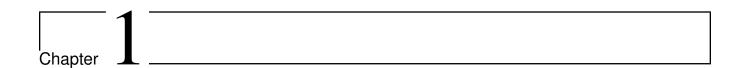
Assignments

Oscaration		Callina camana	Tom
Operation	Comment	Calling sequence	Team
Example			
B := LB	Lower triangular matrix L	Trmm_llnn(L, B)	Robert vdG
Symmetric mat	Symmetric matrix-matrix multiplication		
C := AB + C	A symmetric, stored in lower-triangular part	$Symm_1(A, B, C)$	
C := AB + C	A symmetric, stored in uppertriangular part	Symm_lu(A,B,C)	
C := BA + C	A symmetric, stored in lower-triangular part	Symm_rl(A,B,C)	
C := BA + C	A symmetric, stored in uppertriangular part	Symm_ru(A,B,C)	
Symmetric rank-k update	k-k update		
$C := AA^T + C$	C symmetric, stored in lower-triangular part	Syrk_ln(A, C)	Ayushi Sharma, Grace Simmons, Matthew Martinez
$C := AA^T + C$	C symmetric, stored in uppertriangular part	$Syrk_ut(A,C)$	Paras Arora, Ho-Jae Jung
$C := A^T A + C$	C symmetric, stored in lower-triangular part	Syrk_ln(A,C)	Laurence Zhang, Yvonne Huang, Cindy Truong
$C := A^T A + C$	C symmetric, stored in uppertriangular part	Syrk_ut(A,C)	Melissa Mendez, Irene Jiang

Operation	Comment	Calling sequence	Team
Symmetric rank-2k update	date		
$C := AB^T + BA^T + C$	C symmetric, stored in lower-triangular part	Syr2k_ln(A, B, C)	Ethan Lind, John Tunstall, Anika Gera
$C := AB^T + BA^T + C$	C symmetric, stored in uppertriangular part	Syr2k_un(A, B, C)	Maria Salazar, Arathi Kumar
$C := A^T B + B^T A + C$	C symmetric, stored in lower-triangular part	Syr2k_lt(A, B, C)	Abhilash Chilakamarthi
$C := A^T B + B^T A + C$	C symmetric, stored in uppertriangular part	Syr2k_ut(A, B, C)	Dennis Stasaski, Mridhul Maddela
Triangular matrix-matrix multiplication	rix multiplication		
$B := L^T B$	L stored in lower triangle	Trmm_lltn(A, B)	Aditya Palnitkar, Jamie Xia, Vignesh Kumar
B := UB	U stored in upper triangle	Trmm_lunn(A, B)	Jaydeep Singh, Jonas Wechsler
$B := U^T B$	U stored in upper triangle	Trmm_lutn(A, B)	Ahmed Al Nabil, Abhiram Ilindra
B := BL	L stored in lower triangle	Trmm_rlnn(A, B)	Ted Wong, Cameron Reynoldson
$B := BL^T$	L stored in lower triangle	Trmm_rltn(A, B)	Juan Manuel Penaranda Ortiz, Derek Tolliver
B := BU	U stored in upper triangle	Trmm_runn(A, B)	Shannon Yap, Vinay Valsaraj, Po-Chen Wu
$B:=BU^T$	U stored in upper triangle	Trmm_rutn(A, B)	Drew Poche, Vamsi Ponakala
Triangular-triangular matrix multiplication	natrix multiplication		
B := LU	L lower triangular, U upper triangular	Trtrmm_lunn(L, U, B)	Cristina Anderson, Adam Leslie
B := UL	L lower triangular, U upper triangular	Trtrmm_ulnn(U, L, B)	Jesus Romero, John Knoebel, Raul Camacho
$B := U^T L$	L lower triangular, U upper triangular	Trtrmm_ultn(U, L, B)	
$B := UL^T$	L lower triangular, U upper triangular	Trtrmm_ulnt(U, L, B)	3

Part I

Triangular Matrix-matrix Multiplication



Triangular matrix-matrix multiplication (TRMM) is matrix-matrix multiplication where one of the matrices is square and (lower or upper) triangular.

The full set of triangular matrix-matrix multiplication cases is denoted by TRMM____\, where the letters in the four boxes denote whether the triangular matrix is on the Left or Right, is Lower or Upper triangular, is Not transposed or Transposed, and has a Nonunit or Unit diagonal:

	LL□□	LU□□	RL□□	RU□□
$\square\square$ N \square	B = LB	B = UB	B = BL	B = BU
	$B = L^T B$	$B = U^T B$	$B = BL^T$	$B = BU^T$

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Chapter 2

Example: B := LB — Team: Robert van de Geijn

2.1 Operation

Consider the operation

$$B := LB$$

where L is a $m \times m$ lower triangular matrix and B is a $m \times n$ matrix. This is a special case of triangular matrix-matrix multiplication, with the Lower triangular matrix on the Left, and the triangular matrix is Not transposed. We will refer to this operation as TRMM_LLNN where the LLNN stands for left lower no-transpose nonunit diagonal. The Nonunit diagonal means we will use the entries of the matrix that are stored on the diagonal. In some other cases, the entries on the diagonal may implicitly be taked to all equal one (Unit).

