$ 
62:     end for
63:   end for
64:   iter = 0
65:    $\triangleright$  number of iterations for  $\mu_i$  and  $z_i$ 
66:   if  $i < M$  then
67:     mu_max_iter =  $i - 1$ 
68:   else
69:     mu_max_iter =  $M - 1$ 
70:   end if
71:   for iter =  $\overline{0, 1, \dots, \text{mu\_max\_iter}}$  do  $\triangleright \mu_i, z_i$ 
72:      $\mu_i = \mu_i - \alpha_{\text{coeff}+\text{iter}}^{2M-\text{counter}-\text{iter}} \times \alpha_i^{1+\text{iter}}$ 
73:      $z_i = z_i - z_{\text{coeff}+\text{iter}} \times \alpha_i^{1+\text{iter}}$ 
74:   end for
75:   if !flag then
76:     if  $|\mu_i| < \varepsilon$  then
77:        $\mu_i = \text{symb}$ ; flag = True
78:     end if
79:   end if
80:    $z_i = \frac{z_i}{\mu_i}$ 
81:   iter = 0  $\triangleright$  number of iterations for a
82:    $\triangleright$  particular  $\alpha_i^m$ ,  $m \geq M+1$ 
83:   alpha_counter = 0  $\triangleright$  number of
84:    $\alpha_i^m$ ,  $m \geq M+1$ 
85:   for  $m = \overline{0, 1, \dots, M-1}$  do
86:     num_sub[m] = 0
87:      $\triangleright$  number of subtractions in  $\alpha_i^m$ 
88:   end for
89:   for  $m = \overline{M+1, \dots, 2M-1}$  do
90:     m_index =  $m - M - 1$ 
91:      $\triangleright$  shift the index from 0 to  $M - 2$ 
92:     if  $i \leq M - 1$  then

```

```

92:     num_sub[m_index] =
93:         min(i, M - 1 - alpha_counter)
94:     else
95:         num_sub[m_index] =
96:             M - 1 - alpha_counter
97:     end if
98:     iter = 0
99:     for k = 0, 1, ..., num_sub[m_index] - 1 do
100:         coeff = i - num_sub[m_index] + iter
101:         coeff_1 = 0
102:         if i ≥ M then
103:             coeff_1 = (M - num_sub[m_index]
104:                 + iter) % M + 1
105:         else
106:             coeff_1 = i - num_sub[m_index]
107:                 + iter + 1
108:         end if
109:         ▷ the helping  $\alpha_i^{\text{coeff}_1}$  are with upper index
110:         ▷ up to M, therefore we need to find the
111:             module(M)
112:          $\alpha_i^m = \alpha_i^m$ 
113:             -  $\alpha_{\text{coeff}}^{m+\text{num\_sub}[m\_index]-\text{iter}} \times \alpha_i^{\text{coeff}_1}$ 
114:         iter = iter + 1
115:     end for
116:     alpha_counter = alpha_counter + 1
117:     for k = M + 1, ..., 2M do
118:          $\alpha_i^k = \frac{\alpha_i^k}{\mu_i}$ 
119:     end for
120: end for
121: end for
122:  $x_{N-1} = z_{N-1}$  ▷ Step 2. Solution
123: for i = N - 2, ..., 0 do
124:      $x_i = z_i$ 
125:     iter = 0
126:     for k = 0, ..., M - 1 do
127:         if i + iter > n - 2 then
128:             break
129:         end if
130:          $x_i = x_i - \alpha_i^{M+1+k} \times x_{i+1+k}$ 
131:         iter = iter + 1
132:     end for
133: end for
134: Cancel the common factors in the numerators and
    denominators of x, making them coprime. Substi-
    tute symb := 0 in x and simplify.

```

Remark: If any μ_i expression has been evaluated to be zero or numerically zero, then it is assigned to be a symbolic variable. We cannot compare any of the next μ_i expressions with ε , because any further μ_i is going to be a symbolic expression. To that reason, we use a boolean flag which tells us if any previous μ_i is a symbolic expression. In that case, comparison with ε is not conducted as being not needed.

