Combining Explicit and Recursive Blocking for Solving Triangular Sylvester-Type Matrix Equations on Distributed Memory Platforms



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Outline

- Triangular Sylvester-type matrix equations
- ScaLAPACK-style algorithms
- RECSY recursive blocked algorithms and HPC library
- Experimental results
- Conclusions and summary
- Future work
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Triangular Sylvester-type matrix equations (1)

Arise in many different applications in science and engineering

- Condition estimation of eigenvalue problems
- Control theory

Name	Matrix Equation
Standard Sylvester (CT)	AX - XB = C
Standard Lyapunov (CT)	$AX + XA^T = C$
Standard Sylvester (DT)	$AXB^T - X = C$
Standard Lyapunov (DT)	$AXA^T - X = C$
Generalized Coupled Sylvester	(AX - YB, DX - YE) = (C, F)
Generalized Sylvester	$AXB^T - CXD^T = E$
Generalized Lyapunov (CT)	$AXE^T + EXA^T = C$
Generalized Lyapunov (DT)	$AXA^T - EXE^T = C$

Consider the continuous-time Sylvester equation (SYCT):

$$AX - XB = C, \quad A \in \mathbb{R}^{m \times m}, B \in \mathbb{R}^{n \times n}, C \in \mathbb{R}^{m \times n}$$

A and B in real Schur form

Solution is unique iff $\lambda(A) \cap \lambda(B) = \emptyset$

Triangular Sylvester-type matrix equations (2)

SYCT equivalent to linear system of equations:

$$\begin{bmatrix} Z_{SYCT} \operatorname{vec}(X) = \operatorname{vec}(C), \\ Z_{SYCT} = I_n \otimes A - B^T \otimes I_m \end{bmatrix}$$

 Explicit Kronecker product representation used in kernel solvers, e.g., in blocked algorithms:

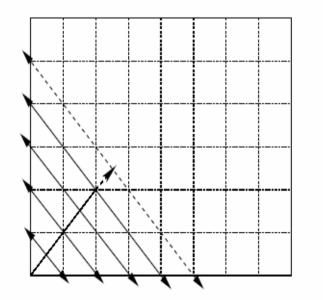
$$A_{ii}X_{ij} - X_{ij}B_{jj} = C_{ij} - \left(\sum_{k=i+1}^{D_a} A_{ik}X_{kj} - \sum_{k=1}^{j-1} X_{ik}B_{kj}\right)$$

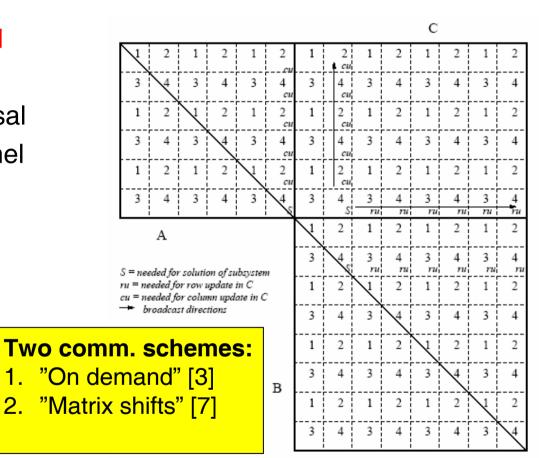
for $j=1, D_b$ for $i=D_a, 1, -1$ {Solve the (i, j)th subsystem using a kernel solver} $A_{ii}X_{ij} - X_{ij}B_{jj} = C_{ij}$ for k=1, i-1{Update block column j of C} $C_{kj} = C_{kj} - A_{ki}X_{ij}$ end for $k=j+1, D_b$ {Update block row i of C} $C_{ik} = C_{ik} + X_{ij}B_{jk}$ end end end

With appropriate blocking, cost is dominated by GEMM-updates of right hand side

ScaLAPACK-style algorithms (1)

- Rectangular process grid
- 2D block cyclic mapping
- Wave-front matrix traversal
- LAPACK's DTRSYL kernel solver (ess. Level-2)





ScaLAPACK-style algorithms (2)

- Communication operations ("On demand"):
 - 1. Implicit redistribution [3] (splitted 2x2 diagonal blocks)
 - Minor impact on total execution time
 - 2. Solving subsystem (i,j)
 - Minor impact on total execution time
 - 3. Broadcast of subsolution (*i*,*j*) in block row *i* and block column *j*
 - Bottleneck causes idle processes waiting for BC to start
 - 4. GEMM-updates of right hand side
 - Bottleneck causes idle processes
- "Matrix shifts" approach combines (2) and (4) into one single (expensive) operation but suffers from broadcast bottleneck
- "OD" allows single transposes in equation, whereas "MS" does not
- This contribution attempts to minimize (3) by means of a faster kernel node solver thus improving the total execution time – and creating ScaLAPACK-style hybrid algorithms

