

Batched Generation of Block-Jacobi Preconditioners for Iterative Sparse Linear System Solvers on GPUs

Hartwig Anzt, Jack Dongarra, <u>Goran Flegar</u>, Enrique S. Quintana-Ortí, Andrés E. Tomás

 Solve sparse linear system using an iterative Krylov method

$$Ax = b, \ A \in \mathbb{R}^{n \times n}$$

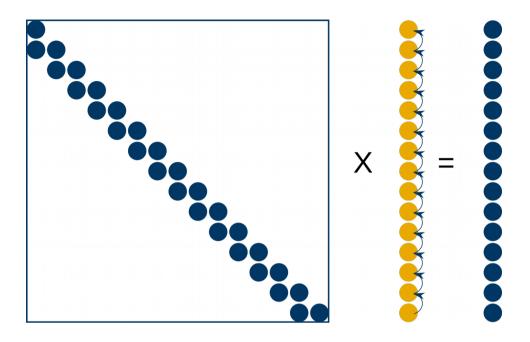
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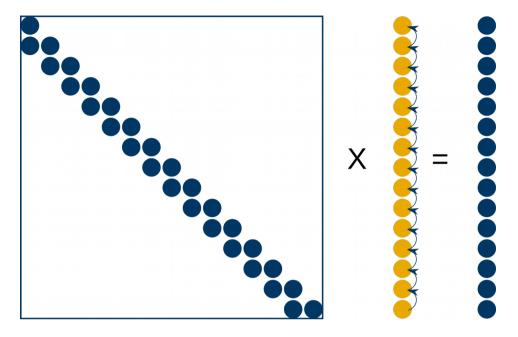
 Convergence typically benefits from using a preconditioner

$$M^{-1}Ax = M^{-1}b$$

- Solve sparse linear system using an iterative Krylov method
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- Solve sparse linear system using an iterative Krylov method
- Convergence typically benefits from using a preconditioner
- Conventional ILU-type preconditioners offer limited concurrency
- Need high degree of parallelism to use a GPU effectively
 - ILU on GPUs is an area of active research





source: devblogs.nvidia.com/parallelforall



GPU architecture (NVIDIA GP100)

- 56 SMs x 64 cores = 3584 cores!
- Need to oversubscribe to hide memory latency!





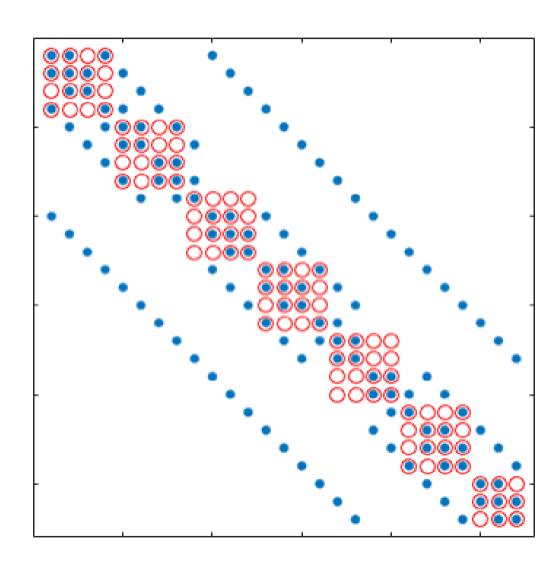
Different approach

Scalar Jacobi

 Scale with inverses of diagonal elements

Block-Jacobi

- Scale with inverses of diagonal blocks (possibly of different sizes!)
- Can reflect the block structure of the problem
- Often superior to scalar Jacobi
- Lower convergence rate improvement compared to ILU
- Can proces each block independently!