3. Justification of the algorithm

Let us make some observations on the correctness of the proposed algorithm. In case the algorithm assigns μ_i for any $i = \overline{0, N-1}$ to be equal to a symbolic variable (in case μ_i is zero or numerically zero), this ensures correctness of the formulae for computing the solution of the considered SLAE (because we are not dividing by (numerical) zero). However, this does not add any additional requirements to the coefficient matrix so as to keep the algorithm stable.

Theorem 2. *The only requirement to the coefficient matrix of a multi-diagonal SLAE so as the multi-diagonal symbolic algorithm to be correct is nonsingularity.*

Proof. As a direct consequence of the transformations done so as the matrix A to be factorized and then the downwards sweep to be conducted, it follows that the determinant of the matrix A in the terms of the introduced notation is:

$$\det(A) = \prod_{i=0}^{N-1} \mu_i|_{\text{symb}=0},$$

because the determinant of an upper triangular matrix is equal to the product of all its diagonal elements [20]. (This formula could be used so as the nonsingularity of the coefficient matrix to be checked.) If μ_i for any i is assigned to be equal to a symbolic variable, then it is going to appear in both the numerator and the denominator of the expression for the determinant and so it can be cancelled:

$$\begin{aligned}
\det(A) &= \mu_0 \mu_1 \mu_2 \dots \mu_{N-2} \mu_{N-1} = M_0 \frac{M_1}{\mu_0} \frac{M_2}{\mu_0 \mu_1} \dots \frac{M_{N-2}}{\mu_0 \mu_1 \dots \mu_{N-3}} \frac{M_{N-1}}{\mu_0 \mu_1 \dots \mu_{N-2}} = \\
&= \frac{\prod_{i=0}^{N-1} M_i}{\mu_0^{N-1} \mu_1^{N-2} \mu_2^{N-3} \dots \mu_{N-3}^2 \mu_{N-2}^1} = \frac{\prod_{i=0}^{N-1} M_i}{M_0^{N-1} \frac{M_1^{N-2}}{\mu_0^{N-2}} \frac{M_2^{N-3}}{\mu_0^{N-3} \mu_1^{N-3}} \dots \frac{M_{N-3}^2}{\mu_0^2 \mu_1^2 \dots \mu_{N-4}^2} \frac{M_{N-2}^1}{\mu_0^1 \mu_1^1 \dots \mu_{N-3}^1}} = \\
&= \frac{\prod_{i=0}^{N-1} M_i}{M_0^{N-1} \frac{M_1^{N-2}}{\mu_0^{N-2}} \frac{M_2^{N-3}}{\mu_0^{N-3} \left(\frac{M_1}{\mu_0}\right)^{N-3}} \dots \frac{M_{N-3}^2}{\mu_0^2 \left(\frac{M_1}{\mu_0}\right)^2 \dots \left(\frac{M_{N-4}}{\mu_{N-3}}\right)^2} \frac{M_{N-2}^1}{\mu_0^1 \left(\frac{M_1}{\mu_0}\right)^1 \dots \left(\frac{M_{N-3}}{\mu_{N-4}}\right)^1}} = \frac{\prod_{i=0}^{N-1} M_i}{\prod_{i=0}^{N-2} M_i} = M_{N-1}
\end{aligned}$$

where M_i is the i -th leading principal minor, and $\mu_0 = M_0$. This means that the only constraint on the coefficient matrix is $M_{N-1} \neq 0$. \square

Remark: above, we have used the following recurrent formula $M_i = \prod_{j=0}^i \mu_j$.

Remark: this theorem coincides with the theorem we have proven in [16], because no matter what the number of diagonals ($2 \times M + 1$) within the coefficient matrix is, the logic remains.

The requirement on the coefficient matrix to be nonsingular is not limiting at all since this is a standard requirement so as the SLAE to have only one solution.

