Rectangular Full Packed Format for Cholesky’s Algorithm: Factorization, Solution and Inversion

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We describe a new data format for storing triangular and symmetric matrices called RFPF (Rectangular Full Packed Format). The standard two dimensional arrays of Fortran and C (also known as full format) that are used to represent triangular and symmetric matrices waste nearly half of the storage space but provide high performance via the use of Level 3 BLAS. Standard packed format arrays fully utilize storage (array space) but provide low performance as there are no Level 3 packed BLAS. We combine the good features of packed and full storage using RFPF to obtain high performance via using Level 3 BLAS as RFPF is a standard full format representation. Also, RFPF requires exactly the same minimal storage as packed format. Each LAPACK full and/or packed symmetric, triangular, and Hermitian routine becomes a single new RFPF routine based on eight possible data layouts of RFPF. This new RFPF routine usually consists of two calls to the corresponding LAPACK full format routine and two calls to Level 3 BLAS routines. This means no new software is required. As examples, we present LAPACK routines for Cholesky factorization, solution and inverse computation in RFPF to illustrate this new work and to describe its performance on several commonly used computer platforms. Performance of LAPACK full routines using RFPF versus LAPACK full routines using standard format for both serial and SMP parallel processing is about the same while using half the storage. Performance gains are roughly one to a factor of 33 for serial and one to a factor of 100 for SMP parallel times faster using LAPACK full routines with RFPF than with using LAPACK packed routines. Existing LAPACK routines and vendor LAPACK routines were used in the serial and the SMP parallel study respectively. In the full and packed studies vendor Level 3 BLAS were used.

Categories and Subject Descriptors: G.1.3 [Numerical Analysis]: Numerical Linear Algebra – Linear Systems (symmetric and Hermitian); G.4 [Mathematics of Computing]: Mathematical Software

General Terms: Algorithms, BLAS, Performance

Additional Key Words and Phrases: real symmetric matrices, complex Hermitian matrices, positive definite matrices, Cholesky factorization and solution, recursive algorithms, novel packed matrix data structures.

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1. INTRODUCTION

A very important class of linear algebra problems deals with a coefficient matrix $A$ that is symmetric and positive definite [Dongarra et al. 1998; Demmel 1997; Golub and Van Loan 1996; Trefethen and Bau 1997]. Because of symmetry it is only necessary to store either the upper or lower triangular part of the matrix $A$.

Fig. 1. The full format array layout of an order $N$ symmetric matrix required by LAPACK. LAPACK requires $LDA \geq N$. Here we set $LDA = N = 7$.

<table>
<thead>
<tr>
<th>Lower triangular case</th>
<th>Upper triangular case</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1 8 15 22 29 36 43</td>
</tr>
<tr>
<td>2 9</td>
<td>9 16 23 30 37 44</td>
</tr>
<tr>
<td>3 10 17</td>
<td>17 24 31 38 45</td>
</tr>
<tr>
<td>4 11 18 25</td>
<td>25 32 39 46</td>
</tr>
<tr>
<td>5 12 19 26 33</td>
<td>33 40 47</td>
</tr>
<tr>
<td>6 13 20 27 34 41</td>
<td>41 48</td>
</tr>
<tr>
<td>7 14 21 28 35 42 49</td>
<td>49</td>
</tr>
</tbody>
</table>

Fig. 2. The packed format array layout of an order 7 symmetric matrix required by LAPACK.

<table>
<thead>
<tr>
<th>Lower triangular case</th>
<th>Upper triangular case</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1 2 4 7 11 16 22</td>
</tr>
<tr>
<td>2 8</td>
<td>3 5 8 12 17 23</td>
</tr>
<tr>
<td>3 9 14</td>
<td>6 9 13 18 24</td>
</tr>
<tr>
<td>4 10 15 19</td>
<td>10 14 19 25</td>
</tr>
<tr>
<td>5 11 16 20 23</td>
<td>15 20 26</td>
</tr>
<tr>
<td>6 12 17 21 24 26</td>
<td>21 27</td>
</tr>
<tr>
<td>7 13 18 22 25 27 28</td>
<td>28</td>
</tr>
</tbody>
</table>

1.1 LAPACK full and packed storage formats

The LAPACK library [Anderson et al. 1999] offers two different kinds of subroutines to solve the same problem: POTRF and PPTRF both factorize symmetric, positive definite matrices by means of the Cholesky algorithm. A major difference in these two routines is the way they access the array holding the triangular matrix (see figures 1 and 2).

In the POTRF case, the matrix is stored in one of the lower left or upper right triangles of a full square matrix ([Anderson et al. 1999, pages 139 and 140] and

1 Four names SPOTRF, DPOTRF, CPOTRF and ZPOTRF are used in LAPACK for real symmetric and complex Hermitian matrices [Anderson et al. 1999], where the first character indicates the precision and arithmetic versions: S – single precision, D – double precision, C – complex and Z – double complex. LAPACK 95 uses one name LAPOTRF for all versions [Barker et al. 2001]. In this paper, POTRF and/or PPTRF express, any precision, any arithmetic and any language version of the PO and/or PP matrix factorization algorithms.
the other triangle is wasted (see figure 1). Because of the uniform storage scheme, blocked LAPACK and Level 3 BLAS [Dongarra et al. 1990b; Dongarra et al. 1990a] subroutines can be employed, resulting in a fast solution.

In the PPTRF case, the matrix is kept in *packed* storage ([Anderson et al. 1999, pages 140 and 141], [Agarwal et al. 1994] and [IBM 1997, pages 74 and 75]), which means that the columns of the lower or upper triangle are stored consecutively in a one dimensional array (see figure 2). Now the triangular matrix occupies the strictly necessary storage space but the nonuniform storage scheme means that use of full storage BLAS is impossible and only the Level 2 BLAS [Lawson et al. 1979; Dongarra et al. 1988] packed subroutines can be employed, resulting in a slow solution.

To summarize: LAPACK offers a choice between high speed with double the waste of memory space versus low speed and no waste of memory space.

1.2 Packed Minimal Storage Data Formats related to RFPF

Recently many new data formats for matrices have been introduced for improving the performance of Dense Linear Algebra (DLA) algorithms. The survey article [Elmqvist, Elmore et al. 2004] gives an excellent overview.

**Recursive Packed Format:** [Andersen et al. 2001; Andersen et al. 2002] A new compact way to store a symmetric or triangular matrix called Recursive Packed Format (RPF) is described in [Andersen et al. 2001] as are novel ways to transform RPF to and from standard packed format. New algorithms, called Recursive Packed Cholesky (RPC) [Andersen et al. 2001; Andersen et al. 2002] that operate on the RPF format are presented here. RPF format operates almost entirely by calling Level 3 BLAS GEMM [Dongarra et al. 1990b; Dongarra et al. 1990a] but requires variants of algorithms TRSM and SYRK [Dongarra et al. 1990b; Dongarra et al. 1990a] that are designed to work on RPF. We call these algorithms RPTRSM and RPSYRK [Andersen et al. 2001] and find that they do most of their FLOPS by calling GEMM [Dongarra et al. 1990b; Dongarra et al. 1990a]. It follows that almost all of execution time of the RPC algorithm is done in calls to GEMM.

