FORTRAN

Subroutines for Out-of-Core Solutions of Large Complex Linear Systems

Elizabeth L. Yip
Boeing Commercial Airplane Company
Seattle, Washington

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### Errata


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Item 5

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Line 10 of text

Equation (6)

Equation (7)

2nd line in Proof

3rd line in Proof

4th line in Proof

5th line in Proof

6th line in Proof

7th line in Proof

Equation (8)

Equation (9)

Inequality (10) and the equation above

\[ k \left( I - V^T B^{-1} U \right) = \left\| I - V^T B^{-1} U \right\|_2 \left\| (I - V^T B^{-1} U)^{-1} \right\|_2 \]

\[ k \left( I - V^T B^{-1} U \right) \leq \min \{ \xi_1, \xi_2, \xi_3 \} k(A) k(B). \quad (10) \]

should be

\[ k \left( I + V^T B^{-1} U \right) = \left\| I + V^T B^{-1} U \right\|_2 \left\| (I + V^T B^{-1} U)^{-1} \right\|_2 \]

\[ k \left( I + V^T B^{-1} U \right) \leq \min \{ \xi_1, \xi_2, \xi_3 \} k(A) k(B). \quad (10) \]
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4th line from bottom: $n_k(k=1,2...)
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Line 9 of Subroutine ETCIT: $\text{will overwrites}
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Line 6: $\text{TRIDIAGONAL}
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Abstract

This document describes the design and usage of two main subprograms using direct methods to solve large linear complex systems, of the form $Ax = b$, whose coefficient matrices are too large to be stored in core. The first main subprogram is for systems whose coefficient matrices are of a particular sparse structure, namely, the matrix $A$ can be written in the form $B + D$, where $B$ is a block-banded system, and $D$ has only a few columns of nonzeros. Key elements of the algorithms used in the subprograms include: the data structure, the strategy for preserving numerical stability, the adaptability of the algorithms for dense systems as well as for block-profile systems.
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1.0 SUMMARY

This document describes the design and usage of two main subprograms using direct methods to solve large linear complex systems of the form $Ax = b$, whose coefficient matrices are too large to be stored in core. In the first main subprobram, the basic idea is to reduce the matrix to an upper triangular matrix and subsequently solve the problem by backward substitution. A row interchanging strategy called "threshold pivoting" is adopted to preserve numerical stability and to minimize disk-transfers. This algorithm does not yield the usual LU factorization of some row permutation of the coefficient matrix. The second main subprogram is designed for linear systems with a certain sparse structure, namely, the matrix can be written as $B + D$ where $B$ is a block-banded matrix, and $D$ has only a few columns of nonzeros. A variant of the Sherman-Morrison updating formula is employed in this case. The second main subprogram calls the first main subprogram to solve $B(x,y) = (b,u)$ where $x,y$ are unknowns, $b$ is the right-hand side of the linear system $(B + D)x = b$, and $U$ is the nonzero columns of $D$. 

2.0 INTRODUCTION

This document describes the design and usage of two FORTRAN main subprograms for the CDC digital computer systems, namely subroutine ETCGP and subroutine ETCSM. These subroutines were written as part of a research effort investigating unsteady transonic flow. References 1 through 5 present a procedure for analyzing the flow about harmonically oscillating airfoils and wings in the transonic regime. This procedure is based on small perturbations. A solution is formulated using finite difference techniques which results in a large set of simultaneous equations, written matrix form as

\[ Ax = b \]  

(1)

where \( A \) is a sparse complex matrix of order equal to the number of mesh points. The matrix \( b \) is a complex matrix with the number of columns equal to the number of modes for which pressure distributions are to be found. The matrix \( A \) may be of order 3,500 for a practical two-dimensional airfoil.

Numerical solutions to equation (1) were initially obtained using relaxation techniques. However, for a significant range of practical values of Mach number and reduced frequency, experience showed that relaxation techniques applied to equation (1) failed to converge. The matrix \( A \) in equation (1) does not possess any of well-known properties (e.g., positive definiteness, diagonal dominance) which guarantee the convergence of relaxation methods. However, the physical origin of equation (1) guarantees the existence of a unique solution. Thus the obvious alternative is a direct solution method, which assumes no properties of the matrix \( A \) other than existence of a unique solution for an equation of the form of (1) when any arbitrary matrix \( b \) is applied.

Out-of-core algorithms have been discussed in standard textbooks of numerical analysis (e.g. Reference 6). These algorithms are numerically as stable as the well-known regular Gaussian elimination, yet they are not designed for efficient execution in modern day computers. The physical origin of equation (1) makes “blocking” of the matrix \( A \) an obvious and attractive data structure. Yet existing “blocked linear equation solvers” at best assume non-singularity and the well-conditioning of the i-th diagonal sub-block at the i-th blocked pivotal step (e.g. Reference 7). This property is unnatural and is not based on the physics of the problem. Our basic algorithm does not require the matrix \( A \) to have this particular property. As stated in the above paragraph, the only property of \( A \) it assumes is non-singularity of the matrix \( A \). It also takes into consideration numerical stability and computational efficiency. Sections 43 and 44 discuss these advantages in detail.

The FORTRAN subroutines described here are applicable to both dense and sparse matrices that are too large to be stored in core. Thus they are useful for solving any equation of the form of equation (1), in particular in the case when there is no useful properties of the matrix \( A \) to guarantee convergence of any iterative methods. For example, subroutine ETCGP (as well as the version for real matrices) has been used in different research and production computer codes at the Boeing Company.
Results of applying the routines of this document together with the program described in reference 5 are presented in reference 4. The development of the routines of this document was in conjunction with work described in references 4, 5, and 8.

The first step in developing the algorithm for solving equation (1) is to partition the augmented matrix $[A|B]$ into blocks so that it can be considered to have the following structure:

$$
\begin{bmatrix}
A_{11} & A_{12} & \ldots & A_{1S} & B_1 \\
A_{21} & A_{22} & \ldots & A_{2S} & B_2 \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
A_{S1} & A_{S2} & \ldots & A_{SS} & B_S
\end{bmatrix}
$$

*Figure 1. - Data Structure*

Each block is stored as a record in mass storage with the requirement that at least three blocks can be held in core simultaneously, and the diagonal blocks are square blocks. The mass storage used should be a random access file (called direct access file in the IBM nomenclature).

The methods used in both main subprograms are direct methods. In the first main subprogram, the basic idea is to reduce the coefficient matrix to an upper triangular system. In the second main subprogram, the coefficient matrix is assumed to have the form $B + D$, where $B$ is a block-banded matrix, and $D$ is a matrix with only a few columns of nonzeros. A variant of the Sherman-Morrison updating formula is employed.

Features of the two subprograms include options to
- Solve more than one set of right-hand sides
- Control the frequency of pivoting
- Access the submoduli of the main subprograms
- Take advantage of the block sparsity of the coefficient matrix.

In section 3.0 we shall list our symbols and nomenclatures. In section 4.0, we shall discuss the algorithms in detail, analyze their numerical stabilities, remark on their FORTRAN implementation and compare our approaches with outer approaches outlined in the literature. In section 5.0, we shall discuss in detail the usage of the two subprograms.
3.0 NOMENCLATURE

A = (a\textsubscript{i, j})

A is a matrix, the entry of the i-th row and j-th column is denoted by a\textsubscript{i, j}.

A = (A\textsubscript{i, j})

A is a matrix partitioned into blocks, the block in the i-th block row and the j-th block column is denoted by A\textsubscript{i, j}.

A\textsuperscript{T}

The transpose of A.

Upper triangular matrix

Matrix with all zeros below the diagonal.

