



Accelerating GMRES via Mixed Precision

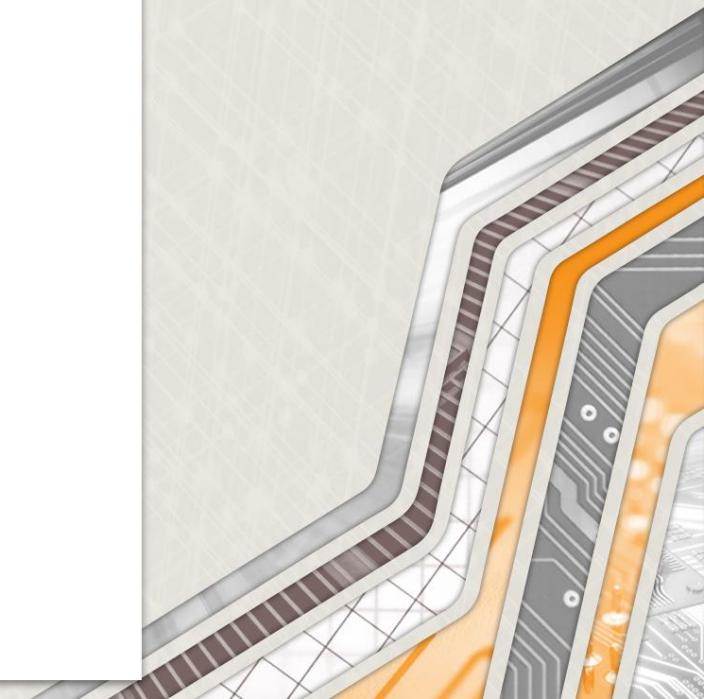


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GMRES

- General purpose, sparse linear solver
 - Iterative, Krylov solver
- Memory bound performance
 - Mix single and double precision

GMRES Algorithm

$\text{GMRES}_{res}(\mathbf{A}, \mathbf{x}_0, \mathbf{b}, \mathbf{M}^{-1})$

for $k = 0, 1, 2, \dots$

$$\mathbf{r}_k \leftarrow \mathbf{b} - \mathbf{A}\mathbf{x}_k$$

$$\mathbf{z}_k \leftarrow \mathbf{M}^{-1}\mathbf{r}_k$$

$$\beta \leftarrow \|\mathbf{z}_k\|_2$$

$$\mathbf{V}_{:,0} \leftarrow \mathbf{z}_k / \beta$$

$$\mathbf{s} \leftarrow [\beta, 0, 0, \dots, 0]^T$$

for $j = 0, 1, 2, \dots$

$$\mathbf{w} \leftarrow \mathbf{M}^{-1}\mathbf{A}\mathbf{V}_{:,j}$$

$$\mathbf{w}, \mathbf{H}_{:,j} \leftarrow \text{orthogonalize}(\mathbf{w}, \mathbf{V}_{:,j})$$

$$\mathbf{H}_{j+1,j} \leftarrow \|\mathbf{w}\|_2$$

$$\mathbf{V}_{:,j+1} \leftarrow \mathbf{w} / \|\mathbf{w}\|_2$$

$$\mathbf{H}_{:,j} \leftarrow \mathbf{G}_0 \mathbf{G}_1 \dots \mathbf{G}_{j-1} \mathbf{H}_{:,j}$$

$$\mathbf{G}_j \leftarrow \text{rotation_matrix}(\mathbf{H}_{:,j})$$

$$\mathbf{H}_{:,j} \leftarrow \mathbf{G}_j \mathbf{H}_{:,j}$$

$$\mathbf{s} \leftarrow \mathbf{G}_j \mathbf{s}$$

$$\mathbf{u}_k \leftarrow \mathbf{V}\mathbf{H}^{-1}\mathbf{s}$$

$$\mathbf{x}_{k+1} \leftarrow \mathbf{x}_k + \mathbf{u}_k$$

Computing $\mathbf{A}\mathbf{x} = \mathbf{b}$. $\mathbf{A}^{-1} \approx \mathbf{M}^{-1}$

