Numerical Solution of General Bordered ABD Linear Systems by Cyclic Reduction

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Abstract: We generalize the cyclic reduction algorithm to the solution of Bordered ABD linear systems with blocks of different sizes and with different overlaps. This kind of system often arises from the numerical approximation of BVPs with nonseparated boundary conditions. In particular, in our experiments we refer to spline collocation techniques with monomial and B-spline basis functions.

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1 Introduction

We analyze the solution of the BABD linear system

\[ Ax = f \]  

with the coefficient matrix A having the structure shown in Figure 1.

This is the most general BABD matrix structure: we recognize boundary blocks \( B_a \) and \( B_b \) on the first row, and block rows \( V_i \) defined such that there are overlapped columns only between successive blocks. For this reason, each block \( V_i \) may be represented as

\[ V_i = \begin{pmatrix} S_{i-1} & T_i & R_i \end{pmatrix}, \]

where the columns of \( S_{i-1} \) and \( R_i \) overlap columns of \( V_{i-1} \) and \( V_{i+1} \), respectively. Also \( V_1 \) and \( V_N \) have the structure in (2): the first columns of \( V_1 \) are overlapped with those of \( B_a \) while the last columns of \( V_N \) are overlapped with those of \( B_b \), see Figure 1. We set

\[ n_i = (m_{i-1} + k_i + m_i), \]

the size of each block \( V_i \), where \( m_i \) is the number of overlapped columns between the blocks \( V_i \) and \( V_{i+1} \), and \( n_0 \) as the number of rows of the boundary blocks. Then,

\[ \sum_{i=0}^{N} n_i = \sum_{i=0}^{N} m_i + \sum_{i=1}^{N} k_i \] (A is square)
We suppose that $N$ is much larger than $n_i$, $m_i$ and $k_i$.

Solvers and packages for solving BABD (and ABD) linear systems have been considered in [1, 7]. In [2, 3] we have already considered the cyclic reduction algorithm applied to a BABD matrix with a simplified structure. In fact, in these papers we suppose that blocks $S_i$ and $R_i$ are square of dimension $m$ and blocks $T_i$ are null blocks. This means that the coefficient matrix is block lower bidiagonal with an additional block in the right-upper corner. In [4] we also show that the code BABDCR developed by two of the authors outperforms the existing codes RSCLAE and COLROW and may be used in place of COLROW when boundary value ODE problems with nonseparated boundary conditions are considered.

On the other hand, BABD matrices with a more general structure often arise from the numerical approximation of BVPs for ODEs and PDEs with nonseparated boundary conditions. Matrices with nonnull blocks $T_i$ arise, for example, from the numerical solution of linear two-point BVPs of order $m$

$$u^{(m)}(x) - \sum_{s=1}^{m} c_s(x)u^{(s-1)}(x) = f(x) \quad \text{for } x \in [a,b]$$

with $m$ linear boundary conditions

$$B_a z(u(a)) + B_b z(u(b)) = d$$

using collocation at $k$ Gaussian points in $N$ subintervals with B-spline or monomial spline basis. For example, when monomial splines are used, each block $V_i$ is of size $(k + m) \times (k + 2m)$ and boundary blocks $B_a$ and $B_b$ are square of dimension $m$. Moreover, $m$ is the number of overlapped columns with the previous and successive blocks ($k$ is the number of non-overlapped columns); the last $m$ columns have the structure

$$R_i = \begin{pmatrix} 0 \\ I \end{pmatrix}.$$
where \( I \) is a \( m \times m \) identity matrix (the last \( m \) rows represent the continuity conditions while the remaining \( k \) rows in each block \( V_i \) arise from the collocation equation in the gaussian points). When B-splines are used, then the block \( V_i \) is \( k \times (k + m) \) (where \( k \geq m \) with \( m \) overlap columns (\( R_i \) and \( S_i \) have \( m \) columns).