Block profile matrix

Matrix of the form:

![Block profile matrix diagram](image)

(where the shaded area indicates non-zero entries)
Block-Banded matrix

Matrix of the form:

\[
\begin{array}{ccc}
\cdot & \cdot & \cdot \\
\cdot & & \\
\cdot & & \\
\cdot & & \\
\end{array}
\]

(Where the shaded area indicates non-zero entries)

Threshold pivoting and pivotal tolerance

Threshold pivoting is a row interchanging strategy controlled by a pivotal tolerance parameter \( \mu \). The \( \mu \) parameter is a real number such that \( 0 \leq \mu \leq 1 \). If \( A = \left( a_{i,j} \right) \), and \( a_{i,j} \geq \max_{j > i} \left| a_{i,j} \right| \), then there is no row interchanging, otherwise interchange rows \( i \) and \( m \) where \( m = \max_{j > i} \left| a_{i,j} \right| \)

Random Access file

Multi-record mass storage input/output file which allows the user to create, access and modify its records on a random basis without regard for their physical positions or internal structures. (They are called direct access files in IBM nomenclature.)

Sherman-Morrison Updating Formula for \( A = B + UV^T \):

\[
A^{-1} = B^{-1} - B^{-1} U \left( I_4 + V^T B^{-1} U \right)^{-1} V^T B^{-1}
\]
Matrix norms:
\[
\|A\| = \max_i \left\{ \sum_j |a_{ij}| \right\}
\]
\[
\|A\|_\infty = \max_{ij} \left\{ |a_{ij}| \right\}
\]
\[
\|A\|_2 = \max_i \left\{ \left( \sum_j a_{ij}^2 \right)^{1/2} \right\}
\]

Condition Number of a nonsingular matrix
\[
K_m(A) = \|A\|_m \cdot \|A^{-1}\|_m \quad m = 1, 2, \ldots, \infty
\]

Growth (growth factor) of a reduction process
The largest absolute value of the numbers generated by the reduction process
4.0 DISCUSSION

In this section, we highlight the special features of our algorithms. In sections 4.1 to 4.4, we shall discuss Algorithm I, which reduces the coefficient matrix to an upper triangular matrix and solves the problem by backward substitution. Then we shall remark on FORTRAN implementation and compare the efficiency of this algorithm with the approaches outlined by Bjorck and Dahquist (ref. 6), and Reid, (private communication).

In sections 4.5 and 4.7, we shall discuss Algorithm II, which employs a variant of the Sherman-Morrison updating formula. Then we shall remark on its FORTRAN implementation, and the numerical stability of our particular application of the Sherman-Morrison updating formula.

4.1 ALGORITHM I (SUBROUTINE ETCGP)

For the convenience of discussion, we shall firstly assume the coefficient matrix $A$ to be dense. We shall show how the algorithm can be modified for a block-profile system.

We shall first illustrate the algorithm with a 3 x 3 block system, and assume no row interchanging is necessary. In this case, the algorithm yields Crout's LU decomposition. The LU factors can replace the original coefficient matrix $A$ on disk. Assume, graphically, $A$ is stored on disk as in the following figure.

\[
\begin{bmatrix}
A_{11} & A_{12} & A_{13} \\
A_{21} & A_{22} & A_{23} \\
A_{31} & A_{32} & A_{33}
\end{bmatrix}
\]

*Figure 2.—Original Matrix*

In the *first* block pivotal step, we form the LU factorization of $A_{11}, A_{11} = L_1 U_1$ and replace $A_{11}, A_{12}, A_{13}, A_{21}, A_{31}$ as in figure 3.
For $i = 2, 3$, write $L_i^{-1}A_{ii}$ as $U_{ii}$, and $U_i^{-1}A_{ii}$ as $L_{ii}$.

In the second block pivotal step, replace $A_{22}$ with $A_{22} - L_{21}U_{12}$, decompose the resultant into its LU factors, and complete the rest of the second block pivotal step as illustrated in fig. 4.

For $i = 2, 3$, write $L_i^{-1}A_{ii}$ as $U_{ii}$, and $U_i^{-1}A_{ii}$ as $L_{ii}$.

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In the second block pivotal step, replace $A_{22}$ with $A_{22} - L_{21}U_{12}$, decompose the resultant into its LU factors, and complete the rest of the second block pivotal step as illustrated in fig. 4.
Thus we have reduced the matrix $A$ to an upper triangular matrix, and thus equation (1) can be solved by backward substitution.

Now we shall explain our row interchanging strategy, which is commonly known as threshold pivoting. A real number $\mu$ is chosen so the $0 \leq \mu \leq 1$. We perform row interchanging only when $|a_{ii}| < \mu \cdot \max_{j \neq i} |a_{ji}|$.

Note that if $\mu = 0$, there is no row interchanging at all; if $\mu = 1$, we have the regular row interchanging. We shall illustrate the application of this row interchanging strategy to a $2 \times 2$ block system:

$$
\begin{bmatrix}
U_1 \\
A_{21}
\end{bmatrix}
= 
\begin{bmatrix}
\bar{u}_{11} & u_{12} \\
0 & a_{22} \\
a_{31} & a_{32} \\
a_{41} & a_{42}
\end{bmatrix}
$$

*Figure 6.—First Pivotal Step of a 2 x 2 Block System*

In the first pivotal step, we first decompose $A_{11}$ into its LU factors with pivoting, i.e., $A_{11} = L_1 U_1 P_1$, then replace $A_{12}$ with $L_1^{-1} P_1 A_{12}$.

Let $U_{12} = L_1^{-1} P_1 A_{12}$. Now suppose

$$
U = 
\begin{bmatrix}
\bar{u}_{11} & u_{12} \\
0 & \bar{u}_{22}
\end{bmatrix}
$$

and

$$
|\bar{u}_{11}| > \mu \cdot \max \{ |a_{31}|, |a_{41}| \} = \mu \cdot |a_{31}|
$$

We have to interchange row 1 of $U_1$ with row 1 of $A_{21}$, then eliminate $u_{11}$ and $a_{41}$ with $a_{31}$ as the pivot, the resultant is as follows:

$$
\begin{bmatrix}
\bar{a}_{31} & \bar{a}_{32} \\
0 & \bar{u}_{22} \\
0 & \bar{u}_{12} \\
0 & \bar{a}_{42}
\end{bmatrix}
$$
Now suppose

$$|u_{22}| < \mu \cdot \max \{ |u_{12}|, |a_{42}| \} = \mu |a_{42}|$$

We have to interchange row 2 with row 2 of $A_{21}$, and eliminate $u_{12}$ and $u_{22}$ with $a_{42}$ as a pivot, the resultant is:

$$\begin{bmatrix}
  a_{31} & a_{32} \\
  0 & a_{42} \\
  0 & 0 \\
  0 & 0
\end{bmatrix}$$

After the first block pivotal step, the disk storage should be as follows:

$$\begin{bmatrix}
  L_1 U_1 & L_1 P_1 A_{12} \\
  Multipliers and pivoting information & A_{22}
\end{bmatrix}$$

(In the formal description of our algorithm, we shall write $L_1 P_1 A_{12}$ as $U_{12}$, and the disk record that stores the multipliers and pivoting information as indicated in the previous diagram $L_{21}$.)

The following is a formal description of Algorithm I:

**Algorithm I**

Let $A$ be an $n \times n$ block system

For $i = 1$ step 1 until $n$ do

  For $j = i$ step 1 until $n$ do

    Read $A_{ij}$ into core

    For $k = 1$ step 1 until $i-1$ do

      Read $L_{jk}, U_{kj}$ into core

      Modify $A_{ij}$ by repeating the row operations done to $A_{kj}$.

      If there is row interchanging between $U_k$, rewrite $U_{kj}$ on disk.

    End For

  End For

End For
Eliminate $A_{ij}$ from $V_i$ with row interchanging if necessary.

Store multipliers and pivoting information as $L_{ji}$ on disk.

Repeat the same row operation to the right-hand side, and solve for $x$ by backward substitution.

Note:

1. If there is no interblock pivoting, it requires two disk read/write's in the innermost do-loop; otherwise, it requires three.

2. Nowhere in the algorithm do we assume non-zeros in the diagonal blocks. Although nonzeros in the final upper triangular form will indicate matrix singularity, and backward substitution will be impossible to carry out.
Now we shall show how the algorithm can be modified for a block-profile system.