General Ideas

Restrict block size to 32x32

 Large block sizes require more memory to store the preconditioner matrix



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- Use a single warp to process the whole block (one thread per column)
 - No need for explicit synchronization



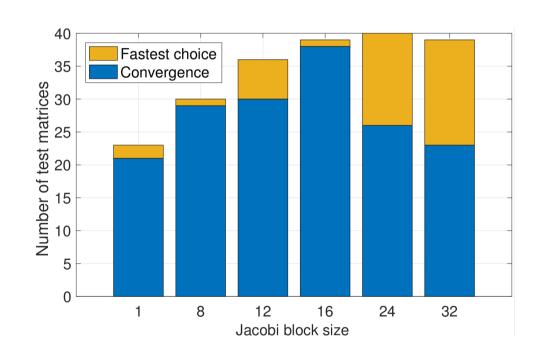
General Ideas

- Restrict block size to 32x32
 - Large block sizes require more memory to store the preconditioner matrix
- Use a single warp to process the whole block (one thread per column)
 - No need for explicit synchronization
- Use the large register file to store the entire block
 - Read/write from mem. once
 - Avoids load/store instructions
 - Com. via warp shuffles



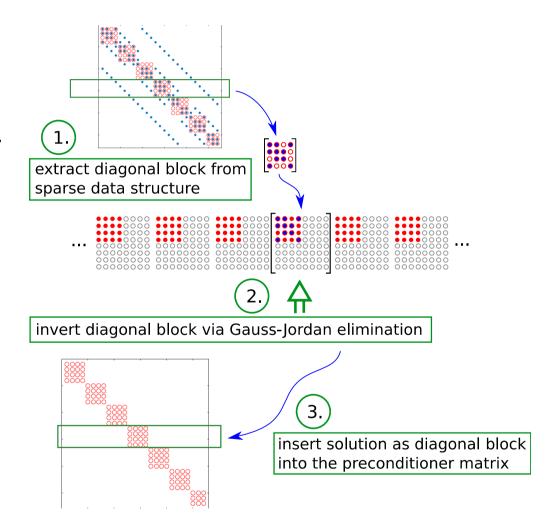
Benefits of block-Jacobi

- 40 matrices from SuiteSparse
- MAGMA-sparse open source library
 - IDR solver
 - Jacobi preconditioner
 - Supervariable blocking
- Block-Jacobi improves the robustness of the solver
 - More problems converge
- Decreases time-to-solution



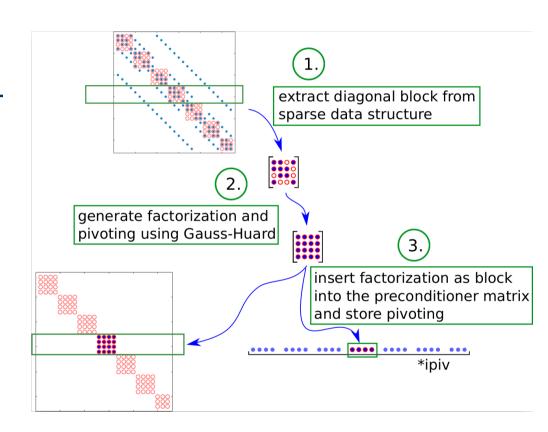
Implementation options

- Inversion in preconditioner setup + matrix-vector product in application
 - (FLOPS: 2n³ setup, 2n² app.)
 - H. Anzt et all., "Batched Gauss-Jordan Elimination for Block-Jacobi Preconditioner Generation on GPUs", PMAM'17