2.2 Precondition and postcondition

In the precondition

$$B = \widehat{B}$$

 \widehat{B} denotes the original contents of B. This allows us to express the state upon completion, the postcondition, as

$$B=L\widehat{B}.$$

It is implicitly assumed that L is nonunit lower triangular.

2.3 Partitioned Matrix Expressions and loop invariants

There are two PMEs for this operation.

2.3.1 PME 1

To derive the second PME, partition

$$L
ightarrow \left(egin{array}{c|c} L_{TL} & 0 \ \hline L_{BL} & L_{BR} \end{array}
ight), \quad ext{and} \quad B
ightarrow \left(egin{array}{c|c} B_T \ \hline B_B \end{array}
ight).$$

Substituting these into the postcondition yields

$$\left(\begin{array}{c|c}
B_T \\
\hline
B_B
\end{array}\right) = \left(\begin{array}{c|c}
L_{TL} & 0 \\
\hline
L_{BL} & L_{BR}
\end{array}\right) \left(\begin{array}{c|c}
\widehat{B}_T \\
\hline
\widehat{B}_B
\end{array}\right)$$

or, equivalently,

$$\left(\frac{B_T}{B_B}\right) = \left(\frac{L_{TL}\widehat{B}_T}{L_{BL}\widehat{B}_T + L_{BR}\widehat{B}_B}\right)$$

so that, upon completion

$$\left(\frac{B_T}{B_B}\right) = \left(\frac{L_{TL}\widehat{B}_T}{L_{BL}\widehat{B}_T + L_{BR}\widehat{B}_B}\right)$$

From this, we can choose two loop invariants:

Invariant 1:
$$\left(\frac{B_T}{B_B}\right) = \left(\frac{\widehat{B}_T}{L_{BR}\widehat{B}_B}\right)$$
.

(The top part has been left alone and the bottom part has been partially computed).

Invariant 2:
$$\left(\frac{B_T}{B_B}\right) = \left(\frac{\widehat{B}_T}{L_{BL}\widehat{B}_T + L_{BR}\widehat{B}_B}\right)$$
.

(The top part has been left alone and the bottom part has been completely computed).

2.3.2 PME 2

To derive the second PME, partition

$$B
ightarrow \left(egin{array}{c|c} B_L & B_R \end{array}
ight)$$

and do not partition L. Substituting these into the postcondition yields

$$\left(\begin{array}{c|c} B_L & B_R \end{array}\right) = L \left(\begin{array}{c|c} \widehat{B}_L & \widehat{B}_R \end{array}\right)$$

or, equivalently,

$$\left(\begin{array}{c|c}B_L & B_R\end{array}\right) = \left(\begin{array}{c|c}L\widehat{B}_L & L\widehat{B}_R\end{array}\right)$$

so that, upon completion

$$\left(\begin{array}{c|c}B_L & B_R\end{array}\right) = \left(\begin{array}{c|c}L\widehat{B}_L & L\widehat{B}_R\end{array}\right)$$

From this, we can choose two more loop invariants:

Invariant 3:
$$(B_L \mid B_R) = (L\widehat{B}_L \mid \widehat{B}_R)$$
.

Invariant 3: $\begin{pmatrix} B_L & B_R \end{pmatrix} = \begin{pmatrix} L\widehat{B}_L & \widehat{B}_R \end{pmatrix}$. (The left part has been completely finished and the right part has been left untouched).

Invariant 4:
$$\begin{pmatrix} B_L & B_R \end{pmatrix} = \begin{pmatrix} \widehat{B}_L & L\widehat{B}_R \end{pmatrix}$$
. (The left part has been completely finished and the right part has been left untouched).

2.3.3 Notes

How do I decide to partition the matrices in the postcondition?

- Pick a matrix (operand), any matrix.
- If that matrix has
 - a triangular structure (in storage), then you want to either partition is into four quadrants, or not at all. Symmetric matrices and triangular matrices have a triangular structure (in storage).
 - no particular structure, then you partition it vertically (left-right), horizontally (top-bottom), or not at all.
- Next, partition the other matrices similarly, but conformally (meaning the resulting multiplications with the parts are legal).

Take our problem here: B := LB. Start by partitioning L in to quadrants:

$$B = \left(\begin{array}{c|c} L_{TL} & 0 \\ \hline L_{BL} & L_{BR} \end{array}\right) \widehat{B}.$$

Now, the way partitioned matrix multiplication works, this doesn't make sense:

$$B = \underbrace{\begin{pmatrix} L_{TL} & 0 \\ L_{BL} & L_{BR} \end{pmatrix}} \widehat{B}.$$

$$\underbrace{\begin{pmatrix} L_{TL} \times \text{something} + 0 \times \text{something} \\ L_{BL} \times \text{something} + L_{BR} \times \text{something} \end{pmatrix}}$$

