

Sparse Direct Solver on Today's Architectures

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Content



- Algorithm flowchart
- Performance on representative machines
 - IBM power5 (bassi @ NERSC)
 - Multicores: Intel Clovertown, Sun Niagra2 (Clusters @ UC Berkeley)

- Acknowledgement
 - Rich Vuduc, Georgia Tech
 - Sam Williams, UC Berkeley

Gaussian Elimination (GE)



- Solving a system of linear equations Ax = b
- First step: (make sure α not too small . . . may need pivoting)

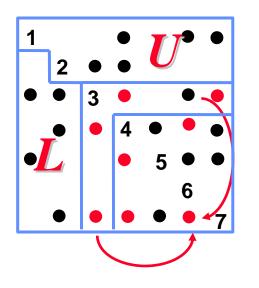
$$A = \begin{bmatrix} \alpha & w^{T} \\ v & B \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ v/\alpha & I \end{bmatrix} \cdot \begin{bmatrix} \alpha & w^{T} \\ 0 & C \end{bmatrix}$$
$$C = B - \frac{v \cdot w^{T}}{\alpha}$$

- Repeats GE on C
- Results in {L\U} decomposition (A = LU)
 - L lower triangular with unit diagonal, U upper triangular
- Then, x is obtained by solving two triangular systems with L and U



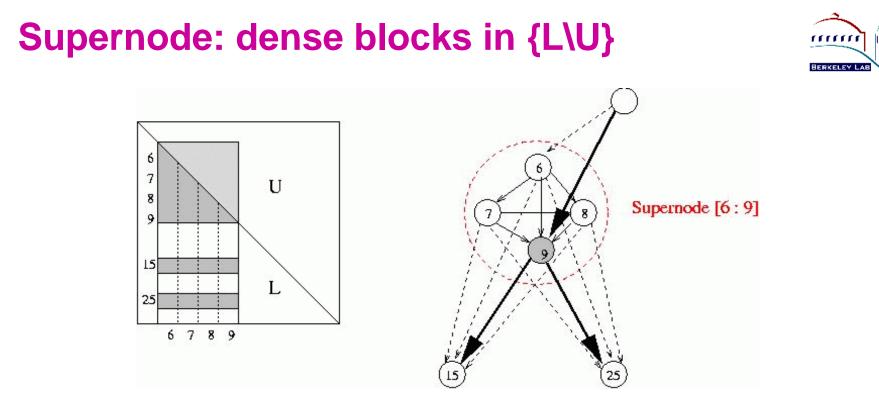


Scalar version : 3 nested loop



for i = 1 to n-1 for j = i+1 to n A(j,i) = A(j,i) / A(i,i)for k = i+1 to n s.t. A(i,k) != 0for j = i+1 to n s.t. A(j,i) != 0A(j,k) = A(j,k) - A(j,i) * A(i,k)

Typical fill ratio: 10x for 2D problems, 30-50x for 3D problems



- Good for high performance
 - Enable use of Level 3 BLAS
 - Reduce inefficient indirect addressing (scatter/gather)
 - Reduce time of the graph algorithms by traversing a coarser graph

SuperLU usage



- Over 6000 downloads each year (FY2005, FY2006)
- Research
 - In other DOE ACTS Tools: Hypre, PETSc, Overture, Trilinos
 - M3D-C1, NIMROD (fusion SciDAC)
 - Omega3P (accelerator SciDAC)
 - -...
- Industrial
 - Cray Scientific Libraries
 - FEMLAB
 - HP Mathematical Library
 - IMSL Numerical Library
 - NAG
 - Sun Performance Library
 - Python (NumPy, SciPy extensions)

SuperLU_DIST major steps

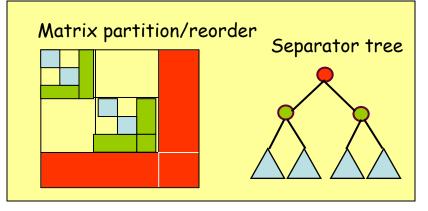
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- (parallelization perspectives)
- Static numerical pivoting: improve diagonal dominance
 - Currently use MC64 (HSL, serial)
 - Being parallelized [Riedy]
- Ordering to preserve sparsity
 - Can use parallel graph partitioning (e.g., ParMetis, Scotch)
- Symbolic factorization: determine pattern of {L\U}
 - Parallelized recently [Grigori]
- Numerics: factorization, triangular solutions, iterative refinement (usually dominate total time)
 - Parallelized a while ago; need to improve load balance, latency-hiding, . . .