The solution of such systems may be computed using COLROW (the size of blocks \( V_i \) is constant) or ABDPACK which was developed to solve the linear systems arising in spline collocation with monomial bases. All these packages are based on the Varah algorithm [8] which was originally proposed to solve ABD systems, that is, for systems arising from the numerical approximation of BVPs with separated boundary conditions. To apply them to a system arising from nonseparated boundary conditions, these codes require to increase the size of each block \( V_i \) of the original system.

## 2 Cyclic reduction algorithm

Let us rewrite the coefficient matrix in Figure 1 as

\[
A = \begin{pmatrix}
B_a & & & & B_b \\
S_0 & T_1 & R_1 & & \\
& S_1 & T_2 & R_2 & \\
& & S_2 & T_3 & R_3 \\
& & & \cdots & \\
& & & S_{N-1} & T_N \end{pmatrix}.
\]  

(4)

In accordance with the structure of (4) we define the right hand side \( f = (f_0^T, f_1^T, \ldots, f_N^T)^T \) where each \( f_i \) is of length \( n_i \), and the solution vector \( x = (z_0^T, w_0^T, z_1^T, w_1^T, \ldots, z_N^T, w_N^T)^T \), where \( z_i \) and \( w_i \) are of length \( m_i \) and \( k_i \), respectively.

We solve system (1) using a block cyclic reduction algorithm, that is, a recursive approach that reduces the original linear system to subsystems with a smaller number of unknowns. The only condition is that the first and the last unknowns \( z_0 \) and \( z_N \) are always among the unknowns of the successive reduced systems, and also the first row containing the boundary blocks is unchanged in the reduction process (therefore, boundary blocks \( B_a \) and \( B_b \) are maintained in the first row of each reduced system).

Following [2, 3], the first step of reduction would be to combine two successive block rows \( V_i \) and \( V_{i+1} \), for \( i = 1, 3, 5, \ldots \), in order to obtain a new block row \( V'_i \) involving at least the unknowns \( z_{i-1} \) and \( z_{i+1} \). In [2, 3] we suggest eliminating the overlapped columns between \( V_i \) and \( V_{i+1} \) by considering the partial pivoting LU factorization of the block \( \begin{pmatrix} R_i & \\ S_i \end{pmatrix} \). This approach is not appropriate here since the blocks \( T_i \) becomes larger in the successive reduced systems and hence the computational cost is large.

It is preferable to consider the partial pivoting LU factorization:

\[
P_i \begin{pmatrix} T_i & R_i & T_{i+1} \end{pmatrix} = \begin{pmatrix} L_i & I \\ G_i & \end{pmatrix} U_i = \begin{pmatrix} I & I \\ F_i & O \end{pmatrix},
\]

(5)

where \( L_i \) and \( U_i \) are square matrices and \( F_i = G_iL_i^{-1} \). Now, from (5) we have

\[
\begin{pmatrix} I & -F_i \\ -F_i & I \end{pmatrix} P_i \begin{pmatrix} S_{i-1} & T_i & R_i \end{pmatrix} = \begin{pmatrix} \tilde{S}_{i-1} & \tilde{L}_i & \tilde{R}_{i+1} \end{pmatrix}
\]

(6)

\[
\begin{pmatrix} I & -F_i \\ -F_i & I \end{pmatrix} P_i \begin{pmatrix} f_i & f_{i+1} \end{pmatrix} = \begin{pmatrix} \tilde{f}_i & \tilde{f}_{i+1} \end{pmatrix}
\]

(7)
From the second row of (6) and (7) we have a new equation involving \( z_{i-1} \) and \( z_{i+1} \), that is, a reduced system with \([N/2]+1\) block rows and unknowns.

If for each index \( i \) we have \( m_i = m \) and \( k_i = k \) and therefore \( n_0 = m, n_i = m + k \) for \( i \geq 1 \), the computational cost of this first step is

\[
\frac{14}{3} m^3 + 16m^2k + 16mk^2 + \frac{16}{3} k^3 - \frac{3}{2} m^2 - 4mk - 2k^2 - \frac{1}{6} m - \frac{1}{3} k
\]

for each of the \( N/2 \) reductions. This case appears when collocation for monomial spline is considered. On the contrary, for B-spline we must set \( m_i = m \) and \( k_i = k \equiv k - m \).