The are three advantages of this storage scheme compared to traditional packed and full storage. First, the RPF storage format uses the minimum amount of storage required for symmetric, triangular, or Hermitian matrices. Second, the RPC algorithm is a Level 3 implementation of Cholesky factorization. Finally, RPF requires no block size tuning parameter. A disadvantage of the RPC algorithm was that it had a high recursive calling overhead. The paper [Gustavson and Jonsson 2000] removed this overhead and added other novel features to the RPC algorithm.

**Square Block Packed Format (SBPF)** [Gustavson 2003]: SBPF is described in Section 4 of [Gustavson 2003]. A strong point of SBPF is that it requires minimum block storage and all its blocks are contiguous and of equal size. If one uses SBPF with kernel routines then data copying is mostly eliminated during Cholesky factorization.

**Block Packed Hybrid Format (BPHF)** [Andersen et al. 2005; Gustavson et al. 2007]: We consider an efficient implementation of the Cholesky solution of symmetric positive-definite full linear systems of equations using packed storage. We

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2 In Fortran column major, in C row major.
take the same starting point as that of LINPACK [Dongarra et al. 1979] and LAPACK [Anderson et al. 1999], with the upper (or lower) triangular part of the matrix being stored by columns. Following LINPACK [Dongarra et al. 1979] and LAPACK [Anderson et al. 1999], we overwrite the given matrix by its Cholesky factor. The paper [Andersen et al. 2005] uses the BPHF where blocks of the matrix are held contiguously. The paper compares BPHF versus conventional full format storage, packed format and the RPF for the algorithms its consider in its timing studies. BPF is a variant of SBPF in which the diagonal blocks are stored in packed format and so its storage requirement is equal to that of packed storage.

We mention that for packed matrices SBPF and BPHF have become the format of choice for multi-core processors when one stores the blocks in register block format [Gustavson et al. 2006]. Recently, there have been many papers published on new algorithms for multi-core processors. This literature is extensive. So, we only mention two projects, PLASMA [Buttari et al. 2007] and FLAME [Chan et al. 2007], and refer the interested reader to the literature for additional references.

In regard to other references on NDS, the survey article [Elmroth et al. 2004] gives an excellent overview. However, since 2005 at least two new data formats for Cholesky type factorization have emerged, [Herrero 2006] and the subject matter of this paper, RFPF [Gustavson and Wasniewski 2006]. In the next sub-section we highlight the main features of the RFPF.

1.3 A novel way of representing symmetric, triangular and Hermitian matrices in LAPACK

LAPACK has two types of subroutines for symmetric, triangular, and Hermitian matrices called packed and full format routines. LAPACK has about 300 these kind of subroutines. So, in either format, a variety of problems can be solved by these LAPACK subroutines. The RFPF can replace both these LAPACK data formats. Furthermore, and this is important, using RFPF does not require any new LAPACK subroutines to be written. Using RFPF in LAPACK only requires the use of already existing LAPACK and BLAS routines.

1.4 Overview of the Paper

First we introduce the RFPF in general, see section 2. Secondly we show how to use RFPF on symmetric and Hermitian positive definite matrices; e.g., for the factorization (section 3), solution (section 4), and inversion (section 5) of these matrices. Section 6 describes LAPACK subroutines for the factorization, solution, and inversion of symmetric and Hermitian positive definite matrices using RFPF. Section 7 indicates that the stability results of using RFPF is unaffected by this format choice as RFPF uses existing LAPACK algorithms which are already known to be stable. Section 8 describes a variety of performance results on commonly used platforms both for serial and parallel SMP execution. These results show that performance of LAPACK full routines using RFPF versus LAPACK full routines using standard format for both serial and SMP parallel processing is about the same while using half the storage. Also, performance gains are roughly one to a factor of 33 for serial and one to a factor of 100 for SMP parallel times faster using LAPACK full routines with RFPF than with using LAPACK packed routines. Existing LAPACK routines and vendor LAPACK routines were used in the serial
and the SMP parallel study respectively. In the full and packed studies vendor Level 3 BLAS were used. Section 9 gives a short summary and brief conclusions.

LAPACK software using the RFPF is already written and well tested. In particular, it has past all the LAPACK test for Cholesky and positive definite Hermitian routines. It will be separately announced by [Gustavson et al. 2007].

2. DESCRIPTION OF RECTANGULAR FULL PACKED FORMAT

We describe Rectangular Full Packed Format (RFPF). It transforms a standard Packed Array $\mathbf{A}_P$ of size $NT = N(N + 1)/2$ to a full 2D array. This means that performance of LAPACK’s [Anderson et al. 1999] packed format routines becomes equal to or better than their full array counterparts. RFPF is a variant of Hybrid Full Packed (HFP) format [Gunnels and Gustavson 2004]. RFPF is a rearrangement of a Standard full format rectangular Array $\mathbf{A}_S$ of size $LDA\times N$ where $LDA \geq N$.

Array $\mathbf{A}_S$ holds a triangular part of a symmetric, triangular, or Hermitian matrix $\mathbf{A}$ of order $N$. The rearrangement of array $\mathbf{A}_S$ is equal to compact full format Rectangular Array $\mathbf{A}_R$ of size $LDA \times N1 = NT$ and hence array $\mathbf{A}_R$ like array $\mathbf{A}_P$ uses minimal storage. Array $\mathbf{A}_R$ will hold a full rectangular matrix $\mathbf{A}_R$ obtained from a triangle of matrix $\mathbf{A}$. Note also that the transpose of the rectangular matrix $\mathbf{A}_R^T$ resides in the transpose of array $\mathbf{A}_R$ and hence also represents $\mathbf{A}$. Therefore, Level 3 BLAS [Dongarra et al. 1990b; Dongarra et al. 1990a] can be used on array $\mathbf{A}_R$ or its transpose. In fact, with the equivalent LAPACK algorithm which uses the array $\mathbf{A}_R$ or its transpose, the performance is slightly better than standard LAPACK algorithm which uses the array $\mathbf{A}_S$ or its transpose. Therefore, this offers the possibility to replace all packed or full LAPACK routines with equivalent LAPACK routines that work on array $\mathbf{A}_R$ or its transpose. For examples of transformations of a matrix $\mathbf{A}$ to a matrix $\mathbf{A}_R$ see the figures in Section 6.