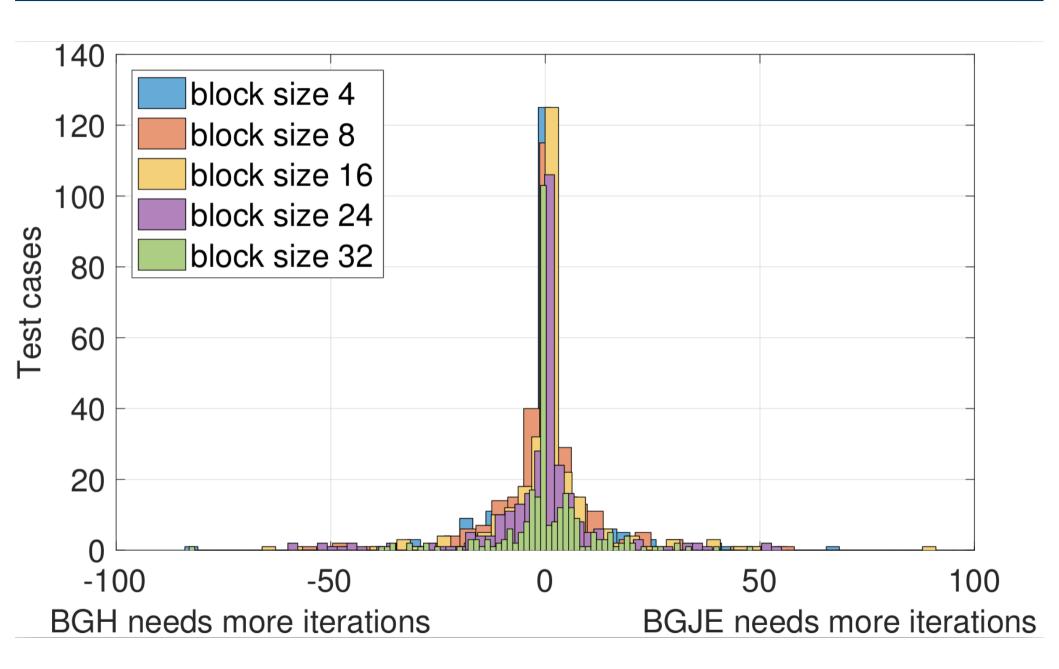


Different options

- Inversion in preconditioner setup + matrix-vector product in application
 - (FLOPS: 2n³ setup, 2n² app.)
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- Matrix decomposition in setup + solve in application
 - (FLOPS: $2/3n^3$ setup, $2n^2$ app.)
 - Gauss-Huard decomposition



Inversion?!



Gauss-Huard decomposition

Decomposition

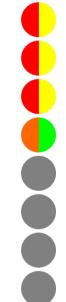
- GEMV (G = G RR)
- SCAL (O = O / B)
- GER (Y = Y BO)
- Column pivoting
 - Do not swap the columns, just remember which thread holds which column of the result

Solve

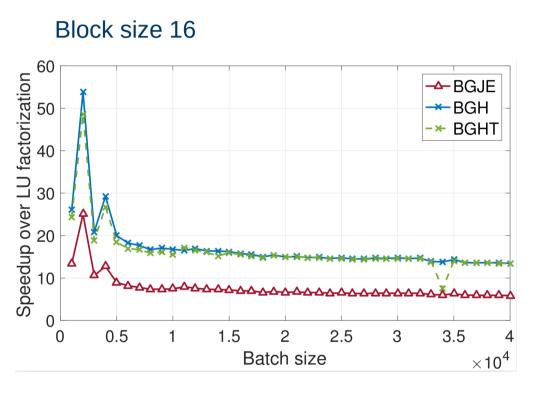
- Store only the solution vector into registers
- DOT (G = G RR)
- SCAL (0 = 0 / B)
- AXPY (Y = Y BO)
- Store lower part transposed wrp. to anti-diagonal for coalesced mem. access



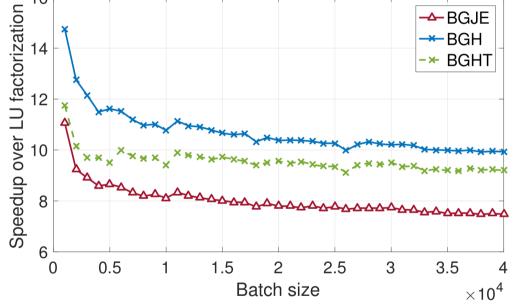




Comparison to batched LU (MAGMA)