Define an $n \times n$ matrix $K$ such that:

- $K(i, j) = 0$ indicates the $(i, j)$ block is a zero block.
- $K(i, j) \neq 0$ indicates $K(i, j)$ is the location of the $(i, j)$ block (a nonzero block) on disk.

For example, for the following block-profile matrix,

```
1 2
3
4 5
6 7 8 9
10 11
```

Where the shaded area indicates the nonzero blocks, its corresponding $K$ matrix can be

We call the $K$ matrix the profile matrix of the matrix $A$.

In subroutine ETCGP, we scan the profile matrix to determine the beginning and end of each block row and block column, and change the "ends" the do-loops in the algorithm accordingly. To handle the "fill-in's" of the zero blocks (i.e., if the $(i, j)$ block is originally a zero block, but the reduction process turns it into a nonzero block), we find the maximum entry of the original $K$ matrix, $m$, and set $K(i, j) = m + 1$, update $m$ by adding 1 to it, and revise the information on the beginnings and ends of the $i$-th block row and $j$-th block column if necessary.

### 4.2 REMARKS ON FORTRAN IMPLEMENTATION OF ALGORITHM I

1. **Storage of the multipliers and pivoting information**

   Suppose the $i, j$ block is a nonzero block that is $n_i \times n_j$, and $i \succ j$. The number of words in its corresponding record on disk is $2*(n_i \times n_j) + n_j + 1$. Before the $j$-th block pivotal step, the first $2*(n_i \times n_j)$ words stored the entries of the $(i, j)$ block, and the last $n_j + 1$ words are zeros. After the $j$-th block pivotal step, the first $2*(n_i \times n_j)$ words store the multipliers used, and the last $n_j + 1$ words store the pivoting information. We can consider this record as consist-
ing of two separate arrays: an array of complex numbers, \( A(n_1, n_j) \), and an array of integers, \( I(n_j+1) \). In the 2 x 2 block system we used to illustrate the threshold pivoting discussion, after the first pivotal step, the (2,1) block of fig 6., that is \( A_{21} \), is stored as an array of complex numbers, \( A(2,2) \), as

\[
\begin{pmatrix}
U_{11}/a_{31} \\
a_{41}/a_{31} \\
U_{12}/a_{42} \\
U_{22}/a_{42}
\end{pmatrix}
\]

and the pivotal information is stored in the integer array, \( I(3) \), as

\[
\begin{bmatrix}
1 \\
2 \\
3
\end{bmatrix}
\]

The two arrays are packed as one record on disk.

The matrix \( A \) stores the actual multipliers used in the reduction process, \( I(1)=1 \) indicates the first row of \( A_{21} \) interchanged with row 1 of the upper triangular matrix \( U_1 \) before the first column of \( A_{21} \) is eliminated. \( I(2)=2 \) indicates the second row of \( A_{21} \) interchanged with row 2 of \( U_1 \) before the second column of \( A_{21} \) is eliminated. \( I(3)=I(1)+I(2)>0 \) indicates there is interblock pivoting between the first block row (the pivotal block row) and the second block row. If there is no interblock pivoting in this step, \( I(1)=0 \), and \( I(2)=0 \), thus \( I(3)=0 \). As it is obvious from this example, \( I(k)=0 \) indicates that no row interchange is needed to eliminate the \( k \)-th column of the \((i,j)\) block. \( I(k)=m>0 \) indicates that the \( m \)-th row of the \((i,j)\) block is interchanged with the \( k \)-th row of the corresponding upper triangular matrix \( U_j \) before the \( k \)-th column of the \((i,j)\) block is eliminated, and \( I(n_j+1)=I(1)+I(2)+\ldots+I(n_j) \). If \( I(n_j+1)=0 \), it indicates that there is no interblock pivoting between the \( i \)-th and \( j \)-th rows; if \( I(n_j+1)>0 \), it indicates that there is interblock pivoting between the \( i \)-th and \( j \)-th block row.

2. Optimization of the Innermost Loop

As indicated by the 3 x 3 block system in section 4.1, if there is no interblock pivoting, the most expensive operation is the matrix operation
In subroutine ETCGP, this operation is carried out by a Compass subroutine CCSMAB which is about five times faster than straight FORTRAN. Even if there is interblock row pivoting, we still can take advantage of CCSMAB. Consider the situation in the innermost do-loop in the description of Algorithm I. We have the following blocks in core:

\[
\begin{bmatrix}
0 & U_{kj} \\
L_{ik} & A_{ij}
\end{bmatrix}
\]

If there is no pivoting between the i-th and j-th block row, the operation can be represented by the following equation:

\[
L \begin{bmatrix}
U_{kj} \\
A_{ij}
\end{bmatrix} = \begin{bmatrix}
U_{kj} \\
A_{ij}
\end{bmatrix}
\]

where \( L \) is such that

\[
L^{-1} = \begin{bmatrix}
I & 0 \\
L_{ik} & 1
\end{bmatrix}
\]

If there is pivoting between the i-th and j-th block rows, then the operation can be represented by the following equation:

\[
\hat{P} L \begin{bmatrix}
U_{kj} \\
A_{ij}
\end{bmatrix} = \begin{bmatrix}
\hat{U}_{kj} \\
\hat{A}_{ij}
\end{bmatrix}
\]

where \( P \) is that permutation matrix that interchanges the rows and \( L \) is such that

\[
\hat{L}^{-1} = I + P \begin{bmatrix}
0 & 0 \\
L_{ik} & 1
\end{bmatrix}
\]
We can write
\[ \hat{L}^{-1} = \begin{bmatrix} L_0 & 0 \\ M & 1 \end{bmatrix} \]
Therefore
\[ \hat{L} = \begin{bmatrix} L_0^{-1} & 0 \\ -ML_0^{-1} & 1 \end{bmatrix} \]
Also write
\[ P \begin{bmatrix} U_{kj} \\ A_{ij} \end{bmatrix} = \begin{bmatrix} \hat{U}_{kj} \\ \hat{A}_{ij} \end{bmatrix} \]
Therefore
\[ \hat{L} \begin{bmatrix} \hat{U}_{kj} \\ \hat{A}_{ij} \end{bmatrix} = \begin{bmatrix} L_0^{-1} \hat{U}_{kj} \\ \hat{A}_{ij} - ML_0^{-1} \hat{U}_{ij} \end{bmatrix} \]
Thus in ETCGP, we compute \( \hat{U}_{kj} = L_0^{-1} \hat{U}_{kj} \) then use CCSMAB to compute \( A_{ij} = \hat{A}_{ij} - M \hat{U}_{kj} \).

Note that \( M \) is just the lower block of \( P \begin{bmatrix} 0 \\ L_{ik} \end{bmatrix} \) and can be computed quite efficiently by forward substitution. (The following describe the computation \( M \) and \( U = L_0^{-1} U_{kj} \).

\( M \) would overwrite \( L_{ik} \) and \( U \) would overwrite \( U_{kj} \). We shall call \( L_{ik}, M \) and \( U_{kj}, U \) and the actual row and column dimension of \( U, NR, \) and \( NC \).)

For \( m=2 \) step 1 until \( NR \) do
  If \( I(m) > 0 \)
  \( k = I(m) \)
  For \( h = 1 \) step 1 until \( NC \) do
    \[ U(m, h) = U(m, h) - \sum_{s=1}^{m-1} M(k, s) \cdot U(s, j) \]
  Enddo
For $h=1$ step 1 until $m-1$ do
  
  $M(k,h)=0$
  
  Enddo
  
  Else
  
  (no operation)
  
  Endif

Enddo

3. Pivotal Tolerance

At this point, we cannot give any theoretical guidance on the choice of the pivotal tolerance $\mu$. Experimentally, $\mu = 0.001$ proves to be satisfactory. In ETCGP, we store $\mu$ in a labelled common block

  COMMON/ETPIVOT/U

and $U$ is set to 0.001 by a DATA statement. The user can reset the value of $U$ by any executable statement before calling ETCGP.