RFPF is closely related to HFP format, see [Gunnels and Gustavson 2004], which represents $\mathbf{A}$ as the concatenation of two standard full arrays whose total size is also $NT$. A basic simple idea leads to both formats. Let $\mathbf{A}$ be an order $N$ symmetric matrix. Break $\mathbf{A}$ into a block $2 \times 2$ form

\[
\mathbf{A} = \begin{bmatrix}
A_{11} & A_{21}^T \\
A_{21} & A_{22}
\end{bmatrix}
\quad \text{or} \quad
\mathbf{A} = \begin{bmatrix}
A_{11} & A_{12} \\
A_{21}^T & A_{22}
\end{bmatrix}
\]

(1)

where $A_{11}$ and $A_{22}$ are symmetric. Clearly, we need only store the lower triangles of $A_{11}$ and $A_{22}$ as well as the full matrix $A_{21} = A_{21}^T$ when we are interested in a lower triangular formulation.

When $N = 2k$ is even, the lower triangle of $A_{11}$ and the upper triangle of $A_{22}^T$ can be concatenated together along their main diagonals into a $(k + 1) \times k$ dense matrix (see the figures where $N$ is even in Section 6). This last operation is the crux of the basic simple idea. The off-diagonal block $A_{21}$ is $k \times k$, and so it can be appended below the $(k+1) \times k$ dense matrix. Thus, the lower triangle of $\mathbf{A}$ can be stored as a single $(n+1) \times k$ dense matrix $\mathbf{A}_R$. In effect, each block matrix $A_{11}$, $A_{21}$ and $A_{22}$ is now stored in ‘full format.’ This means all entries of matrix $\mathbf{A}_R$ in array $\mathbf{A}_R$ of size $LDA = n+1$ by $N1 = k$ can be accessed with constant row and column strides. So, the full power of LAPACK’s block Level 3 codes are now available for RFPF which uses the minimum amount of storage. Finally, matrix $\mathbf{A}_R^T$ which has
size $k \times (N + 1)$ is represented in the transpose of array $A_R$ and hence has the same desirable properties. There are eight representations of RFPF. The matrix $A$ can have either odd or even order $N$, or it can be represented either in standard lower or upper format or it can be represented by either matrix $A_A$ or its transpose $A_R^T$ giving $2^3 = 8$ representations in all.

All eight cases or representations are presented in Section 6. The RFPF matrices are in the upper right part of the figures. We have introduced colors and horizontal lines to try to visually delineate triangles $T_1$, $T_2$ representing lower, upper triangles of symmetric matrices $A_{11}$, $A_{22}$ respectively and square or near square $S_1$ representing matrices $A_{21}$. For an upper triangle of $A$, $T_1$, $T_2$ represents lower, upper triangles of symmetric matrices $A_{11}^T$, $A_{22}$ respectively and square or near square $S_1$ representing matrices $A_{12}$. For both lower and upper triangles of $A$ we have, after each $a_{i,j}$, added its position location in the arrays holding matrices $A$ and $A_R$.

We now consider performance aspects of using RFPF in the context of using LAPACK routines on triangular matrices stored in RFPF. Let $X$ be a Level 3 LAPACK routine that operates either on standard packed or full format. $X$ has a full Level 3 LAPACK block $2 \times 2$ algorithm, call it $FX$. We write a simple related partition algorithm (SRPA) with partition sizes $n1$ and $n2$ where $n1 + n2 = N$. Apply the new SRPA using the new RFPF. The new SRPA almost always has four major steps consisting entirely of calls to existing full format LAPACK routines in two steps and calls to Level 3 BLAS in the remaining two steps, see fig. 3.

![Fig. 3. Simple related partition algorithm (SRPA) of RFPF](image)

Section 6 shows $FX$ algorithms equal to factorization, solution and inversion algorithms on symmetric positive definite or Hermitian matrices.

3. CHOLESKY FACTORIZATION USING RECTANGULAR FULL PACKED FORMAT

The Cholesky factorization of a symmetric and positive definite matrix $A$ can be expressed as

$$A = LL^T \text{ or } A = U^TU \text{ (in the symmetric case)}$$
$$A = LL^H \text{ or } A = U^HU \text{ (in the Hermitian case)}$$

(2)

where $L$ and $U$ are lower triangular and upper triangular matrices.

Break the matrices $L$ and $U$ into $2 \times 2$ block form in the same way as was done for the matrix $A$ in equation (1):

$$L = \begin{bmatrix} L_{11} & 0 \\ L_{21} & L_{22} \end{bmatrix} \text{ and } U = \begin{bmatrix} U_{11} & U_{12} \\ 0 & U_{22} \end{bmatrix}$$

(3)

We now have
Rectangular Full Packed Data Format (Cholesky Algorithm)  

\[ LL^T = \begin{bmatrix} L_{11} & 0 \\ L_{21} & L_{22} \end{bmatrix} \begin{bmatrix} L_{11}^T & L_{12}^T \\ 0 & L_{22}^T \end{bmatrix} \quad \text{and} \quad U^T U = \begin{bmatrix} U_{11}^T & 0 \\ U_{12}^T & U_{22} \end{bmatrix} \begin{bmatrix} U_{11} & U_{12} \\ 0 & U_{22} \end{bmatrix} \]

(4)

and the Hermitian case:

\[ LH^T = \begin{bmatrix} L_{11} & 0 \\ L_{21} & L_{22} \end{bmatrix} \begin{bmatrix} L_{11}^H & L_{12}^H \\ 0 & L_{22}^H \end{bmatrix} \quad \text{and} \quad U^H U = \begin{bmatrix} U_{11}^H & 0 \\ U_{12}^H & U_{22} \end{bmatrix} \begin{bmatrix} U_{11} & U_{12} \\ 0 & U_{22} \end{bmatrix} \]

where \( L_{11}, L_{22}, U_{11}, \) and \( U_{22} \) are lower and upper triangular submatrices, and \( L_{21} \) and \( U_{12} \) are square or almost square submatrices. The block values of \( L \) or \( U \) are stored over the of \( A \).

Using equations (2) and equating the blocks of equations (1) and equations (4) gives us the basis of a \( 2 \times 2 \) block algorithm for Cholesky factorization using RFPF. We can now express each of these four block equalities by calls to existing LAPACK and Level 3 BLAS routines. An example, see Section 6, of this is the three block equations is \( L_{11}L_{11}^T = A_{11}, L_{21}L_{11}^T = A_{21} \) and \( L_{21}L_{21}^T + L_{22}L_{22} = A_{22} \). The first and second of these block equations are handled by calling LAPACK’s POTRF routine \( L_{11} \leftarrow A_{11} \) and by calling Level 3 BLAS TRSM via \( L_{21} \leftarrow L_{21}L_{11}^T \). In both these block equations the Fortran equality of replacement (\( \leftarrow \)) is being used so that the lower triangle of \( A_{11} \) is being replaced \( L_{11} \) and the nearly square matrix \( A_{21} \) is being replaced by \( L_{21} \). The third block equation breaks into two parts: \( A_{22} \leftarrow L_{21}L_{21}^T \) and \( L_{22} \leftarrow A_{22} \) which are handled by calling Level 3 BLAS SYRK or HERK and by calling LAPACK’s POTRF routine. At this point we can use the flexibility of the LAPACK library. In RFPF \( A_{22} \) is in upper format (upper triangle) while in standard format \( A_{22} \) is in lower format (lower triangle). Due to symmetry, both formats of \( A_{22} \) contain equal values. This flexibility allows LAPACK to accommodate both formats. Hence, in the calls to SYRK or HERK and POTRF we set uplo = 'U' even though the rectangular matrix of SYRK and HERK comes from a lower triangular formulation.