3.1. Number of computational steps

The calculation of $\alpha_i^k, \mu_i, \alpha_i^{M+1}, \alpha_i^{M+2}, \dots, \alpha_i^{2M}$, and z_i depends on the results of the calculation of α_{i-j}^{M+k} , and z_{i-j} . On the other hand, the calculation of x_i depends on the results of the calculation of $\alpha_i^{M+1}, \alpha_i^{M+2}, \dots, \alpha_i^{2M}, z_i$, and $x_{i+1}, x_{i+2}, \dots, x_{i+M}$. This makes the multi-diagonal numerical method inherently serial. It takes $2 \times N$ steps overall, where N is the number of equations in the initial SLAE.

3.2. Complexity

The amount of operations per expression are summarized in Table 1. Thus, the overall complexity of the multi-diagonal numerical algorithms is:

$$2NM^2 + 5NM + N - \frac{4M^3}{3} - \frac{7M^2}{2} - \frac{13M}{6},$$

where N is the number of rows in the initial coefficient matrix. Hence, the multi-diagonal numerical method requires only $O(N)$ operations (provided that $M \ll N$) for finding the solution, and beats the Gaussian elimination which requires $O(N^3)$ operations.

4. Results

Within this paper we formulated the multi-diagonal numerical solver which is a generalization of the Thomas method for multi-diagonal SLAEs. In Algorithm 1, we introduced the pseudocode of the the generalised symbolic algorithm for solving SLAEs with multi-diagonal coefficient matrices.

It was proven that the only requirement to the coefficient matrix of a multi-diagonal SLAE so as the multi-diagonal symbolic algorithm to be correct is nonsingularity.

The multi-diagonal numerical method takes $2 \times N$ steps overall, where N is the number of equations in the initial SLAE.

Table 1

Complexity per expression for the multi-diagonal numerical algorithm

expression	# operations	simplified form of # ops	examples		
			$M = 2$	$M = 3$	$M = 4$
$\alpha_i^k, i < M,$ $k = 2, \dots, M$	$\sum_{k=1}^{M-1} (2 \times (1 + 2 + \dots k - 1)) =$ $\sum_{k=1}^{M-1} \left(2 \times \frac{(k-1) \times k}{2} \right)$	$\frac{(M-1) \times M \times (2M-1)}{6}$ $-\frac{(M-1) \times M}{2}$	0	2	8
$\alpha_i^k, i \geq M,$ $k = 2, \dots, M$	$(N-1-M+1)$ $\times \sum_{k=1}^M ((k-1) \times 2)$	$(N-M) \times (M^2 - M)$	$2(N-2)$	$6(N-3)$	$12(N-4)$
$\mu_i, i < M$	$\sum_{k=0}^{M-1} (2k)$	$M^2 - M$	2	6	12
$\mu_i, i \geq M$	$(N-1-M+1) \times 2M$	$2NM - 2M^2$	$4N - 8$	$6N - 18$	$8N - 32$
$\alpha_i^{M+k}, i < M,$ $k = 1, 2, \dots, M$	$\sum_{k=1}^M (\sum_{i=0}^{M-k} (i \times 2 + 1))$ $+(k-1) \times ((M-k) \times 2 + 1)$	M^3 $-\frac{M \times (M+1) \times (2M+1)}{6}$ $+\frac{M \times (M+1)}{2}$	6	19	44
$\alpha_i^{M+k}, i \geq M,$ $k = 1, 2, \dots, M$	$\sum_{k=1}^M ((N-(M+k))$ $\times ((M-k) \times 2 + 1))$	$\frac{NM^2 - 2M^3 - M^2}{3}$ $+\frac{M \times (M+1) \times (2M+1)}{3}$ $-\frac{M \times (M+1)}{2}$	$4N - 13$	$9N - 41$	$16N - 94$
$z_i, i < M$	$\sum_{k=0}^{M-1} (2k+1)$	M^2	4	9	16
$z_i, i \geq M$	$(N-1-M+1) \times (2M+1)$	$2NM + N - M - 2M^2$	$5N - 10$	$7N - 21$	$9N - 36$
$x_{N-k},$ $k = 1, \dots, M$	$\sum_{k=1}^M ((k-1) \times 2)$	$M^2 - M$	2	6	12
$x_{N-k},$ $k = M+1, \dots, N$	$(N-(M+1)+1) \times 2M$	$2NM - 2M^2$	$4N - 8$	$6N - 18$	$8N - 32$
Total	$2NM^2 + 5NM + N - \frac{4M^3}{3} - \frac{7M^2}{2} - \frac{13M}{6}$		$19N - 29$	$34N - 74$	$53N - 150$

The complexity of the multi-diagonal numerical algorithms was found to be:

$$2NM^2 + 5NM + N - \frac{4M^3}{3} - \frac{7M^2}{2} - \frac{13M}{6},$$