4. Monitoring of Stability

ETCGP keeps track of the "growth" of the reduction process (which is the largest absolute value of all the numbers generated by the reduction process). Our motivation for keeping track of the growth is explained in section 4.3.

ETCGP also generates an extra right-hand side row which is the row sum of the coefficient matrix $A$. It subtracts $(1.0,0)$ from each element in the solution of this extra right-hand side. The resultant gives some indication of the "smallness" of the residue $Ax-b$ of the actual problem.

4.3 NUMERICAL STABILITY OF ALGORITHM I

Reid (ref. 9) and Wilkinson (ref. 10) have analyzed the stability of the regular Gaussian elimination. If $L$ and $U$ are respectively computed lower and upper triangular factors of $A$, the $A$, $L$ and $U$ are related as follows:

  $A = LU + E$

with $E=(e_{i,j})$ and $|e_{i,j}| \leq 3.01 \cdot m \cdot \varepsilon \cdot g$

where $m$ is the order of $A$, and $\varepsilon$ is the machine precision, (in CDC equipment, $\varepsilon = 10^{-14}$), $g$ is the growth of the reduction process.
Reid and Wilkinson analysis cannot be extended in any obvious manner to give a practical bound for \( E \) for Algorithm I of this document. The best we can do at present is the following:

\[
A = M_1^{-1}M_2^{-1} \cdots M_n^{-1}U + M_1^{-1}M_2^{-1} \cdots M_n^{-1}E_n + \cdots + M_1^{-1}M_2^{-1}E_2 + M^{-1}E_1
\]

where \( M_i \) is the matrix that performs the \( i \)th block pivotal step

\[
E_i = (e_{kl}^i) \quad \left| e_{kl}^i \right| \leq 3.01 * e * g_i \quad \text{with} \quad g_i \text{ being the growth of the first } i \text{th block pivotal step.}
\]

The main difficulty is due to the fact that we do not use the same pivots for eliminating the subdiagonal elements in the same column. Thus

\[
M_1^{-1}M_2^{-1} \cdots M_i^{-1}E_i \neq E_i \quad \text{for } i = 1, 2, \ldots, M
\]

whilst equality holds for the above expression in the regular Gaussian elimination. However each block pivotal step the process of eliminating the lower triangular elements of \( A_{i,i} \) from

\[
\begin{array}{cccc}
A_{i,i} & A_{i,i+1} & \cdots & A_{i,n} \\
\end{array}
\]

and the process of elimination \( A_{j,i} \) from

\[
\begin{array}{cccc}
U_{i,i} & U_{i,i+1} & \cdots & U_{i,n} \\
A_{j,i} & A_{j,i+1} & \cdots & A_{j,n} \\
\end{array}
\]

for \( i \leq j \leq n \) are regular Gaussian elimination. Thus the "growth" of Algorithm I at least gives the local stability of these substeps. The norm of the difference of the computed solution from the actual solution give a realistic bound for the norm of \( E \), because the actual solution \( x \) satisfies the equation \( Ax = b \), the computed solution \( y \) satisfies the equation \( (A + E)y = b \).

Thus

\[
A x = (A + E)y
\]

\[
A(x - y) = Ey
\]

\[
\|Ey\| \leq \|A \| \|x - y\|
\]

In all our test problems

\[
\|x - y\| \approx 10^{-11}
\]
4.4 COMPARISON OF ALGORITHM I WITH OTHER APPROACHES

Bjorck and Dahlquist (ref. 6) and Reid (ref. 8) have proposed approaches to the problem of solutions of large dense linear systems. Bjorck and Dahlquist’s approach follows.

The matrix is partitioned into block rows:

\[
\begin{array}{c}
A_1 \\
A_2 \\
\vdots \\
\vdots \\
\vdots \\
A_n
\end{array}
\]

with the requirement that a minimum of two block rows can be held in core simultaneously.

The first block pivotal step can be described as follows:

1. Read \( A_1 \) into core and reduce it to \( \tilde{A_1} \) with regular Gaussian elimination.

2. For \( j=2,3, \ldots, n \) read \( A_j \) into core reduce

\[
\begin{array}{c}
\tilde{A_1} \\
A_j
\end{array}
\]

to

\[
\begin{array}{c}
A_1 \\
A_j
\end{array}
\]

with row-interchanges when necessary, and write \( A_j \) back onto disk.

3. Write \( A_1 \) back onto disk.

The other pivotal steps are similar.

Reid proposed the same data structure as we do. His first block pivotal substeps consists of the following substeps to be performed for \( i=2,3, \ldots, n \).

1. Read into core the blocks \( \begin{pmatrix} A_{11} \\ A_{1i} \end{pmatrix} \) and apply normal Gaussian elimination, overwriting
eliminated elements by multipliers and using row interchanges. Note that for $i > 3$ the modified block $A_{11}$ is upper triangular and advantage may be taken of this.

2. For $j=2,3,...,n$ read in blocks $\begin{pmatrix} A_{1j} \\ A_{ij} \end{pmatrix}$ modify them using the multipliers and interchanges held in $\begin{pmatrix} A_{11} \\ A_{i1} \end{pmatrix}$ and then write them back to disk.

3. Write modified block $A_{11}$ to disk.

The remaining block pivotal steps are performed similarly. Operations on the right-hand side vector may be performed subsequently using the stored multipliers or at the same time as the elimination.

Reid has proved that given the same amount of core, the block row storage scheme is only efficient for matrices of order of less than 300. For large matrices, Algorithm I requires the least amount of disk input/output. The following table summarizes the features of the three approaches.

<table>
<thead>
<tr>
<th>Approaches</th>
<th>Features</th>
<th>No. of disk access</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dahlquist and Bjorck</td>
<td>two block rows are needed in core</td>
<td>$4 \frac{m}{N}^4$</td>
</tr>
<tr>
<td>Reid</td>
<td>four blocks are needed in core</td>
<td>$32 \left( \frac{m}{\sqrt{N}} \right)^3$</td>
</tr>
<tr>
<td></td>
<td>four disk read/write's in the inner most do-loop</td>
<td></td>
</tr>
<tr>
<td>Algorithm I</td>
<td>three blocks are needed in core</td>
<td>$2 \left( \frac{m}{\sqrt{N}} \right)^3$</td>
</tr>
<tr>
<td>(without pivoting)</td>
<td>two disk read/write's in the inner most do-loop</td>
<td></td>
</tr>
<tr>
<td>Algorithm I</td>
<td>three disk read/write in the inner most do-loop</td>
<td>$5 \frac{\sqrt{3}}{2} \left( \frac{m}{\sqrt{N}} \right)^3$</td>
</tr>
<tr>
<td>(with the worst possible case of pivoting)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

* $m$ is the order of the matrix, and $N$ is one half of the number of words available in core.
Algorithm II is designed for coefficient matrix with a special sparsity structure:

\[
A = \begin{bmatrix}
\times & \times & \times & \times & \times \\
\times & \times & \times & \times & \times \\
\times & \times & \times & \times & \times \\
\times & \times & \times & \times & \times \\
\times & \times & \times & \times & \times \\
\end{bmatrix}^{
*}
\]

We can write \( A = B + D \) where \( B \) is a banded matrix:

\[
B = \begin{bmatrix}
\times & \times & \times & \times & \times \\
\times & \times & \times & \times & \times \\
\times & \times & \times & \times & \times \\
\times & \times & \times & \times & \times \\
\times & \times & \times & \times & \times \\
\end{bmatrix}
\]

and \( D = A - B \):

\[
D = \begin{bmatrix}
\times \\
\times \\
\times & \times \\
\times & \times \\
\times & \times \\
\end{bmatrix}
\]

*The x's are single elements*
D can be written as $D = UV^T$, where $U$ and $V^T$ are as follows:

$$U = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ x & x \\ x & x \\ x & x \\ x & x \\ x & x \end{bmatrix}$$

$$V^T = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Thus to solve $x = b$, we apply the Sherman-Morrison updating formula

$$x = A^{-1}b = (B + UV^T)^{-1}b$$

$$= B^{-1}b - B^{-1}u(I + V^TB^{-1}U)^{-1}V^TB^{-1}b$$

The algorithm II can be described as follows:

1. Reduce $B$ to an upper triangular matrix with Algorithm I.
2. Repeat the row operation in 1 to $[b|U]$. Solve for $B [z,y] = [b,U]$ by backward substitution.
4. Decompose $(I+T)$ into its LU factors with row interchanging.
5. Solve for $w$ in the equation $(I-T) w = S$ by forward and substitution.
6. Compute $x = S - T \ast x$. 