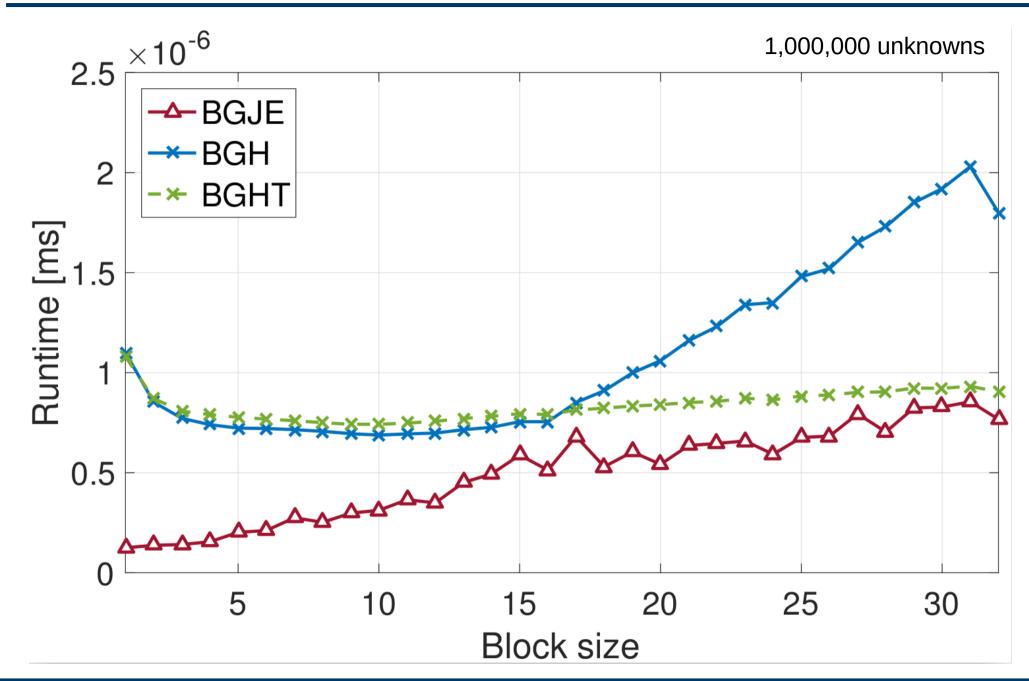






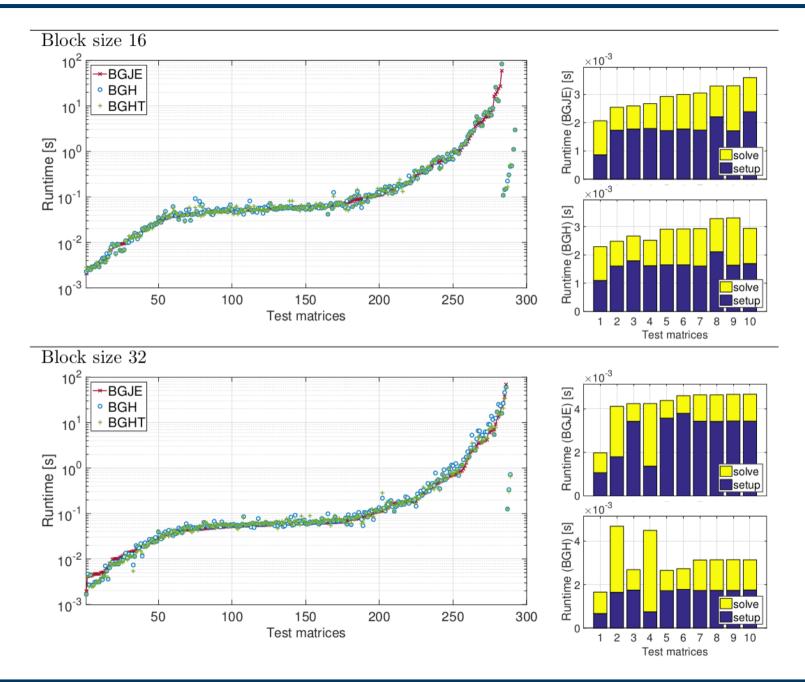
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Application time





Total runtime of block-Jacobi preconditioned BiCGSTAB



Thank you! Questions?

All functionalities are part of the MAGMA-sparse project.

MAGMA SPARSE

ROUTINES BiCG, BiCGSTAB, Block-Asynchronous Jacobi, CG,

CGS, GMRES, IDR, Iterative refinement, LOBPCG,

LSQR, QMR, TFQMR

PRECONDITIONERS ILU / IC, Jacobi, ParlLU, ParlLUT, Block Jacobi, ISAI

KERNELS SpMV, SpMM

DATA FORMATS CSR, ELL, SELL-P, CSR5, HYB

This research is based on a cooperation between Hartwig Anzt, Jack Dongarra (University of Tennessee), Goran Flegar, Enrique S. Quintana-Ortí and Adrés E. Tomás (Universidad Jaume I)