So, we need to also partition *B* into a top part and a bottom part:

$$\left(\begin{array}{c|c}
B_T \\
B_B
\end{array}\right) = \underbrace{\left(\begin{array}{c|c}
L_{TL} & 0 \\
L_{BL} & L_{BR}
\end{array}\right) \left(\begin{array}{c}
\widehat{B}_T \\
\widehat{B}_B
\end{array}\right)}_{\left(\begin{array}{c|c}
L_{TL}B_T + B_B \\
L_{BL}B_T + L_{BR}B_B
\end{array}\right)}_{\left(\begin{array}{c}
L_{BL}B_T + L_{BR}B_B
\end{array}\right)}.$$

Alternatively, what if you don't partition L? You have to partition something so let's try partitioning B:

$$\left(\frac{B_T}{B_B}\right) = L\left(\frac{\widehat{B}_T}{\widehat{B}_B}\right)$$

But that doesn't work... Instead

$$\left(\begin{array}{c|c}B_L & B_R\end{array}\right) = L\left(\begin{array}{c|c}\widehat{B}_L & \widehat{B}_R\end{array}\right) = \left(\begin{array}{c|c}L\widehat{B}_L & L\widehat{B}_R\end{array}\right)$$

works just fine.

2.4 Deriving all unblocked algorithms

The below table

Part II Symmetric Matrix-Matrix Multiplication



Symmetric matrix-matrix multiplication (SYMM) is matrix-matrix multiplication where one of the matrices is square and symmetric, stored in the lower or upper triangular part of the matrix.

The full set of symmetric matrix-matrix multiplication cases is denoted by SYMM__\, where the letters in the two boxes denote whether the symmetric matrix is on the Left or Right, and is stored in the Lower or Upper triangular part of that matrix:

	L□	R□	
\Box L	C := AB + C	C := BA + C	A symmetric, stored in lower triangle
□U	C := AB + C	C := BA + C	A symmetric, stored in upper triangle

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Part III Symmetric Rank-k Update



Symmetric rank-k update (SYRK) is matrix-matrix multiplication of a matrix with its transpose, where the matrix *C* being updated is symmetric, stored in the lower or upper triangular part of the matrix.

The full set of symmetric rank-k cases is denoted by SYRK___\, where the letters in the two boxes denote whether the matrices begin multiplied are Not transposed or Tranksposed, and whether the matrix being updated is stored in the Lower or Upper triangular part:

	N□	т□	
	$C := AA^T + C$	$C := A^T A + C$	C symmetric, stored in lower triangle
□U	$C := AA^T + C$	$C := A^T A + C$	C symmetric, stored in upper triangle

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Part IV Symmetric Rank-2k Update



Symmetric rank-2k update (SYR2K) is matrix-matrix multiplication of two matrices where both the result and the transpose of that result are added to a symmetric matrix C, stored in the lower or upper triangular part of the matrix.

The full set of symmetric rank-2k cases is denoted by SYR2K_\[
\sum_\], where the letters in the two boxes denote whether the matrices begin multiplied are Not transposed or Transposed, and whether the matrix being updated is stored in the Lower or Upper triangular part:

	N□	т□	
	$C := AB^T + BA^T + C$	$C := A^T B + B^T A + C$	C symmetric, stored in lower triangle
□U	$C := AB^TBA^T + C$	$C := A^T B + B^T A + C$	C symmetric, stored in upper triangle

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Part V

Triangular Solve with Multiple Right-hand Sides



Triangular solve with multiple right-hand sides (TRSM) is a matrix-matrix multiplication in disguise: Not only is one of the matrices square and (lower or upper) triangular, but in addition, one multiplies with the inverse of the matrix.

The full set of TRSM cases is denoted by TRSM. \(\subseteq \subseteq \subseteq\), where the letters in the four boxes denote whether the triangular matrix is on the Left or Right, is Lower or Upper triangular, is Not transposed or Transposed, and has a Nonunit or Unit diagonal:

		LU□□	$RL\Box\Box$	RU□□
$\square\square$ N \square	$B := L^{-1}B$	B := U - 1B	$B := BL^{-1}$	$B := BU^{-1}$
	$B := L^{-T}B$	$B :: U^{-T}B$	$B := BL^{-T}$	$B := BU^{-T}$

We will only consider the case where we compute with the diagonal.

Where does the name triangular solve with multiple right-hand sides come from? If one wants to compute $B := A^{-1}B$ one can also think of this as solving AX = B, overwriting the matrix A with the solution matrix X. Now, partition X and B by columns. Then

$$\underbrace{A\left(\begin{array}{c|c}x_0 & x_1 & \cdots & x_{n-1}\end{array}\right)}_{\left(\begin{array}{c|c}Ax_0 & Ax_1 & \cdots & Ax_{n-1}\end{array}\right)} = \left(\begin{array}{c|c}b_0 & b_1 & \cdots & b_{n-1}\end{array}\right)$$

and hence, for each column j, $Ax_j = b_j$. This means that one can instead solve with matrix A, and because B has many columns, this becomes a *solve with multiple right-hand sides*. If A is triangular, then it is a *triangular solve with multiple right-hand sides*. Importantly, the inverse of the matrix is never computed.