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4.6 REMARKS ON FORTRAN IMPLEMENTATION OF ALGORITHM II

In this section, we shall use the notation we have defined so far.

1. Storage of U and V

We partitioned U conformable to the partitioning of the coefficient matrix A:

\[
U = \begin{bmatrix}
    U_1 \\
    U_2 \\
    \vdots \\
    U_n
\end{bmatrix}
\]

and store them on disk in the same block which stores the corresponding partition of the right-hand side.

\[
\begin{bmatrix}
    b_1 & U_1 \\
    b_2 & U_2 \\
    \vdots & \vdots \\
    b_n & U_n
\end{bmatrix}
\]

Since the matrix V only consists of ones and zeros, we do not store V explicitly.

We use two-dimensional array:

\[
\text{INTEGER N(MR,MC)}
\]

where MR \(\geq \) the number of block columns that contain element columns in the matrix \(D=UV^T\), and MC \(\geq 2 + \) the total number of columns in the matrix U. The contents of N are defined as follows:

Let \(i_1,i_2,\ldots\) be the indices of the blocks that contains columns of D, and then

\[
\begin{align*}
    N(1,1) &= i_1 \\
    N(2,1) &= i_2 \\
    \vdots \\
\end{align*}
\]
Suppose for \( j=1,2 \ldots \) all \((i_j,m)\) blocks for some \( m > k_j \) contain the nonzero elements of \( D \), then
\[
N(1,2) = k_1 \\
N(2,2) = k_2 \\
\ldots
\]

Suppose for \( j=1,2 \ldots \), \( D \) contains the following columns of the \( i_j \)-th block:
\[
m_{j,1}, m_{j,2}, \ldots, m_{j,n(j)} \text{ with } m_{j,1} < m_{j,2} < \ldots < m_{j,n(j)}.
\]
Then
\[
N(1,3) = m_{1,1}, N(1,4) = m_{1,2}, \ldots, N(1,n(1)+2) = m_{1,n(1)}, N(1,n(1)+3) = 0 \\
N(2,3) = m_{2,1}, N(2,4) = m_{2,2}, \ldots, N(2,n(2)+2) = m_{2,n(2)}, N(2,n(1)+3) = 0 \\
\ldots
\]

It is important that \( N(j,n(j)+3) \) be set to zero, so that the program know the \( m_{j,n(j)} \) is the last column from the \( i_j \)-th block column to go into \( D \). The following 3 x 3 block system will illustrate our scheme:

\[
A = \begin{bmatrix}
xx \\
xxx \\
xxxx \\
xxx \\
xxx \\
xxx \\
xxx \\
xxx \\
\end{bmatrix}
\]
N(1,1)=1 meaning the first block column contains columns for D
N(2,1)=2 meaning the second block column contains columns for D
N(1,2)=2 meaning in the first block column, only (m,1),m > 2 contains columns for D
N(2,2)=3 meaning in the second block column, only (m,2),m > 3, contains columns for D
N(1,2)=3,N(1,4)=5,N(1,5)=0 meaning only the third and fifth columns of the first block
column belongs to D
N(2,3)=3,N(2,4)=0 meaning only the third column of the second block column belongs
to D

2. Computation of $V^T(z,y)$ (Step 3 of Algorithm II)
Since V is not stored explicitly, we use the following:
(Let NB be the number of block columns that contain column of D, L be the total number
of nonzero columns of D, and the two-dimensional array N, is as defined as before, and
NC=L+number of right-hand sides.)
g=1
For i=1 step 1 until NB do
  k=N(i,1)
  Read $T_k=b_k,U_k$ into core.
  For j=3 step 1 until LU+2 do
    m=N(i, j)
    If m=0, then exit do-loop with index j
Otherwise
S(s,h)=T_K(m,h)
Enddo
g=g+1
Enddo
Note that the array S contains ν^T(z,y)

4.7 NUMERICAL STABILITY OF ALGORITHM II

Suppose A and B are matrices of order n and are related in the following manner:

\[ A = B + UV^T \]  \hspace{1cm} (2)

where U and V are nxr matrices with r ≪ n, and both U and V are of rank r. Then it follows from a variant of the well known Sherman-Morrison formula (ref. 6) that

\[ A^{-1} = B^{-1} - B^{-1}U(I_r + V^TB^{-1}U)^{-1}V^TB^{-1} \]  \hspace{1cm} (3)

where \( I_r \) is the identity matrix of order r. Assume B is well-conditioned and has been decomposed into LU factors B=LU, then equation (3) provides a very efficient method to solve the linear system

\[ Ax = b. \]  \hspace{1cm} (4)

However, a general concern when using equation (3) is that the matrix \( (I_r + Y^TB^{-1}U) \) may be ill-conditioned. In this section, we prove a sufficient condition for the well conditioning of \( (I_r + V^TB^{-1}C) \)

Before we proceed, we state the usually accepted definition of the condition number of a matrix and an inequality related to it.

Let A be an nxm, n≥m, matrix with linearly independent columns, then the conditional number of A, denoted by \( k(A) \) is such that

\[ k(A) = \frac{\max. \|Ax\|}{\min. \|Ax\|} \quad \|x\| = 1 \]
It is also true that

\[ k(A) \geq \|A\|_2 \cdot \|A^+\|_2 \]  

(5)

where \( A^+ = \left( A^T A \right)^{-1} A^T \), the generalized inverse of \( A \).

Let \( R(A) \) be the range of \( A \). Inequality (5) is obvious from the fact that:

\[
\min_{\|x\| \neq 0} \frac{\|Ax\|}{\|x\|} = \frac{\max_{\|x\| \neq 0} \|x\| \|Ax\|}{\|A\|_2} = \frac{\|A^+A\|}{\|A\|_2} \leq \max_{\|y\| = 1} \|A^+y\|.
\]

Further, if \( A \) is nonsingular, equality will hold.

We now prove the following:

**Theorem 1.** If \( A \) and \( B \) are related as in equation (2), and \( A \) and \( B \) are invertible, then

\[
(I - Y^T B^{-1} u) = U^+ AB^{-1} U \quad \text{(6)}
\]

\[
(I - Y^T B^{-1} u) = \left( I_n - U^T B^{-1} \right) B \quad \text{(7)}
\]

where \( U^+ = \left( U^T U \right)^{-1} U^T \), and \( \left( V^T \right)^+ = V \left( V^T V \right)^{-1} \) are the generalized inverses of \( U \) and \( V \), respectively.