RECSY – recursive blocked algorithms (1)

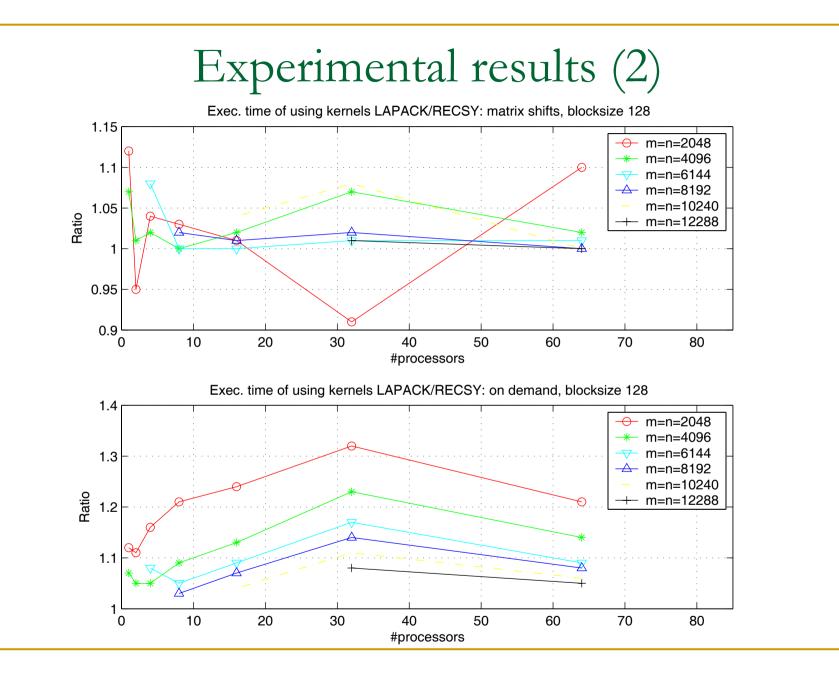
- HPC library which solves 42 sign- and transpose variants of 8 Sylvester-type matrix equations
 - Recursive blocking approach
 - Automatic variable-sized blocking matches memory hierarchies of today's modern computers
 - Significant speedup compared to standard library routines
 - For example SYCT solver RECSYCT: 10-folded speedup compared to LAPACK's DTRSYL
- Based on work by Jonsson-Kågström [4,5,6]
- Library was presented at EuroPar 2003.
- Documentation, download and installation instructions at <u>http://www.cs.umu.se/research/parallel/recsy</u>

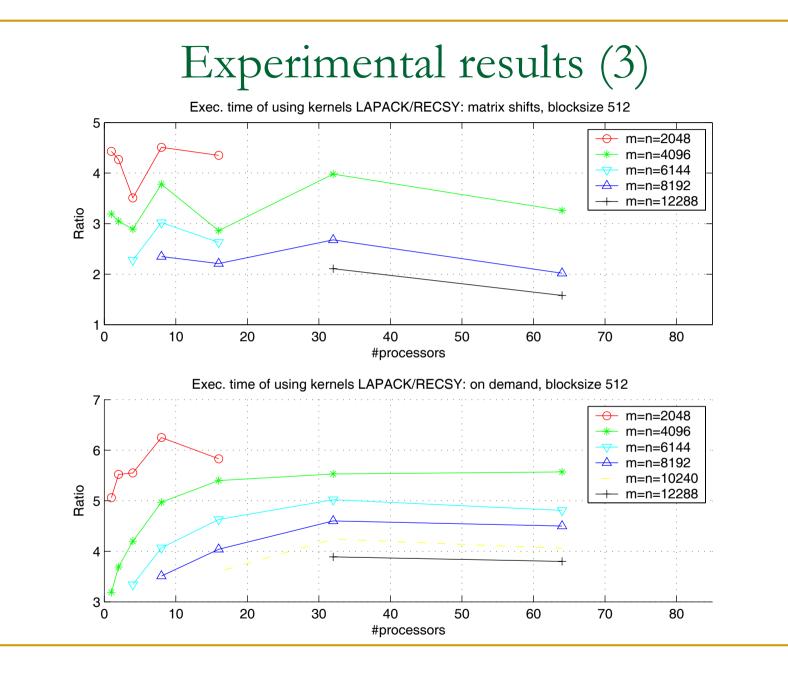
RECSY – recursive blocked algorithms (2)