Restarts

Iteration count

GMRES Algorithm

GMRES_{res}(A, x_0, b, M^{-1})

for $k = 0, 1, 2, \dots$

$$r_k \leftarrow b - Ax_k$$

$$z_k \leftarrow M^{-1}r_k$$

$$\beta \leftarrow \|z_k\|_2$$

$$V_{:,0} \leftarrow z_k / \beta$$

$$s \leftarrow [\beta, 0, 0, \dots, 0]^T$$

for $j = 0, 1, 2, \dots$

$$w \leftarrow M^{-1}AV_{:,j}$$

$$w, H_{:,j} \leftarrow \text{orthogonalize}(w, V_{:,j})$$

$$H_{j+1,j} \leftarrow \|w\|_2$$

$$V_{:,j+1} \leftarrow w / \|w\|_2$$

$$H_{:,j} \leftarrow G_0 G_1 \dots G_{j-1} H_{:,j}$$

$$G_j \leftarrow \text{rotation_matrix}(H_{:,j})$$

$$H_{:,j} \leftarrow G_j H_{:,j}$$

$$s \leftarrow G_j s$$

$$u_k \leftarrow VH^{-1}s$$

$$x_{k+1} \leftarrow x_k + u_k$$

Computing $Ax = b$. $A^{-1} \approx M^{-1}$

Restarts

Double:

Single:

Double:

Iteration count

GMRES Simplified Algorithm

$\text{GMRES}_{res}(A, x_0, b, M^{-1})$

for $k = 0, 1, 2, \dots$

Double: $r_k \leftarrow b - Ax_k$

Single: $u_k \leftarrow \text{GMRES}_{no\ res}(A, \vec{0}, r_k, M^{-1})$

Double: $x_{k+1} \leftarrow x_k + u_k$

GMRES Simplified Algorithm

$\text{GMRES}_{res}(A, x_0, b, M^{-1})$

for $k = 0, 1, 2, \dots$

Double: $r_k \leftarrow b - Ax_k$

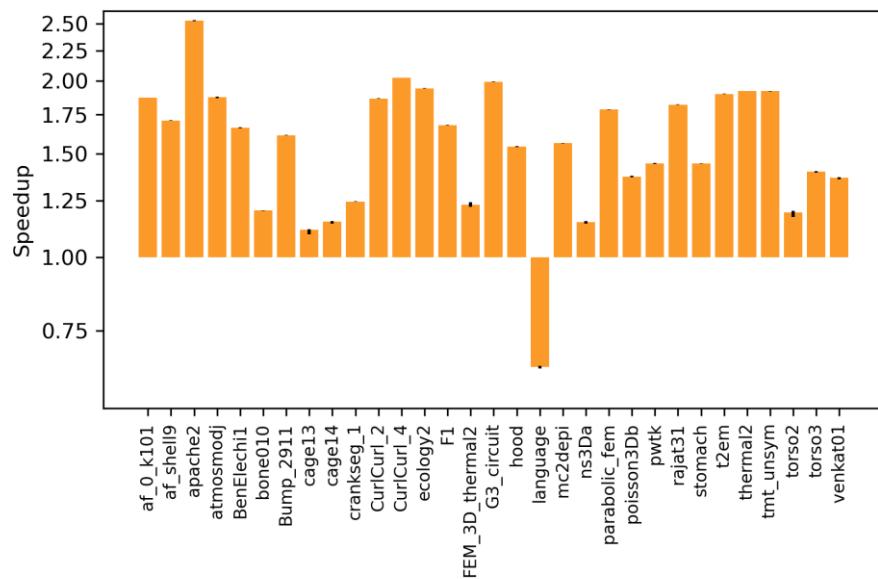
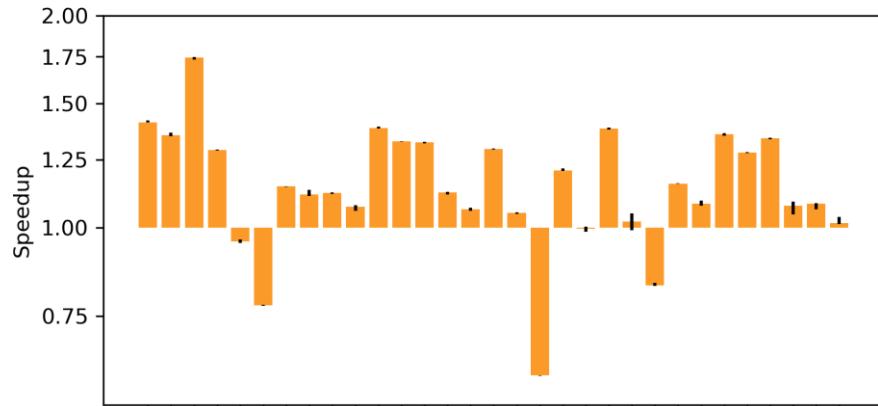
Single: $u_k \leftarrow A^{-1} r_k$