**Proof:** Note

\[
A = B - UV^T
\]

\[
= \left( I_n - U^T B^{-1} \right) B
\]

\[
AB^{-1} = \left( I_n - U^T B^{-1} \right)
\]

\[
\therefore AB^{-1} U = \left( U - UV^T B^{-1} \right) U
\]

\[
\therefore AB^{-1} U = \left( I - V^T B^{-1} \right) U
\]

\[
\therefore \left( I - V^T B^{-1} \right) = U^+ AB^{-1} U
\]

Thus we have proved the validity of equation (6). The proof of equation (7) is similar.
The following equations are direct derivations from equations (6) and (7):

\[
(I - VT^{-1}B^{-1}U)^{-1} = U + BA^{-1}U
\]  
(8)

\[
(I - VT^{-1}B^{-1}U)^{-1} = VTA^{-1}B\left(V^T\right)^+
\]  
(9)

Let \( \xi_1 = \left(\|U\|_2,\|U^+\|_2\right)^2, \xi_2 = \left(\|V^T\|_2,\|(V^T)^+\|_2\right)^2, \xi_3 = \|U\|_2\|U^+\|_2\|V^T\|\|(V^T)^+\|_2 \)

Combining equations (6), (7), (8), and (9) with the fact that \( \|CD\|_2 \leq \|C\|_2\|D\|_2 \)
for any two matrices C and D,

\[
k\left(I - VT^{-1}B^{-1}U\right) = \left\|I - VT^{-1}B^{-1}U\right\|_2 \left\|(I - VT^{-1}B^{-1}U)^{-1}\right\|_2
\]

\[
k\left(I - VT^{-1}B^{-1}U\right) \leq \min \{ \xi_1, \xi_2, \xi_3 \} k(A) k(B)
\]  
(10)

Inequality (10) states that if A and B are well-conditioned, and either U or V is well-conditioned then \(I + VT^{-1}B^{-1}U\) is well-conditioned.

Inequality (10) is useful because \(\|U\|_2,\|U^+\|_2,\|V^T\|\) and \(\|(V^T)^+\|\) can be computed quite economically and the physical problems that yield matrices A and B usually give some indication of their well-conditioning. The well-conditioning of B and \(\left(I - VT^{-1}B^{-1}U\right)\) ensures that the solutions of \(Bx_o = b\)

\[
By = U
\]

\[
\left(I + VT^{-1}B^{-1}U\right)Z = V^TX_o
\]
can be computed with satisfactory accuracy. Thus the solution of \(Ax = b\) via equation (3) can be computed with satisfactory accuracy (11).

In our particular application of the Sherman-Morrison updating formula, \(VT\) is a permutation matrix, thus \(\|V^+\|_2\|V\|_2 = 1\). Therefore as long as A and B are both well-conditioned, Algorithm II is always stable.
5.0 COMPUTER PROGRAM USAGE

5.1 MACHINE AND EXECUTION ENVIRONMENT

The algorithms described in the previous section are programmed as a FORTRAN subroutine library (which we call OCSLIB out-of-core solution library). The subroutines are written for the FTN compiler of the CDC computers.

5.2 OPERATING SYSTEM

This subroutine library is designed for NOS 1.1. Its compass subroutines are optimized for the CDC 6600.

5.3 TIMING

Timing is hardware and operating system dependent. The following formula gives a very rough estimate for the timing:

\[ \text{CP second} = \frac{1}{2} \times n \times p^2 \times k \]

where \( n \) is the order of the linear system, and \( p \) is the band-width and \( k \) is machine-dependent. To estimate \( k \), make a sample run and compute \( k \) using the above formula. For the Cyber 175, \( k \approx 8 \times 10^{-5} \)

5.4 FILES AND FILE FORMAT

OCSLIB uses at least one random access file. OCSLIB has each block of the coefficient matrix as a record. If the block is \( n_i \times n_j \), then the record length is \( 2n_i n_j + n_j + 1 \).

5.5 USAGE

5.5.1 USING ALGORITHM I (SUBROUTINE ETCGP)

We recommend the following procedure:

Step 1. Define the block system of the coefficient matrix: choose a sequence of positive
integers \( n_1, n_2, \ldots, n_N \) to partition the matrix into a block system like fig. 1 of section 1.0, so that the block \( A_{p,j} \) is \( n_i \times n_j \). The partition is to be chosen so that at least three blocks can be held in core simultaneously.

Step 2. Define the block profile of the matrix: define an array

\[
\text{INTERGER INDEX (M,M+1)}
\]

such that \( M \geq N \), and \( \text{INDEX}(i, j) = 0 \) if the \( A_{p,j} \) is a zero block.

\[
\text{INDEX}(i, j) > 0 \text{ if the } A_{p,j} \text{ has at least one nonzero entries.}
\]

(here we consider \( b_k = A_{i,N+1,k} \) for \( k = 1, 2, \ldots, N \))

Step 3. Write the augmented matrix on disk for \( \text{ETCGP,} \ nk \ (k = 1, 2, \ldots, N) \) at a time. In this step we provide a subroutine \text{ETBMGEN} to write a block row of the augmented matrix on disk in a format that is acceptable by \text{ETCGP}. Thus we shall describe the usage and argument list of \text{ETBMGEN}.

\[
\text{COMPLEX W(MXR,MXC)}
\]

\[
\text{INTEGER INDEX(MT,MT+1),JN(NT2),NDBLK(NT1) (where NT1>NTBLKS+1}
\]

\[
\text{NT2>NTBLKS**2 + NTBLKS+3)}
\]

\[
\text{.}
\]

\[
\text{.}
\]

\[
\text{DO 10 I=1, NTBLKS}
\]

(Code to generate the i-th block row of the augmented matrix.)

\[
\text{CALL ETBMGEN(I,NTAPE,IN,MT,NTBLKS,WORK,MXR,MXC)}
\]

10 \text{CONTINUE}
The argument list for ETBMGEN is described as follows:

```
SUBROUTINE ETBMGEN(I,NTAPE,INDEX,MT,JN,NTBLKS,NDBLK,WORK,
   $MXR,MXC)
C**************************************************************************
C PURPOSE TO WRITE THE NON-ZERO BLOCKS OF A BLOCK ROW OF MATRIX
C IN ETCGP FORMAT
C ARGUMENT LIST
C
   I - BLOCK ROW INDEX
   NTAPE - OUTPUT DISK FILE
   OUTPUT - ARGUMENTED MATRIX IN ETCGP FORMAT
   INDEX - 2 DIMENSIONAL ARRAY FOR THE PROFILE OF THE MATRIX
          ROW DIMENSION = MT, COLUMN DIMENSION = MT+1
   INDEX(I,J)=0 (I,J)-TH BLOCK IS ZERO
   INDEX(I,J).GT.0 LOCATION OF (I,J)-TH BLOCK ON NTAPE
   MT - ROW DIMENSION OF THE ARRAY INDEX
   JN - INTEGER ARRAY FOR THE RANDOM ACCESS FILE NTAPE, AT LEAST
          (NTBLKS+NTBLKS*NTBLKS+31) MANY WORDS LONG
   NTBLKS - NO. OF BLOCK ROWS
   NDBLK - ARRAY TO STORE BLOCK SIZES, NDBLK(NTBLKS+1)=NO.OF RHS
   WORK - 12 DIMENSIONAL TO USER
          INPUT ARRAY FOR THE NON-ZERO BLOCKS OF THE I-TH BLOCK ROW
   MXR - ROW DIMENSION OF WORK
   MXC - COLUMN DIMENSION OF WORK
C NOTE SUBROUTINE ETCGP ASSUMES ALL THE NON-ZERO BLOCKS TO BE
DENSE BLOCKS. (SEE THE SAMPLE CALLING PROGRAM)
C**************************************************************************
ETBMGEN alters the contents of the array INDEX.

Step 4. To solve the linear system via ETCGP

```COMPLEX WORK (LW)
INTEGER INDEX(MT,MT,+1),NDBLK(NT1)
...
CALL ETCGP(NTAPE,INDEX,NTBLKS,NDBLK,WORK,LW,ITAG)
```

The calling sequence of ETCGP is as follows:

```
SUBROUTINE ETCGP(NTAPE,INDEX,NTBLKS,NDBLK,WORK,LWORK,ITAG)
```
SUBROUTINE ETCGP(NTAPE, INDEX, NTBLKS, NDBLK, WORK, LWORK, ITAG)
OPTION, KEEP=OFF, INLIST=ON

C..................................................
C PURPOSE MAIN SUBROUTINE
C REDUCE MATRIX TO AN UPPER TRIANGULAR MATRIX - CALL ETCGPROM
C REPEAT ROW OPERATION ON RHS - CALL ETCGPFS
C BACKWARD SUBSTITUTION - CALL ETCGPBS
C
C ARGUMENT LIST
C NTAPE - INPUT/OUTPUT DISK FILES FOR MATRIX
C INPUT - ORIGINAL ARGUMENTED SYSTEM
C OUTPUT - ROW OPERATIONS PERFORMED AND UPPER
C TRIANGULAR FORM AND SOLUTION
C INDEX - 2 DIMENSIONAL ARRAY FOR THE PROFILE OF THE MATRIX
C ROW DIMENSION = NTBLKS, COLUMN DIMENSION = NTBLKS+1
C INDEX(I,J)=0 (I,J)-TH BLOCK IS ZERO
C INDEX(I,J).GT.0 LOCATION OF (I,J)-TH BLOCK ON NTAPE
C NTBLKS - NO. OF BLOCK ROWS
C NDBLK - ARRAY TO STORE BLOCK SIZES, NDBLK(NTBLKS+1)=NO. OF RHS
C WORK - WORKING ARRAY (COMPLEX TO USER)
C LWORK - LENGTH OF THE ARRAY WORK REGARDED AS COMPLEX
C LWORK.GE. (NTBLKS**2 + 3*NTBLKS + 3*MXBLK**2 + 3*MXBLK + 3*N)
C WHERE MXBLK = MAXIMUM OF ALL THE ENTRIES OF THE ARRAY NDBLK, AND WHERE
C N IS THE ORDER OF THE SYSTEM
C ITAG - COMPUTATIONAL PATH
C ITAG=1 REDUCE MATRIX TO UPPER TRIANGULAR FORM AND
C SOLVE AX=B
C ITAG=2 REDUCE MATRIX ONLY
C ITAG=3 SOLVE AX=B ASSUMING A HAS BEEN REDUCED
C
C..................................................

(The array WORK should be equivalenced to the arrays W and JN as follows:
EQUIVALENCE (WORK,W) (WW(1,MX+1),JN)

Step 5. Read the solution from NT APE. In this subroutine we provide a subroutine
ETRDSOL which has exactly the same calling sequence as ETBMGEN. Its usage is described
below:

DO 10 I=1,NTBLKS
  CALL ETRDSOL(NTAPE, INDEX, NTBLKS, WORK, MNR, MXR, MXC)
  (code to process the part of the solution corresponding to the i-th block row)
10 CONTINUE
5.5.2 USING ALGORITHM II (SUBROUTINE ETCSM)

We recommend the following procedure:

Step 1. Same as step 1 of section 5.5.1.

Step 2. Same as step 2 of section 5.5.1, except call subroutine ETSMGEN instead of ETBMGEN. The argument list of ETSMGEN is as follows:

ARGUMENT LIST

I, NTAPE, INDEX, MT, JN, NTBLKS, NDBLK, WORK, $MXR, MXC, NB, MLB, LB, LU)

SUBROUTINE ETSMGEN(I, NTAPE, INDEX, MT, JN, NTBLKS, NDBLK, WORK, $MXR, MXC, NB, MLB, LB, LU)

PURPOSE: WRITE THE MATRIX IN A FORM ACCEPTABLE BY ETCSM

ARGUMENT LIST

I, NTAPE, INDEX, MT, JN, NTBLKS, NDBLK, WORK, $MXR, MXC - SEE SUBROUTINE ETBMGEN

NB, MLB, LB, LU GIVE THE SPARSITY STRUCTURE OF THE MATRIX D

NB - DIMENSIONED AS INTEGER NB(MLB, LU+2)

NB(I, 1) = K, THE K-TH BLOCK ROW CONTAINS THE COLUMNS OF D

NB(I, 2) = J, ONLY THE BLOCKS (M, K), M.GE.J CONTAINS THE COLUMNS OF D


MLB - ROW DIMENSION OF NB

LB - TOTAL NO. OF BLOCK COLUMNS THAT CONTAINS COLUMNS OF D

LU - TOTAL NO. OF NON-ZERO COLUMNS OF D

C*************************************************************

Step 3. Solve the linear system via ETCSM. The argument list of ITCSM is as follows:

SUBROUTINE ETCSM(NTAPE, INDEX, NTBLKS, NDBLK, WORK, LWORK, ITAG, $KNB, MLB, LB, LU)

C*************************************************************

PURPOSE: APPLY THE SHERMAN-MORRISON UPDATING FORMULA TO SOLVE (BE + D)X = B, WHERE BE IS A BLOCKED BENDED MATRIX, AND D ONLY CONSISTS OF A FEW COLUMNS OF NON-ZEROES. THE NON-ZERO COLUMNS OF D IS ASSUMED TO BE STORED WITH THE RHS.

ARGUMENT LIST

NTAPE, INDEX, NTBLKS, NDBLK, WORK, LWORK, ITAG

KNB - DIMENSIONED AS INTEGER KNB(MLB, LU+2)

KNB(I, 1) = K, THE K-TH BLOCK ROW CONTAINS THE COLUMNS OF D

KNB(I, 2) = J, ONLY THE BLOCKS (M, K), M.GE.J CONTAINS THE COLUMNS OF D


MLB - ROW DIMENSION OF KNB

LB - TOTAL NO. OF BLOCK COLUMNS THAT CONTAINS COLUMNS OF D

LU - TOTAL NO. OF NON-ZERO COLUMNS OF D

C*************************************************************
5.5.3 USING THE ITERATIVE REFINEMENT SUBROUTINE (ETCIT)

ETCIT requires the original matrix and the output matrix from TECGP to be in different files. The usage of ETCIT is obvious from the explanation of its argument list:

SUBROUTINE ETCIT(NTAPEI, INDEXI, NTAPEO, INDEXO, NDBLK, NTBLKS, WORK, LWORK)
C******************************************************************************************
C PURPOSE ITERATIVE REFINEMENT
C ARGUMENT LIST
C NTAPEI - INPUT DISK FILE STORES ORIGINAL MATRIX
C INDEXI - 2 DIMENSIONAL ARRAY FOR THE PROFILE OF THE MATRIX IN NTAPEI (SEE EXPLANATION FOR THE ARGUMENT INDEX IN SUBROUTINE ETCGP)
C NTAPEO - OUTPUT DISK FILE FROM ETCGP, STORES THE UPPER TRIANGULAR FORM AND THE ROW OPERATIONS PERFORMED BY ETCGP, ALSO THE SOLUTION FROM ETCGP.
C INDEXO - 2 DIMENSIONAL ARRAY FOR THE PROFILE OF THE MATRIX IN NTAPEO (SEE EXPLANATION FOR THE ARGUMENT INDEX IN SUBROUTINE ETCGP)
C NTBLKS - NO. OF BLOCK ROWS
C NDBLK - ARRAY TO STORE BLOCK SIZES; NDBLK(NTBLKS+1)=NO. OF RHS
C WORK - WORKING ARRAY (COMPLEX TO USER)