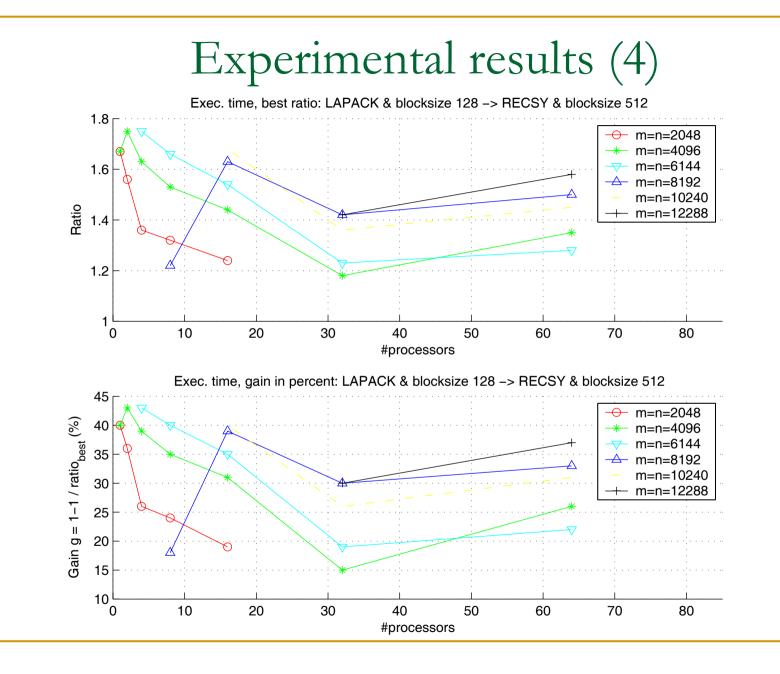
- Recursive blocking makes RECSY rich in GEMM-operations => high performance (if good GEMM impl. is available)
- Keep subsystems "squarish" by testing dimensions in each step of the recursion. At leaf node apply superscalar kernel.
- For example n/2 < m < 2n:

Experimental results (1)

- Comparing ScaLAPACK-style SYCT-solving algorithm with LAPACK or RECSY node solvers
- Target computer system: HPC2N Linux Super Cluster
 - 120 dual 1.667MHz nodes with 1GB RAM
 - 667 Mbyte/sec peak network bandwidth
 - Quite unbalanced
 - Time(flop) / (Memory bandwith)^(-1) ratio: 0.31
 - Time(flop) / (Network bandwidth)^(-1) ratio: 0.10
 - Efficient use of memory hierarchy and network necessary for good performance
- We present ratios of the parallel execution time with the node solvers from LAPACK and RECSY, respectively, for both communication schemes going from normal blocksize (128) to large (512)
 - Ratio > 1.0 represents speedup when going from LAPACK to RECSY







Conclusions and summary

- RECSY solver improves overall parallel performance by 15-43% for the results presented (Notice the misprint in contribution abstract!)
 - Allows larger blocks in the parallel algorithm
 - Close(r) to GEMM-performance for subsystem solves
 - Decreases synchronization cost for both "On demand" and "Matrix shifts"
 - But affects scalability (harder to hide comm. during comp.)
- Improves comparision with iterative methods [2] on unbalanced systems
- With RECSY both communications schemes perform equally good
- With LAPACK:
 - "On demand" best for small blocks
 - "Matrix shifts" best for larger blocks
- RECSY's LAPACK/SLICOT wrappers allow the user to choose kernel node solver without modifying source code

Future Work

- New implementations for all transpose and sign-variants of the non-generalized standard matrix equations
- User may choose communication scheme
- Future software package *SCASY* will also contain generalized solvers and parallel condition estimators
- By linking with RECSY, hybrid algorithms come for free

$op(A)X \pm Xop(B) = C$	SYCT	\checkmark
$op(A)X + Xop(A^T) = C$	LYCT	\checkmark
$op(A)Xop(B) \pm X = C$	SYDT	\checkmark
$op(A)Xop(A^T) - X = C$	LYDT	\checkmark
$\int op(A)X \pm Yop(B) = C,$	GCSY	
$\int op(D)X \pm Yop(E) = F$		
$op(A)Xop(B) \pm op(D)Xop(E) = C$	GSYL	
$op(A)Xop(A^{T}) - op(E)Xop(E^{T}) = C$	GLYCT	
$op(A)X(E^{T}) + op(E)Xop(A^{T}) = C$	GLYDT	

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