In (8) we have considered any power of \( \mu \), and the inverse of the lower triangular matrix in the last term

\[
L_i U_i \frac{1}{e_i} \mathbf{w}_i = \frac{1}{f_i} \mathbf{e}_i - S_{i-1} z_{i-1} - R_i z_i
\]

for \( i = 0 \) to \( i = \frac{N}{2} - 1 \). Therefore, this condensation should be considered as a (completely parallelizable) initial step to be applied to ABD or BABD matrices in order to simplify their structure before factorization.

\[
\tilde{L}_i U_i w_i = \tilde{f}_i - \tilde{S}_{i-1} z_{i-1} - \tilde{R}_i z_i.
\]
Returning to the solution of (1), system (11) may be further on reduced by considering the cyclic reduction algorithm in [2, 3] (even if blocks $\hat{S}_i'$ and $\hat{R}_i'$ are not square). At the first step, we consider the LU factorization of the $(n_i - k_i + n_{i+1} - k_{i+1}) \times m_i$ matrix (of rank $m_i$)

$$P_i \begin{pmatrix} \hat{R}_i \\ \hat{S}_i \end{pmatrix} = \begin{pmatrix} L_i \\ G_i \end{pmatrix} U_i = \begin{pmatrix} I & I \\ F_i & I \end{pmatrix} \begin{pmatrix} L_i U_i \\ O \end{pmatrix}$$

that, applied to whole block rows $i$ and $i + 1$ in (11), gives

$$\begin{pmatrix} I \\ -F_i \end{pmatrix} P_i \begin{pmatrix} \hat{S}_{i-1} \\ \hat{R}_i \\ \hat{S}_i \\ \hat{R}_{i+1} \end{pmatrix} = \begin{pmatrix} \tilde{S}_{i-1} \\ L_i U_i \\ \tilde{S}_i \\ \tilde{R}_i \end{pmatrix},$$

(12)

whose second row is independent of $z_i$. The first row may be used to compute $z_i$ from $z_{i-1}$ and $z_{i+1}$.

This last factorization requires additional memory for the fill-in blocks $F_i$. After this step we obtain the same reduced matrix as by the previous approach, but with a number of operations (in case of constant $m$ and $k$)

$$4m^2k + 2mk^2 + \frac{2}{5}k^3 - mk - \frac{1}{2}k^2 - \frac{1}{6}k$$

for each of the $N$ blocks $V_i$, and

$$\frac{14}{3}m^3 - \frac{3}{2}m^2 - \frac{1}{6}m$$

(14)

for each of the $N/2$ reductions. Hence, we save $(4m^2k + 6mk^2 + 2k^3 - mk - \frac{1}{2}k^2)N$ operations.

Iterating this last step on the successively reduced systems we obtain, after $\lceil \log_2 N \rceil$ steps, a $2 \times 2$ block full linear system

$$\begin{pmatrix} B_a & B_b \\ S_0 & R_N \end{pmatrix} \begin{pmatrix} z_0 \\ z_N \end{pmatrix} = \begin{pmatrix} f_0^* \\ f_N^* \end{pmatrix}.$$ 

(15)

The factorization and the solution of (15) gives the first and the last unknown of $x$. Successively, a back substitution phase allows us to compute, in reverse order, all the other unknowns.

If $m$ and $k$ are constant, the computational cost of the algorithm, that we call GBABDCR, is

$$\frac{14}{3}m^3 + 4m^2k + 2mk^2 + \frac{2}{5}k^3 - \frac{3}{2}m^2 - mk - \frac{1}{2}k^2 - \frac{1}{6}m - \frac{1}{6}k).N.$$ 

(16)

If $k = 0$ (all columns overlap), from (16) the cost is $(\frac{14}{3}m^3 - \frac{3}{2}m^2 - \frac{1}{6}m)N$ as in the BABDCR algorithm in [3]. The additional memory requirement (fill-in) is $m^2N$ since the factorization of the $T_i$ blocks does not require fill-in.