C LWORK - LENGTH OF THE ARRAY WORK REGARDED AS COMPLEX
C LWORK .GE. (NTBLKS**2 + 3*NTBLKS + 3*MXB + 3*N)
C WHERE MXB = (MXB**2) + (MXB+2)/2 WITH MXB = THE MAXIMUM OF ALL THE ENTRIES OF THE ARRAY NDBLK, AND WHERE N IS THE ORDER OF THE SYSTEM
C******************************************************************************************
5.6 ERROR MESSAGES

5.6.1 ERROR MESSAGES FROM ETCGP

1. WORKING SPACE TOO SMALL, AT LEAST ************* WORDS ARE NEEDED RETURN FROM ETCGP.

2. WRONG CHOICE OF COMPUTATION PATH, ITAG SHOULD EQUAL TO 1, 2 OR 3.

3. MATRIX SEEMED SINGULAR, EXIT FROM ETCGP.

5.6.2 ERROR MESSAGES FROM ETCSM

1. WORKING SPACE TOO SMALL, AT LEAST ************* WORDS ARE NEEDED RETURN FROM ETCSM.

2. MATRIX SEEMED SINGULAR, EXIT FROM ETCSM.

5.6.3 ERROR MESSAGES FROM ETCIT

1. WORKING SPACE TOO SMALL, AT LEAST ************* WORDS ARE NEEDED RETURN FROM ETCIT.

2. CONVERGENCE TOO SLOW, RETURN FROM ETCIT.
5.7 SAMPLE PROBLEMS

5.7.1 SAMPLE PROBLEM 1

PROGRAM TGP(INPUT,OUTPUT,TAPE5=INPUT,TAPE6=OUTPUT,TAPE8

THIS PROGRAM ILLUSTRATES THE USE OF ETCGP.
THE PROFILE OF THE MATRIX IS BLOCK TRIADIONAL.
THE SUBBLOCKS ARE 2 X 2 BLOCKS.
WE CALL THE RANDOM NUMBER GENERATOR TO GENERATE THE MATRIX ENTRIES
WE USE THE ROW SUM OF THE MATRIX AS OUR RHS.

COMPLEX W(1000), WW(10,100)
INTEGER IN(10,11), NDBLK(11), JN(115), KN(110), NC(10)

NOTE THE W ARRAY SHOULD BE EQUIVALENCED TO WW
AND JN SHOULD BE EQUIVALENCED TO WW(1,MVC+1)
EQUIVALENCE (WW, W), (KN, IN), (JN, WW(1,12))
DATA KN/110*0/, W/1000*(0.,0.)/
DATA NTBLKS/5/, NDBLK/2, 2, 2, 2, 2, 1/

NOTE THE W ARRAY SHOULD BE EQUIVALENCED TO WW
AND JN SHOULD BE EQUIVALENCED TO WW(1,MVC+1)
EQUIVALENCE (WW, W), (KN, IN), (JN, WW(1,12))
DATA

GENERATE INDEX ARRAY IN IN

NT1=NTBLKS
NT1=NTBLKS+1
NTM1=NTBLKS-1
IN(1,1)=1
IN(1,2)=2
IN(1,NT1)=3
NS=3
DO3I=2, NTM1
IN(I, I-1)=NS+1
IN(I, I)=NS+2
IN(I, I+1)=NS+3
IN(I, NT1)=NS+4
NS=NS+4
3 CONTINUE
IN(NTBLKS, NTM1)=NS+1
IN(NTBLKS, NTBLKS)=NS+2
IN(NTBLKS, NT1)=NS+3
NS=NS+3
FIND OUT HOW MANY NON-ZERO COLUMNS IN THE BLOCK ROW

D04I=1,NTBLKS
NC(I)=0
D04J=1,NTBLKS
4 IF (IN(I,J).GT.0) NC(I)=NC(I)+NDBLK(J)

WRITE(6,100)((IN(I,J),J=1,11),I=1,10)
100 FORMAT(*INDEX ARRAY IN :*,/*,(11I5))

GENERATE NTBLKS BLOCK ROW OF THE MATRIX

D010I=1,NTBLKS
NC1=NC(I)+1

GENERATE NC(I)*20 RANDOM NUMBERS
CALL NOGEN(W,NC(I)*20)

ZERO OUT WW(.,NC1) FOR ROW SUM
ND=NDBLK(I)

COMPUTE ROW SUM
MC=NC(I)

D06J=1,ND
WW(J,NC1)=(0.,0.)
D06K=1,MC
WW(J,NC1)=WW(J,NC1)+WW(J,K)

WRITE I-TH BLOCK ROW FOR ETCGP

CALL ETEBMGEN (I,8,IN,10,JN,NTBLKS,NDBLK,WW,10,11)

10 CONTINUE

CALL ETCGP(8,IN,NTBLKS,NDBLK,W,1000,1)

TO RED SOLUTION FROM TAPE8

D011I=1,NTBLKS
CALL ETRDSOL (I,8,IN,10,JN,NTBLKS,NDBLK,WW,10,11)
ND=NDBLK(I)

WRITE(6,101)I,(WW(J,1),J=1,ND)
101 FORMAT(1X,15,*-TH BLOCK SOLUTION*,/*,(8E10.4))
STOP
END
5.7.2 SAMPLE PROBLEM 2

PROGRAM TSM(INPUT, OUTPUT, TAPE5=INPUT, TAPE6=OUTPUT, TAPE8)

C THIS PROGRAM ILLUSTRATES THE USE OF ETCM.
C THE PROFILE OF THE MATRIX IS OF THE FORM \( A = B + D \), WHERE B IS
C BLOCK TRIDIAGONAL, AND D ONLY CONSISTS OF 2 NON-ZERO ROWS.
C THE SUBBLOCKS ARE 2 X 2 BLOCKS.
C WE CALL THE RANDOM NUMBER GENERATOR TO GENERATE THE MATRIX ENTRIES
C WE USE THE ROW SUM OF THE MATRIX AS OUR RHS.

COMPLEX W(1000), WW(10, 100)
INTEGER IN(10, 11), NDBLK(11), JN(115), KN(110), NC(10), NB(2, 5)

C NOTE THE W ARRAY SHOULD BE EQUIVALENCED TO WW
C AND JN SHOULD BE EQUIVALENCED TO WW(1, MXC+1)
EQUIVALENCE (WW, W), (KN, IN), (JN, WW(1, 12))
DATA KN/110*0/, W/1000*(0., 0.)/
DATA NTBLKS/5/, NDBLK/2, 2, 2, 2, 2, 1/

C GENERATE INDEX ARRAY IN IN

NT1=NTBLKS
NT1=NTBLKS+1
NTM1= NTBLKS-1
IN(1, 1)=1
IN(1, 2)=2
IN(1, NT1)=3
NS=3
D03I=2, NTM1
IN(I, I-1)=NS+1
IN(I, I)=NS+2
IN(I, I+1)=NS+3
IN(I, NT1)=NS+4
NS=NS+4
3 CONTINUE
IN(NTBLKS, NTM1)=NS+1
IN(NTBLKS, NTBLKS)=NS+2
IN(NTBLKS, NT1)=NS+3
NS=NS+3

C D05I=4, NTBLKS
NS=NS+1
5 IN(I, 2)=NS

C DEFINE THE STRUCTURE OF D

NB(1, 1)=2
NB(1, 2)=4
NB(1, 3)=1
NB(1, 4)=2
C NB(1, 5)=0
C FIND OUT HOW MANY NON-ZERO COLUMNS IN THE BLOCK ROW

37
D04I=1, NTBLKS
NC(I)=0
D04J=1, NTBLKS
4 IF (IN(I,J).GT.O) NC(I)=NC(I)+NDBLK(J)
C WRITE(6,100)((IN(I,J),J=1,11),I=1,10)
100 FORMAT(*INDEX ARRAY IN*,I,(11I5))
C GENERATE NTBLKS BLOCK ROW OF THE MATRIX
C D010I=1, NTBLKS
NC1=NC(I)+1
C GENERATE NC(I)*20 RANDOM NUMBERS
CALL NGEN(W, NC(I)*20)
C ZERO OUT WW(., NC1) FOR ROW SUM
ND=NDBLK(I)
C COMPUTE ROW SUM
MC=NC(I)
C D06J=1, ND
WW(J, NC1)=(0., 0.)
D06K=1, MC
6 WW(J, NC1)=WW(J, NC1)+WW(J, K)
C WRITE I-TH BLOCK ROW FOR ETCGP
C CALL ETSMGEN (I, 8, IN, 10, JN, NTBLKS, NDBLK, WW, 10, 11, NB, 2, 1, 2)
C 10 CONTINUE
C CALL ETCSM(8, IN, NTBLKS, NDBLK, W, 1000, 1, NB, 2, 1, 2)
C TO READ SOLUTION FROM TAPE8
C D011I=1, NTBLKS
CALL ETRDSOL (I, 8, IN, 10, JN, NTBLKS, NDBLK, WW, 10, 11)
ND=NDBLK(I)
11 WRITE(6,101)I, WW(J, 1), J=1, ND)
101 FORMAT(1X,I5, **-TH BLOCK SOLUTION*,/(8E10.4))
STOP
END
REFERENCES