New LAPACK like routine PFTRF [Gustavson et al. 2007] performs these four computations. PF stands for Positive Full, and was chosen to fit with LAPACK’s use of PO and PP. The PF routine covers the Cholesky Factorization algorithm for the eight cases of the RFPF. Section 6 has a figure 4 with four subfigures. Here we are interested here in the first and second subfigure. The first subfigure contains the layouts of matrices \( A \) and \( A_R \). The second subfigure has the Cholesky factorization algorithm obtained by simple algebraic manipulations of the three block equalities obtained above.

4. SOLUTION

In Section 3 we obtained the \( 2 \times 2 \) Cholesky factorization (3) of matrix \( A \). Now, we can solve the equation \( AX = B \):

- If \( A \) has lower triangular format then
  \[ LY = B \quad \text{and} \quad L^T X = Y \quad \text{(in the symmetric case)} \]
  \[ LY = B \quad \text{and} \quad L^H X = Y \quad \text{(in the Hermitian case)} \]

(5)
If $A$ has an upper triangular format then
\[
U^T Y = B \text{ and } UX = Y \text{ (in the symmetric case)}
\]
\[
U^H Y = B \text{ and } UX = Y \text{ (in the Hermitian case)}
\]  

(6)

$B$, $X$ and $Y$ are either vectors or rectangular matrices. $B$ contains the RHS values. $X$ and $Y$ contain the solution values. $B$, $X$ and $Y$ are vectors when there is one RHS and matrices when there are many RHS. The values of $X$ and $Y$ are stored over the values of $B$.

Expanding (5) and (6) using (3) gives the forward substitution equations

in the symmetric case:
\[
\begin{bmatrix}
L_{11} & 0 \\
L_{21} & L_{22}
\end{bmatrix}
\begin{bmatrix}
Y_1 \\
Y_2
\end{bmatrix}
= \begin{bmatrix}
B_1 \\
B_2
\end{bmatrix}
\quad \text{and} \quad
\begin{bmatrix}
U_{11}^T & 0 \\
U_{12}^T & U_{22}^T
\end{bmatrix}
\begin{bmatrix}
Y_1 \\
Y_2
\end{bmatrix}
= \begin{bmatrix}
B_1 \\
B_2
\end{bmatrix}
\]

(7)

and in the Hermitian case:
\[
\begin{bmatrix}
L_{11} & 0 \\
L_{21} & L_{22}
\end{bmatrix}
\begin{bmatrix}
Y_1 \\
Y_2
\end{bmatrix}
= \begin{bmatrix}
B_1 \\
B_2
\end{bmatrix}
\quad \text{and} \quad
\begin{bmatrix}
U_{11}^H & 0 \\
U_{12}^H & U_{22}^H
\end{bmatrix}
\begin{bmatrix}
Y_1 \\
Y_2
\end{bmatrix}
= \begin{bmatrix}
B_1 \\
B_2
\end{bmatrix}
\]

and the back substitution equations

in the symmetric case:
\[
\begin{bmatrix}
L_{11}^T & L_{21}^T \\
0 & L_{22}^T
\end{bmatrix}
\begin{bmatrix}
X_1 \\
X_2
\end{bmatrix}
= \begin{bmatrix}
Y_1 \\
Y_2
\end{bmatrix}
\quad \text{and} \quad
\begin{bmatrix}
U_{11} & U_{12} \\
0 & U_{22}
\end{bmatrix}
\begin{bmatrix}
X_1 \\
X_2
\end{bmatrix}
= \begin{bmatrix}
Y_1 \\
Y_2
\end{bmatrix}
\]

(8)

and in the Hermitian case:
\[
\begin{bmatrix}
L_{11}^H & L_{21}^H \\
0 & L_{22}^H
\end{bmatrix}
\begin{bmatrix}
X_1 \\
X_2
\end{bmatrix}
= \begin{bmatrix}
Y_1 \\
Y_2
\end{bmatrix}
\quad \text{and} \quad
\begin{bmatrix}
U_{11} & U_{12} \\
0 & U_{22}
\end{bmatrix}
\begin{bmatrix}
X_1 \\
X_2
\end{bmatrix}
= \begin{bmatrix}
Y_1 \\
Y_2
\end{bmatrix}
\]

The equations (7) and (8) give the basis of a $2 \times 2$ block algorithm for Cholesky solution using RFFP format. We can now express these two sets of two block equalities by using existing Level 3 BLAS routines. An example, see Section 6, of the first set of these two block equalities is $L_{11}Y_1 = B_1$ and $L_{21}Y_1 + L_{22}Y_2 = B_2$. The first block equality is handled by Level 3 BLAS TRSM: $Y_1 \leftarrow L_{11}^{-1} B_1$. The second block equality is handled by Level 3 BLAS GEMM and TRSM: $B_2 \leftarrow B_2 - L_{21} Y_1$ and $Y_2 \leftarrow L_{22}^{-1} Y_2$. The backsubstitution routines are similarly derived. One gets $X_2 \leftarrow L_{22}^{-T} Y_2$, $Y_1 \leftarrow Y_1 - L_{21}^T X_2$ and $X_1 \leftarrow L_{11}^{-T} Y_1$.

New LAPACK like routine PFTRF performs these two solution computations for the eight cases of RFFP. PFTRF calls a new Level 3 BLAS TFTRSM in the same way that POTRS calls TRSM. The third subfigure in Section 6 gives the Cholesky solution algorithm using RFFP obtained by simple algebraic manipulation of the block equations (7) and (8).

5. INVERSION

Following LAPACK we consider the following three-stage procedure:

(1) Factorize the matrix $A$ and overwrite $A$ with either $L$ or $U$ by calling PFTRF; see section 3.

(2) Compute the inverse of either $L$ or $U$. Call these matrices $W$ or $V$ and overwrite either $L$ or $U$ with them. This is done by calling new routine new LAPACK like TFTRF or HFTRF.
(3) Calculate either the product $W^T W$ or $V V^T$ and overwrite either $W$ or $V$ with them.

As in Sections 3 and 4 we examine 2 by 2 block algorithms for the steps two and three above. In Section 3 we obtain either matrices $L$ or $U$ in RFPF. Like LAPACK inversion algorithms for POTRI and PPTRI, this is our starting point for our LAPACK inversion algorithm using RFPF. The LAPACK inversion algorithms for POTRI and PPTRI also follow from steps two and three above by first calling in the full case LAPACK TRTRI and then calling LAPACK LAUUM.

Take the inverse of equation (2) and obtain

\[ A^{-1} = W^T W \quad \text{or} \quad A^{-1} = V V^T \quad \text{(in the symmetric case)} \]
\[ A^{-1} = W^H W \quad \text{or} \quad A^{-1} = V V^H \quad \text{(in the Hermitian case)} \]

where $W$ and $V$ are lower and upper triangular matrices.