### 3 Numerical comparisons

First, we compare the two different cyclic reduction based approaches proposed above. We aim to prove, rather than simply theoretically (GBABDCR requires less computational cost and storage), that condensation may reduce the execution time. We call GBABDCR-0 the non optimized algorithm that, at the first reduction step, uses the factorization (5). Supposing that $m$ and $k$ are constant, then recall that the computational cost of GBABDCR-0 is

$$\frac{14}{3}m^3 + 8m^2k + 8mk^2 + \frac{8}{5}k^3 - \frac{3}{2}m^2 - 2mk - k^2 - \frac{1}{6}m - \frac{1}{6}k).N$$ 

(17)

which is slight larger than that in (16).
Second, we compare these codes with COLROW [6] which is one of the fastest available codes for linear systems with BABD coefficient matrix

\[
A = \begin{pmatrix}
B_{\text{top}} & T_1 & R_1 \\
S_0 & T_1 & R_1 \\
& S_1 & T_2 & R_2 \\
& & \ddots & \\
& & & S_{N-1} & T_N & R_N \\
& & & & b_{\text{bot}} &
\end{pmatrix}
\]  

(18)

where blocks \(S_i\), \(T_i\) and \(R_i\) have constant size (\(m\) is the overlap between different rows and \(k\) is the number of non overlapped columns). COLROW uses a modified alternate row and column elimination which does not create fill-in. For the matrix (18) it has computational cost

\[
\frac{2}{3}(m + k)^3 + (m + k)^2 m_{\text{bot}} + m^2 m_{\text{top}} + m m_{\text{top}} m_{\text{bot}},
\]

where \(m_{\text{top}}\) and \(m_{\text{bot}}\) are the number of rows of \(B_{\text{top}}\) and \(B_{\text{bot}}\), respectively, with \(m_{\text{top}} + m_{\text{bot}} = m\). To apply COLROW to a BABD matrix, we rearrange the system as follows:

\[
\tilde{A} = \begin{pmatrix}
-I & I & & & \\
S_0 & 0 & T_1 & R_1 & \\
-I & 0 & 0 & I \\
S_1 & 0 & T_2 & R_2 & \\
-I & 0 & 0 & I \\
& \ddots & \\
S_{N-1} & 0 & T_N & R_N & \\
-I & 0 & 0 & I \\
& & & & b_{\text{bot}} &
\end{pmatrix}
\]

(19)

\[
\tilde{x} = \begin{pmatrix}
z_0 \\
y_0 \\
w_1 \\
z_1 \\
y_1 \\
w_2 \\
z_{N-1} \\
y_{N-1} \\
w_N \\
\vdots \\
z_N \\
y_N
\end{pmatrix}
\]

\[
\tilde{f} = \begin{pmatrix}
f_0 \\
f_1 \\
f_2 \\
f_3 \\
0 \\
\vdots \\
0
\end{pmatrix}
\]

where \(S_i\), \(T_i\), \(R_i\), \(f_i\), \(z_i\), \(w_i\) are the same blocks and vectors as above and vectors \(y_i\) are new unknowns \(y_N = \ldots = y_1 = y_0 = z_0\). Then, we can apply COLROW to the ABD linear system (19) and compare its results to those of BABDCRG and BABDCR-0 applied to the original BABD linear system. In this case, the cost of COLROW is \(\frac{49}{3}m^3 + \frac{20}{3}m^2k + 4mk^2 + \frac{2}{3}k^3\) \(N\).

We used a AlphaServer DS20E with a 667 MHz EV67 processor and a Compaq Fortran 90 (formerly Digital Fortran 90) compiler. For each test, the elements of the BABD coefficient matrix \(A\) were generated randomly. The right hand side \(f\) was set so that the solution \(x = (1, \ldots, 1)^T\).