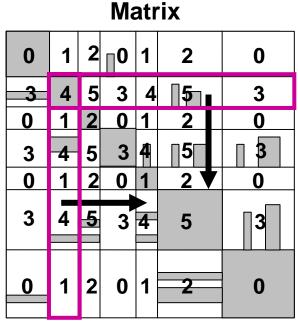
Data distribution

Ordering & Symbfact





Numerical phases



Processors

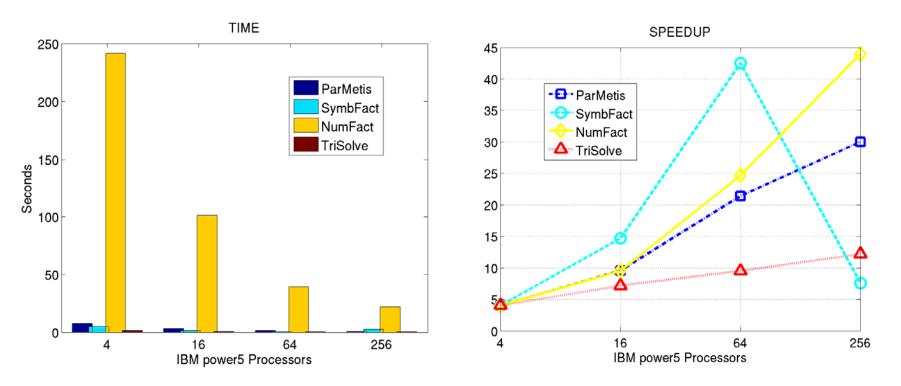
0	1	2
3	4	5

Time & speedup on IBM power5

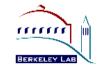


Matrix181 (M3D-C1, Fusion), N=589,698, fill-ratio=9.3

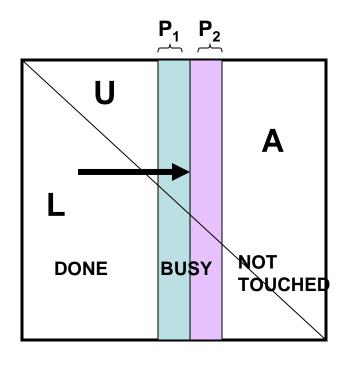
- NumFact dominates total time, scales better
- How to make TriSolve scale better? ... O(1) ops/message

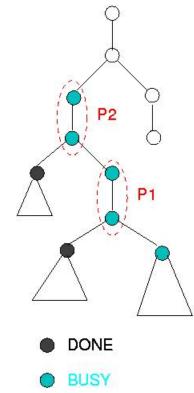


Multicore: SuperLU_MT [Li/Demmel/Gilbert]



- Pthread or OpenMP
- Left looking : many more reads than writes
- Use shared task queue to schedule ready columns in the elimination tree (bottom up)



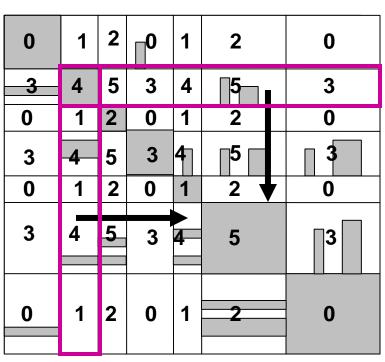


Multicore: SuperLU_DIST [Li/Demmel]

- MPI
- Right looking : many more writes than reads

Matrix

• 2D block cyclic layout



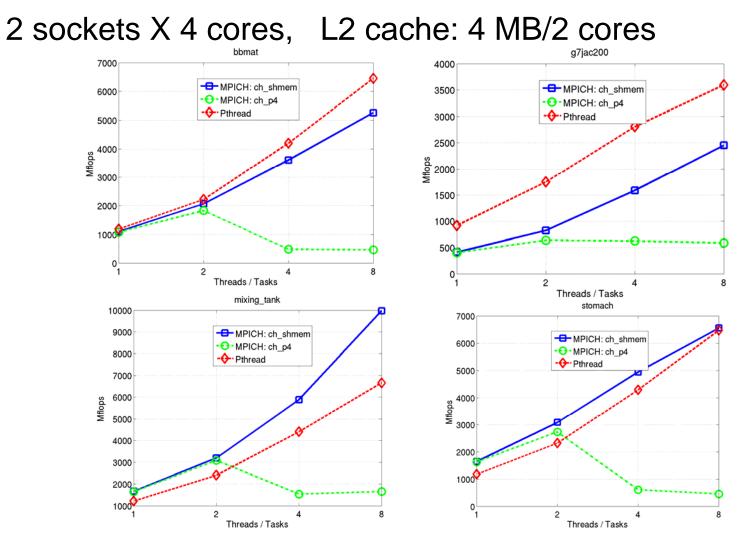


0	1	2
3	4	5



Intel Clovertown: 2.33 GHz Xeon



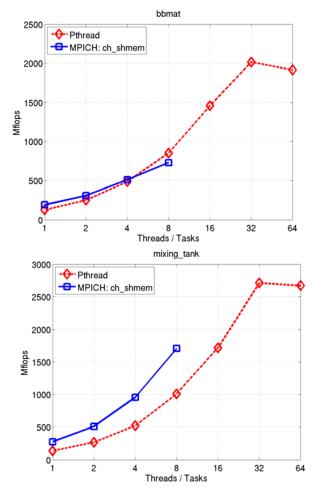


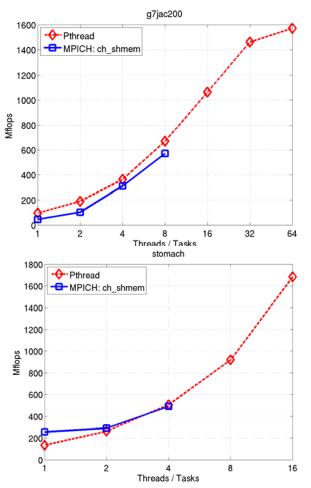
- Important to configure MPICH in shared memory mode
- Achieves only 13% peak

Sun Niagara2: 1.4 GHz UltraSparc T2



8 cores, 8 hw-threads/core, L2 cache shared: 4 MB





• Achieves 24% peak

Observations



- Explicit thread programming is beneficial
- MPI better be configured in both modes
- MPI + Pthread requires significant code rewriting
- How to do auto-tuning?