Using the 2 by 2 blocking for either $L$ or $U$ in equation (3) we obtain the following 2 by 2 blocking for $W$ and $V$:

\[
W = \begin{bmatrix} W_{11} & 0 \\ W_{21} & W_{22} \end{bmatrix} \quad \text{and} \quad V = \begin{bmatrix} V_{11} & V_{12} \\ 0 & V_{22} \end{bmatrix}
\] (10)

From the identities $WL = LW = I$ and $VU = UV = I$ and the 2 by 2 block layouts of equations (3) and 3 we obtain three block equations for $W$ and $V$ which can be solved using LAPACK routines for TRTRI and Level 3 BLAS TRMM. An example, see Figure 4, of these three block equations is $L_{11} W_{11} = I$, $L_{21} W_{11} + L_{22} W_{21} = 0$ and $L_{22} W_{22} = I$. The first and third of these block equations are handled by LAPACK TRTRI routines as $W_{11} \leftarrow L_{11}^{-1}$ and $V_{22} \leftarrow U_{22}^{-1}$. In the second inverse computation we use the fact that $L_{22}$ is equally represented by it transpose $L_{22}^T$ which is $U_{22}$ in RFPF. The second block equation leads to two calls to Level 3 BLAS TRMM via $L_{21} \leftarrow -L_{21} W_{11}$ and $W_{21} = W_{22} L_{21}$. In the last two block equations the Fortran equality of replacement (\leftarrow) is being used so that $W_{21} = -W_{22} L_{21} W_{11}$ is replacing $L_{21}$.

Now we turn to part three of the three stage LAPACK procedure above. For this we use the 2 by 2 blocks layouts of equation (10) and the matrix multiplications indicated by following block equations (11) giving

\[
W^T W = \begin{bmatrix} W_{11}^T & W_{21}^T \\ 0 & W_{22}^T \end{bmatrix} \begin{bmatrix} W_{11} & 0 \\ W_{21} & W_{22} \end{bmatrix} \quad \text{and} \quad V V^T = \begin{bmatrix} V_{11} & V_{12} \\ 0 & V_{22} \end{bmatrix} \begin{bmatrix} V_{11}^T & 0 \\ V_{12} & V_{22} \end{bmatrix}
\] (11)

and the Hermitian case:

\[
W^H W = \begin{bmatrix} W_{11}^H & W_{21}^H \\ 0 & W_{22}^H \end{bmatrix} \begin{bmatrix} W_{11} & 0 \\ W_{21} & W_{22} \end{bmatrix} \quad \text{and} \quad V V^H = \begin{bmatrix} V_{11}^H & V_{12} \\ 0 & V_{22} \end{bmatrix} \begin{bmatrix} V_{11} & V_{12} \\ V_{12}^H & V_{22} \end{bmatrix}
\]

where $W_{11}$, $W_{22}$, $V_{11}$, and $V_{22}$ are lower and upper triangular submatrices, and $W_{21}$ and $V_{12}$ are square or almost square submatrices. The values of the indicated block multiplications of $W$ or $V$ in equation (11) are stored over the block values of $W$ or $V$.

Performing the indicated 2 by 2 block multiplications of equation (11) leads to three block matrix computations. An example, see Section 6, of these three block
computations is \(W_{11}^T W_{11} + W_{21}^T W_{21}, W_{21}^T W_{21}\) and \(W_{22}^T W_{22}\). Additionally, we want to overwrite the values of these block multiplications on their original block operands. Block operand \(W_{11}\) only occurs in the 11 block operand computation and hence can be overwritten by a call to LAPACK LAUUM, \(W_{11} \leftarrow W_{11}^T W_{11}\), followed by a call to Level 3 BLAS SYRK or HERK, \(W_{11} \leftarrow W_{11} + W_{21}^T W_{21}\). Block operand \(W_{21}\) now only occurs in the 21 block computation and hence can be overwritten by a call to Level 3 BLAS TRMM, \(W_{21} \leftarrow W_{21}^T W_{21}\). Finally, block operand \(W_{22}\) can be overwritten by a call to LAPACK LAUUM, \(W_{22} \leftarrow W_{22}^T W_{22}\).

The fourth subfigure in Section 6 has the Cholesky inversion algorithms using RFPF based on the results of this Section. New LAPACK routine, PFTRI, performs this computation for the eight cases of RFPF.

6. RFP DATA FORMATS AND ALGORITHMS

This section contains three figures.

(1) The first figure describes the RFPF (Rectangular Full Packed Format) and gives algorithms for Cholesky factorization, solution and inversion of symmetric positive definite matrices, where \(N\) is odd, uplo = 'lower', and 'trans' = 'no transpose'. This figure has four subfigures.

(a) The first subfigure depicts the lower triangle of a symmetric positive definite matrix \(A\) in standard full and its representation by the matrix \(A_R\) in RFPF.

(b) The second subfigure gives the RFPF factorization algorithm and its calling sequences of the LAPACK and BLAS subroutines, see Section 3.

(c) The third subfigure gives the RFPF solution algorithm and its calling sequences to the LAPACK and BLAS subroutines, see Section 4.

(d) The fourth subfigure in each figure gives the RFPF inversion algorithm and its calling sequences to the LAPACK and BLAS subroutines, see Section 5.

(2) The second figure shows the transformation from full to RFPF of all "no transform" cases.

(3) The third figure depicts all eight cases in RFPF.

The data format for \(A\) has \(LDA = N\). Matrix \(A_R\) has \(LDAR = N\) if \(N\) is odd and \(LDAR = N + 1\) if \(N\) is even and \(n_1 = \lfloor N/2 \rfloor\). Hence, matrix \(A_R\) always has \(LDAR\) rows and \(n_1\) columns. Matrix \(A_R^T\) always has \(n_1\) rows and \(LDAR\) columns and its leading dimension is equal to \(n_1\). Matrix \(A_R\) always has \(LDAR \times n_1 = NT = N(N + 1)/2\) elements as does matrix \(A_R^T\).

The order \(N\) of matrix \(A\) in the first figure is seven and six or seven in the remaining two figures.