In Table 1 we show the execution times, in seconds, as functions of \(k\), \(m\) and \(N\), varying each parameter in turn and fixing the others. The timings confirm our expectations from theory: GBABDCR is faster than GBABDCR-0 and COLROW, though the speed-ups are lower than theory suggests. Moreover, the gap between GBABDCR and GBABDCR-0 increases for larger values of \(k/m\); conversely, GBABDCR improves over COLROW when \(m/k\) increases. The ratios of costs between GBABDCR-0 and GBABDCR and between COLROW and GBABDCR are approximately 1.5 and 2.8, respectively. In Table 2 we show errors for the same tests. Note that the order of the error is the same for all the methods. The case \(m = 5, k = 10, N = 2000\) gives a slight larger than predicted error, but this is true for all the methods and probably depends on the chosen random matrix.
Table 1: Execution times (in seconds) to solve linear systems with varying $m$, $k$ and $N$ using GBABDCR, GBABDCR-0 and COLROW.

<table>
<thead>
<tr>
<th>problem</th>
<th>GBABDCR</th>
<th>GBABDCR-0</th>
<th>COLROW</th>
</tr>
</thead>
<tbody>
<tr>
<td>$m = 5, k = 10, N = 2000$</td>
<td>6.53E-2</td>
<td>9.86E-2</td>
<td>0.148</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 2000$</td>
<td>0.156</td>
<td>0.222</td>
<td>0.439</td>
</tr>
<tr>
<td>$m = 20, k = 10, N = 2000$</td>
<td>0.529</td>
<td>0.651</td>
<td>1.581</td>
</tr>
<tr>
<td>$m = 10, k = 5, N = 2000$</td>
<td>0.106</td>
<td>0.135</td>
<td>0.317</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 2000$</td>
<td>0.156</td>
<td>0.222</td>
<td>0.439</td>
</tr>
<tr>
<td>$m = 10, k = 20, N = 2000$</td>
<td>0.305</td>
<td>0.490</td>
<td>0.733</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 1000$</td>
<td>7.12E-2</td>
<td>0.105</td>
<td>0.208</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 2000$</td>
<td>0.156</td>
<td>0.222</td>
<td>0.439</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 4000$</td>
<td>0.345</td>
<td>0.463</td>
<td>0.876</td>
</tr>
</tbody>
</table>

Table 2: Computed errors in the solution of linear systems with varying $m$, $k$ and $N$ using GBABDCR, GBABDCR-0 and COLROW.

<table>
<thead>
<tr>
<th>problem</th>
<th>GBABDCR</th>
<th>GBABDCR-0</th>
<th>COLROW</th>
</tr>
</thead>
<tbody>
<tr>
<td>$m = 5, k = 10, N = 2000$</td>
<td>4.12E-07</td>
<td>4.09E-07</td>
<td>9.95E-07</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 2000$</td>
<td>5.56E-10</td>
<td>9.75E-10</td>
<td>3.53E-10</td>
</tr>
<tr>
<td>$m = 20, k = 10, N = 2000$</td>
<td>1.32E-10</td>
<td>6.40E-11</td>
<td>8.15E-11</td>
</tr>
<tr>
<td>$m = 10, k = 5, N = 2000$</td>
<td>4.72E-11</td>
<td>7.88E-11</td>
<td>3.47E-11</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 2000$</td>
<td>5.56E-10</td>
<td>9.75E-10</td>
<td>3.53E-10</td>
</tr>
<tr>
<td>$m = 10, k = 20, N = 2000$</td>
<td>4.13E-11</td>
<td>3.41E-11</td>
<td>3.71E-11</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 1000$</td>
<td>7.08E-11</td>
<td>4.04E-11</td>
<td>4.61E-11</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 2000$</td>
<td>5.56E-10</td>
<td>9.75E-10</td>
<td>3.53E-10</td>
</tr>
<tr>
<td>$m = 10, k = 10, N = 4000$</td>
<td>5.56E-10</td>
<td>1.09E-09</td>
<td>3.53E-10</td>
</tr>
</tbody>
</table>

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References