where N is the number of rows in the initial coefficient matrix. Hence, the multi-diagonal numerical method requires $O(N)$ operations (provided that $M \ll N$) for finding the solution. The amount of operations per expression were summarized in Table 1. In the Table 1 one can also find the complexity per expression in the cases when $M = 2$, $M = 3$, and $M = 4$.

5. Discussion

Within this paper we formulated the multi-diagonal numerical solver which is a generalization of the Thomas method for multi-diagonal SLAEs. Next, we introduced the pseudocode of the generalised symbolic algorithm for solving SLAEs with multi-diagonal coefficient matrices. There, as a remedy of the stability issue which arises within the Thomas method in the case of non-diagonally dominant matrices, if we obtain a zero (or numerically zero) quotient of two subsequent leading principal minors, a symbolic variable is assigned instead and the calculations are continued. At the end of the

algorithm, this symbolic variable is substituted with zero. The generalised multi-diagonal symbolic algorithm is the novelty that addresses the need of a direct method which solves multi-diagonal SLAEs without putting any requirements for the characteristics of the coefficient matrix. This algorithm is a generalization of the algorithms presented in [3, 10, 16].

It was proven that the only requirement to the coefficient matrix of a multi-diagonal SLAE so as the multi-diagonal symbolic algorithm to be correct is nonsingularity. Asking for nonsingularity of the coefficient matrix is a standard requirement so as the SLAE to have only one solution. Hence, this does not limit the significance of the formulated symbolic algorithm.

The multi-diagonal numerical method takes $2 \times N$ steps overall, where N is the number of equations in the initial SLAE.

The multi-diagonal numerical method requires $O(N)$ operations (provided that $M \ll N$) for finding the solution, and beats the Gaussian elimination which requires $O(N^3)$ operations.

6. Conclusion

A generalised symbolic algorithm for solving systems of linear algebraic equations with multi-diagonal coefficient matrices was formulated and presented in pseudocode. Some notes on the correctness of the algorithm were made. Formula for the complexity of the multi-diagonal numerical algorithm was obtained.

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Символьный алгоритм решения СЛАУ с многодиагональными матрицами коэффициентов

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Аннотация. Системы линейных алгебраических уравнений с многодиагональными матрицами коэффициентов возникают во многих прикладных и теоретических задачах науки и техники, а также в задачах вычислительной линейной алгебры, где их решение представляет собой одну из ключевых проблем. В данной работе представлен обобщённый символьный алгоритм решения систем линейных алгебраических уравнений с многодиагональными матрицами коэффициентов. Алгоритм приведён в виде псевдокода. Сформулирована и доказана теорема, определяющая условие корректности алгоритма. Получена формула, описывающая вычислительную сложность соответствующего численного алгоритма для многодиагональных систем.

Ключевые слова: численные методы, вычислительные методы для разреженных матриц, методы численного математического программирования, вычислительная сложность численных алгоритмов