7. STABILITY OF THE RFPF ALGORITHM

The algebraic manipulation, how we developed the RFPF factorization (Section 3), solution (Section 4), and inversion (Section 5) algorithms are equivalent to the traditional algorithms in the books [Dongarra et al. 1998; Demmel 1997; Golub and Van Loan 1996; Trefethen and Bau 1997]. The whole theory of the traditional Cholesky factorization, solution, inversion and BLAS algorithms carries over to this three Cholesky and BLAS algorithms described in Sections 3, 4, and 5. The
Fig. 4. The Cholesky factorization algorithm using the Rectangular Full Packed Format (RFPPF) if \( N \) is odd, uplo='lower', and trans='no transpose'.

\[
A_{RF} \text{ of Rectangular full packed}
\begin{align*}
\text{LDA} & = N, \text{memory needed} \\
\text{LDA} \times N & = 28 \\
\begin{pmatrix}
\begin{array}{cccccc}
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}
\end{pmatrix}
\end{align*}
\]

Matrix \( A_R \)

\[
A_{RF} \text{ of Rectangular full packed}
\begin{align*}
\text{LDA} & = N, \text{memory needed} \\
\text{LDA} \times N & = 28 \\
\begin{pmatrix}
\begin{array}{cccccc}
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}
\end{pmatrix}
\end{align*}
\]

Matrix \( A_R \)

Factorization Algorithm

1) \textbf{factor} \( L_{11}^T = A_{11} \); \\
\hspace{1em} call \textsc{potrf}('L', n1, AR, N, & \\
\hspace{2em} info);

2) \textbf{solve} \( L_{11}L_{11}^T = A_{11} \); \\
\hspace{1em} call \textsc{trsm}('R', 'L', 'T', 'N', n1, & \\
\hspace{2em} N', n2, & \\
\hspace{3em} AR(n1 + 1, 1), N);

3) \textbf{update} \( A_{22} := A_{22} - L_{11}L_{11}^T \); \\
\hspace{1em} call \textsc{syrk}/\textsc{herk}('U', 'N', n2, n1, & \\
\hspace{2em} AR(n1 + 1, 1), N, one, AR(1, 2), N);

Solution Algorithm,

where \( B(LDB, nr) \) and \( LDB \geq N \) (here \( LDB = N \)):

\[
\begin{align*}
LY & = B \\
L'X & = Y
\end{align*}
\]

1) \textbf{solve} \( L_{11}^T X_1 = Y \); \\
\hspace{1em} call \textsc{trsm}('U', 'U', 'N', n2, & \\
\hspace{2em} N', n1, & \\
\hspace{3em} N, one, AR(1, 2), N);

2) \textbf{Multiply} \( B_2 = B_2 - L_{21}Y_1 \); \\
\hspace{1em} call \textsc{gemm}('N', 'N', n2, n1, & \\
\hspace{2em} one, AR(n1 + 1, 1), N, B, N, one, & \\
\hspace{3em} AR(n1 + 1, 1), N);

3) \textbf{solve} \( L_{22} X_2 = B_2 \); \\
\hspace{1em} call \textsc{trsm}('L', 'L', 'T', 'N', n2, & \\
\hspace{2em} N, n1, & \\
\hspace{3em} one, AR(1, 2), N, B(n1 + 1, 1), N);

The Inversion Algorithm:

1) \textbf{invert} \( W_{11} = L_{11}^{-1} \); \\
\hspace{1em} call \textsc{trtri}('U', 'N', n1, AR, N, & \\
\hspace{2em} info);

2) \textbf{Multiply} \( W_{21} = -L_{21}W_{11} \); \\
\hspace{1em} call \textsc{trmm}('R', 'L', 'T', 'N', n2, & \\
\hspace{2em} n1, & \\
\hspace{3em} one, AR(n1 + 1, 1), N);

3) \textbf{invert} \( V_{22} = U_{22}^{-1} \); \\
\hspace{1em} call \textsc{trtri}('U', 'N', n2, AR(1, 2), & \\
\hspace{2em} N, info);

4) \textbf{invert} \( V_{22} = U_{22}^{-1} \); \\
\hspace{1em} call \textsc{trmm}('L', 'U', 'T', 'N', n2, & \\
\hspace{2em} n1, & \\
\hspace{3em} one, AR(1, 2), N, AR(n1 + 1, 1), N);

\[
\begin{align*}
\text{Triangular matrix multiplication}
\end{align*}
\]

1) \textbf{Triang. Mult.} \( W_{11} = W_{11} W_{11} \); \\
\hspace{1em} call \textsc{lauum}('L', n1, AR, N, & \\
\hspace{2em} info);

2) \textbf{update} \( W_{11} = W_{11} + W_{21}W_{21} \); \\
\hspace{1em} call \textsc{syrk}/\textsc{herk}('L', 'T', n1, & \\
\hspace{2em} AR(n1 + 1, 1), N, one, & \\
\hspace{3em} AR(n1 + 1, 1), N);

3) \textbf{Multiply} \( W_{21} = V_{21}W_{21} \); \\
\hspace{1em} call \textsc{trmm}('L', 'U', 'T', 'N', n2, & \\
\hspace{2em} n1, & \\
\hspace{3em} one, AR(1, 2), N, AR(n1 + 1, 1), N);

4) \textbf{Triang. Mult.} \( V_{11} = V_{11} V_{11} \); \\
\hspace{1em} call \textsc{lauum}('U', n2, AR(1, 2), N, & \\
\hspace{2em} info);
Fig. 5. Eight two-dimensional arrays for storing the matrices $A$ and $A_R$ that are needed by the LAPACK subroutine POTRF (full format) and PFTRF RFPF respectively. The leading dimension LDA is $N$ for LAPACK, and LDAR for RFPF. $LDAR = N$ for $N$ odd, and $N + 1$ for $N$ even. Here $N$ is 7 or 6. The memory needed is $LDA \times N$ for full format and $LDAR \times n1 = (N + 1)N/2$ for RFPF. Here 49 and 36 for full format and 28 and 21 for RFPF. The column size of RFPF is $n1 = \lceil N/2 \rceil$, here 4 and 3.

5.1 The matrices $A$ of order $N$ and $A_R$ of size $LDAR$ by $n1$, here $N = 7$.

5.1.1 Full Format

\[
\begin{bmatrix}
  a_{1,1} & 0 & 0 & 0 & 0 & 0 & 0 \\
  a_{2,1} & a_{2,2} & 0 & 0 & 0 & 0 & 0 \\
  a_{3,1} & a_{3,2} & a_{3,3} & 0 & 0 & 0 & 0 \\
  a_{4,1} & a_{4,2} & a_{4,3} & a_{4,4} & 0 & 0 & 0 \\
  a_{5,1} & a_{5,2} & a_{5,3} & a_{5,4} & a_{5,5} & 0 & 0 \\
  a_{6,1} & a_{6,2} & a_{6,3} & a_{6,4} & a_{6,5} & a_{6,6} & 0 \\
  a_{7,1} & a_{7,2} & a_{7,3} & a_{7,4} & a_{7,5} & a_{7,6} & a_{7,7}
\end{bmatrix}
\]

5.1.2 RFPF

\[
\begin{bmatrix}
  a_{1,1} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{2,1} & a_{2,2} & a_{6,6} & a_{7,7} \\
  a_{3,1} & a_{3,2} & a_{3,3} & a_{7,7} \\
  a_{4,1} & a_{4,2} & a_{4,3} & a_{4,4} \\
  a_{5,1} & a_{5,2} & a_{5,3} & a_{5,4} \\
  a_{6,1} & a_{6,2} & a_{6,3} & a_{6,4} \\
  a_{7,1} & a_{7,2} & a_{7,3} & a_{7,4}
\end{bmatrix}
\]

5.2 The matrices $A$ of order $N$ and $A_R$ of size $LDAR$ by $n1$, here $N = 6$.

5.2.1 Full format

\[
\begin{bmatrix}
  a_{1,1} & 0 & 0 & 0 & 0 & 0 \\
  a_{2,1} & a_{2,2} & 0 & 0 & 0 & 0 \\
  a_{3,1} & a_{3,2} & a_{3,3} & 0 & 0 & 0 \\
  a_{4,1} & a_{4,2} & a_{4,3} & a_{4,4} & 0 & 0 \\
  a_{5,1} & a_{5,2} & a_{5,3} & a_{5,4} & a_{5,5} & 0 \\
  a_{6,1} & a_{6,2} & a_{6,3} & a_{6,4} & a_{6,5} & a_{6,6}
\end{bmatrix}
\]

5.2.2 RFPF

\[
\begin{bmatrix}
  a_{1,1} & a_{5,5} & a_{6,6} \\
  a_{2,1} & a_{2,2} & a_{6,6} \\
  a_{3,1} & a_{3,2} & a_{3,3} \\
  a_{4,1} & a_{4,2} & a_{4,3} \\
  a_{5,1} & a_{5,2} & a_{5,3} \\
  a_{6,1} & a_{6,2} & a_{6,3}
\end{bmatrix}
\]

5.3 The matrices $A$ of order $N$ and $A_R$ of size $LDAR$ by $n1$, here $N = 7$.

5.3.1 Full format

\[
\begin{bmatrix}
  a_{1,1} & a_{1,2} & a_{1,3} & a_{1,4} & a_{1,5} & a_{1,6} & a_{1,7} \\
  a_{2,1} & a_{2,2} & a_{2,3} & a_{2,4} & a_{2,5} & a_{2,6} & a_{2,7} \\
  a_{3,1} & a_{3,2} & a_{3,3} & a_{3,4} & a_{3,5} & a_{3,6} & a_{3,7} \\
  a_{4,1} & a_{4,2} & a_{4,3} & a_{4,4} & a_{4,5} & a_{4,6} & a_{4,7} \\
  a_{5,1} & a_{5,2} & a_{5,3} & a_{5,4} & a_{5,5} & a_{5,6} & a_{5,7} \\
  a_{6,1} & a_{6,2} & a_{6,3} & a_{6,4} & a_{6,5} & a_{6,6} & a_{6,7} \\
  a_{7,1} & a_{7,2} & a_{7,3} & a_{7,4} & a_{7,5} & a_{7,6} & a_{7,7}
\end{bmatrix}
\]

5.3.2 RFPF

\[
\begin{bmatrix}
  a_{1,1} & a_{1,2} & a_{1,3} & a_{1,4} & a_{1,5} & a_{1,6} & a_{1,7} \\
  a_{2,1} & a_{2,2} & a_{2,3} & a_{2,4} & a_{2,5} & a_{2,6} & a_{2,7} \\
  a_{3,1} & a_{3,2} & a_{3,3} & a_{3,4} & a_{3,5} & a_{3,6} & a_{3,7} \\
  a_{4,1} & a_{4,2} & a_{4,3} & a_{4,4} & a_{4,5} & a_{4,6} & a_{4,7} \\
  a_{5,1} & a_{5,2} & a_{5,3} & a_{5,4} & a_{5,5} & a_{5,6} & a_{5,7} \\
  a_{6,1} & a_{6,2} & a_{6,3} & a_{6,4} & a_{6,5} & a_{6,6} & a_{6,7} \\
  a_{7,1} & a_{7,2} & a_{7,3} & a_{7,4} & a_{7,5} & a_{7,6} & a_{7,7}
\end{bmatrix}
\]

5.4 The matrices $A$ of order $N$ and $A_R$ of size $LDAR$ by $n1$, here $N = 6$.

5.4.1 Full format

\[
\begin{bmatrix}
  a_{1,1} & a_{1,2} & a_{1,3} & a_{1,4} & a_{1,5} & a_{1,6} \\
  a_{2,1} & a_{2,2} & a_{2,3} & a_{2,4} & a_{2,5} & a_{2,6} \\
  a_{3,1} & a_{3,2} & a_{3,3} & a_{3,4} & a_{3,5} & a_{3,6} \\
  a_{4,1} & a_{4,2} & a_{4,3} & a_{4,4} & a_{4,5} & a_{4,6} \\
  a_{5,1} & a_{5,2} & a_{5,3} & a_{5,4} & a_{5,5} & a_{5,6} \\
  a_{6,1} & a_{6,2} & a_{6,3} & a_{6,4} & a_{6,5} & a_{6,6}
\end{bmatrix}
\]

5.4.2 RFPF

\[
\begin{bmatrix}
  a_{1,1} & a_{5,5} & a_{6,6} \\
  a_{2,1} & a_{2,2} & a_{6,6} \\
  a_{3,1} & a_{3,2} & a_{3,3} \\
  a_{4,1} & a_{4,2} & a_{4,3} \\
  a_{5,1} & a_{5,2} & a_{5,3} \\
  a_{6,1} & a_{6,2} & a_{6,3}
\end{bmatrix}
\]
Fig. 6. Eight two-dimensional arrays for storing the matrices \( A_R \) and \( A_R^T \) in RFPF. The leading dimension \( LDAR \) of \( A_R \) is \( N \) when \( N \) is odd and \( N + 1 \) when \( N \) is even. For the matrix \( A_R^T \) it is \( n1 = \lceil N/2 \rceil \). The memory needed for both \( A_R \) and \( A_R^T \) is \( (N + 1)/2 \times N \). This amount is 28 for \( N = 7 \) and 21 for \( N = 6 \).

6.1 RFPF for the matrices of rank odd, here \( N = 7 \) and \( n1 = 4 \)

**Lower triangular**

\[
LDAR = N
\]

\[
\begin{bmatrix}
  a_{1,1} & a_{2,2} & a_{3,3} & a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{2,1} & a_{3,2} & a_{4,3} & a_{5,4} & a_{6,5} & a_{7,6} \\
  a_{3,1} & a_{4,2} & a_{5,3} & a_{6,4} & a_{7,5} \\
  a_{4,1} & a_{5,2} & a_{6,3} & a_{7,4} \\
  a_{5,1} & a_{6,2} & a_{7,3} \\
  a_{6,1} & a_{7,2} \\
  a_{7,1}
\end{bmatrix}
\]

**transpose, \( lda = n1 \)**

\[
\begin{bmatrix}
  a_{1,1} & a_{2,1} & a_{3,1} & a_{4,1} & a_{5,1} & a_{6,1} & a_{7,1} \\
  a_{2,2} & a_{3,2} & a_{4,2} & a_{5,2} & a_{6,2} & a_{7,2} \\
  a_{3,3} & a_{4,3} & a_{5,3} & a_{6,3} & a_{7,3} \\
  a_{4,4} & a_{5,4} & a_{6,4} & a_{7,4} \\
  a_{5,5} & a_{6,5} \\
  a_{6,6} \\
  a_{7,7}
\end{bmatrix}
\]

**Upper triangular**

\[
LDAR = N
\]

\[
\begin{bmatrix}
  a_{1,4} & a_{2,4} & a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{2,4} & a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{5,5} & a_{6,6} \\
  a_{6,6} \\
  a_{7,7}
\end{bmatrix}
\]

**transpose, \( lda = n1 \)**

\[
\begin{bmatrix}
  a_{1,4} & a_{2,4} & a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{2,4} & a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{5,5} & a_{6,6} \\
  a_{6,6} \\
  a_{7,7}
\end{bmatrix}
\]

6.2 RFPF for the matrices of rank even, here \( N = 6 \) and \( n1 = 3 \)

**Lower triangular**

\[
LDAR = N + 1
\]

\[
\begin{bmatrix}
  a_{1,1} & a_{2,2} & a_{3,3} & a_{4,4} & a_{5,5} & a_{6,6} \\
  a_{2,1} & a_{3,2} & a_{4,3} & a_{5,4} & a_{6,5} \\
  a_{3,1} & a_{4,2} & a_{5,3} & a_{6,4} \\
  a_{4,1} & a_{5,2} & a_{6,3} \\
  a_{5,1} & a_{6,2} \\
  a_{6,1}
\end{bmatrix}
\]

**transpose, \( lda = n1 \)**

\[
\begin{bmatrix}
  a_{1,1} & a_{2,1} & a_{3,1} & a_{4,4} & a_{5,5} & a_{6,6} \\
  a_{2,2} & a_{3,2} & a_{4,4} & a_{5,5} & a_{6,6} \\
  a_{3,3} & a_{4,4} & a_{5,5} & a_{6,6} \\
  a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{5,5} & a_{6,6} \\
  a_{6,6} \\
  a_{7,7}
\end{bmatrix}
\]

**Upper triangular**

\[
LDAR = N + 1
\]

\[
\begin{bmatrix}
  a_{1,4} & a_{2,4} & a_{3,4} & a_{4,4} \\
  a_{2,4} & a_{3,4} & a_{4,4} \\
  a_{3,4} & a_{4,4} \\
  a_{4,4} \\
  a_{5,5} \\
  a_{6,6}
\end{bmatrix}
\]

**transpose, \( lda = n1 \)**

\[
\begin{bmatrix}
  a_{1,4} & a_{2,4} & a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} \\
  a_{2,4} & a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} \\
  a_{3,4} & a_{4,4} & a_{5,5} & a_{6,6} \\
  a_{4,4} & a_{5,5} & a_{6,6} & a_{7,7} \\
  a_{5,5} & a_{6,6} \\
  a_{6,6} \\
  a_{7,7}
\end{bmatrix}
\]
error analysis and stability of these algorithms is very well described in the book of [Higham 1996]. The difference between LAPACK algorithms PO, PP and RFPF\(^3\) is how inner products are accumulated. In each case a different order is used. They are all mathematically equivalent, and, stability analysis shows that any summation order is stable.

8. A PERFORMANCE STUDY USING RFP FORMAT

There are 11 tables giving performance results of LAPACK and RFPF routines. The LAPACK Library routines POTRF, PPTRF, POTRI, TRTRI, LAUUM, PPTRI, POTRS and PPTRS are compared with the RFPF like PFTRF, PFTRI, TFTRI and PFTRS for Cholesky factorization, inverse and solution respectively. In all cases real long precision arithmetic (double precision) is used. Results were

\(^3\)full, packed and rectangular full packed.
<table>
<thead>
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Table 3. Performance in Mflop/s of Cholesky Solution on SUN UltraSPARC-III computer.

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Table 4. Performance in Mflop/s of Cholesky Factorization on ia64 Itanium computer.

obtained on several different computers using everywhere the vendor Level 3 and Level 2 BLAS.

Due to space limitations, we cannot present all of our timing results. We noticed a few anomalies in the performance runs for POTRS on SUN in Table 3, the PP runs on NEC (Tables 7, 8 and 9) and the PP and PO runs on SUN SMP parallel, Table 11. We have re-run these cases and have obtained the same results. At this time we do not have a rational explanation for these anomalies. Finally, our timings do not include the cost of sorting any LAPACK data formats to RFPF data formats and vice versa.

The tables from 1 to 9 show the performance comparison in Mflop/s of factorization, inversion and solution on SUN UltraSPARC-III (clock rate: 1200 MHz; L1 cache: 64 kB 4-way data, 32 kB 4-way instruction, and 2 kB Write, 2 kB Prefetch; L2 cache: 8 MB; TLB: 1040 entries), ia64 Itanium (CPU: Intel Itanium2: 1.3 GHz,
Table 5. Performance in Mflop/s of Cholesky Inversion on ia64 Itanium computer

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Table 6. Performance in Mflop/s of Cholesky Solution on ia64 Itanium computer

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cache: 3 MB on-chip L3 cache), and NEC SX-6 computer (8 CPU's, per CPU peak : 8 Gflops, per node peak : 64 Gflops, vector register length: 256). The tables 10 and 11 show the SMP parallelism of these subroutines on the IBM Power4 (clock rate: 1300 MHz; two CPUs per chip; L1 cache: 128 KB (64 KB per CPU) instruction, 64 KB 2-way (32 KB per CPU) data; L2 cache: 1.5 MB 8-way shared between the two CPUs; L3 cache: 32 MB 8-way shared (off-chip); TLB: 1024 entries) and SUN UltraSPARC-IV (the same hardware parameters as SUN UltraSPARC-III except the clock rate: 1350 MHz) computers respectively. They compare SMP times of PFTRF, POTRF and PPTRF. The tables 10 and 11 also show the times of the four operations (POTRF, TRSM, SYRK and again POTRF) inside the new algorithm PFTRF.
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Table 7. Performance in Mflop/s of Cholesky Factorization on SX-6 NEC computer with Vector Option

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Table 9. Performance in Mflop/s of Cholesky Solution on SX-6 NEC computer with Vector Option

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Table 10. Performance Times and Mflop/s of Cholesky Factorization on an IBM Power 4 computer using SMP parallelism on 1, 5, 10 and 15 processors. Here vendor codes for Level 2 and 3 BLAS and POTRF are used, ESSL library version 3.3. PPTRF is LAPACK code. UPLO = 'L'.

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Table 11. Performance in Times and MFlop/s of Cholesky Factorization on SUN UltraSPARC-IV computer with a different number of Processors, testing the SMP Parallelism. PPTRF does not show any SMP parallelism. UPLO = 'L'.

9. SUMMARY AND CONCLUSIONS
This paper describes RFPF as a standard minimal full format for representing both symmetric and triangular matrices. Hence, these matrix layouts are a replacement for both the standard formats of DLA, namely full and packed storage. These new layouts possess three good features: they are supported by Level 3 BLAS and LAPACK full format routines, and they require minimal storage.

10. ACKNOWLEDGMENTS
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