Double: $x_{k+1} \leftarrow x_k + u_k$

Performance

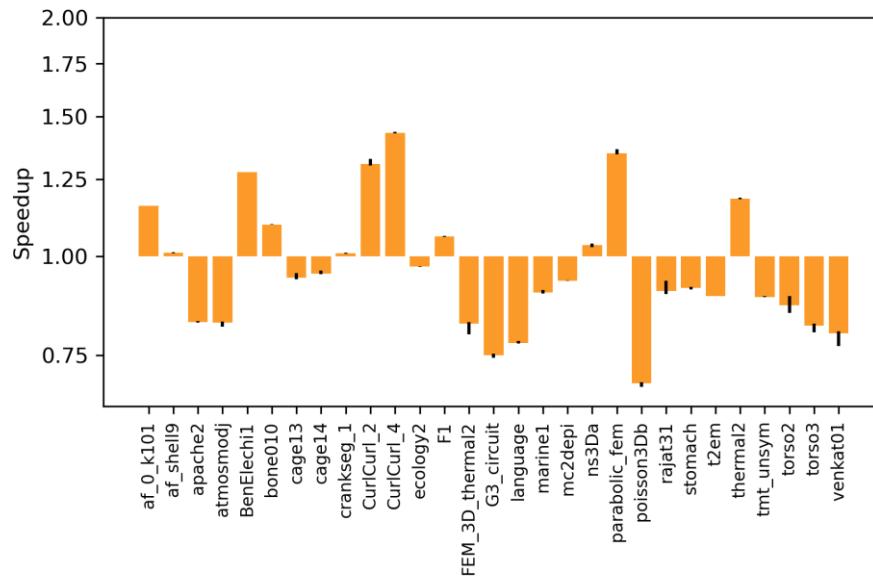
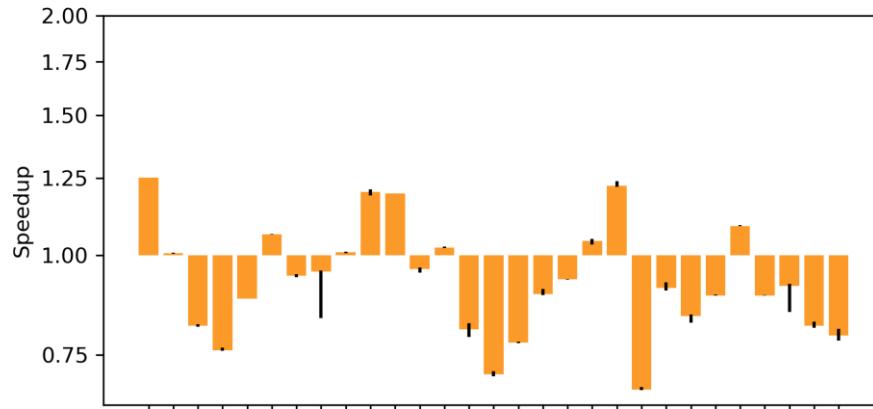
- Target accuracy $10^{-10} = \frac{\|b - Ax\|_2}{\|A\|_F \|x\|_2 + \|b\|_2}$
- Restart strategies:
 - I. 100 inner iterations
 - II. 100 inner iterations or residual estimate of 10^{-10}
 - III. First: 100 inner iterations or residual estimate of 10^{-6}
Then: same number of inner iterations
- 20-core Haswell node with NVIDIA V100 GPU
 - cuSparse, cuBLAS, Kokkos
- CSR matrix format

Performance – Scalar Jacobi



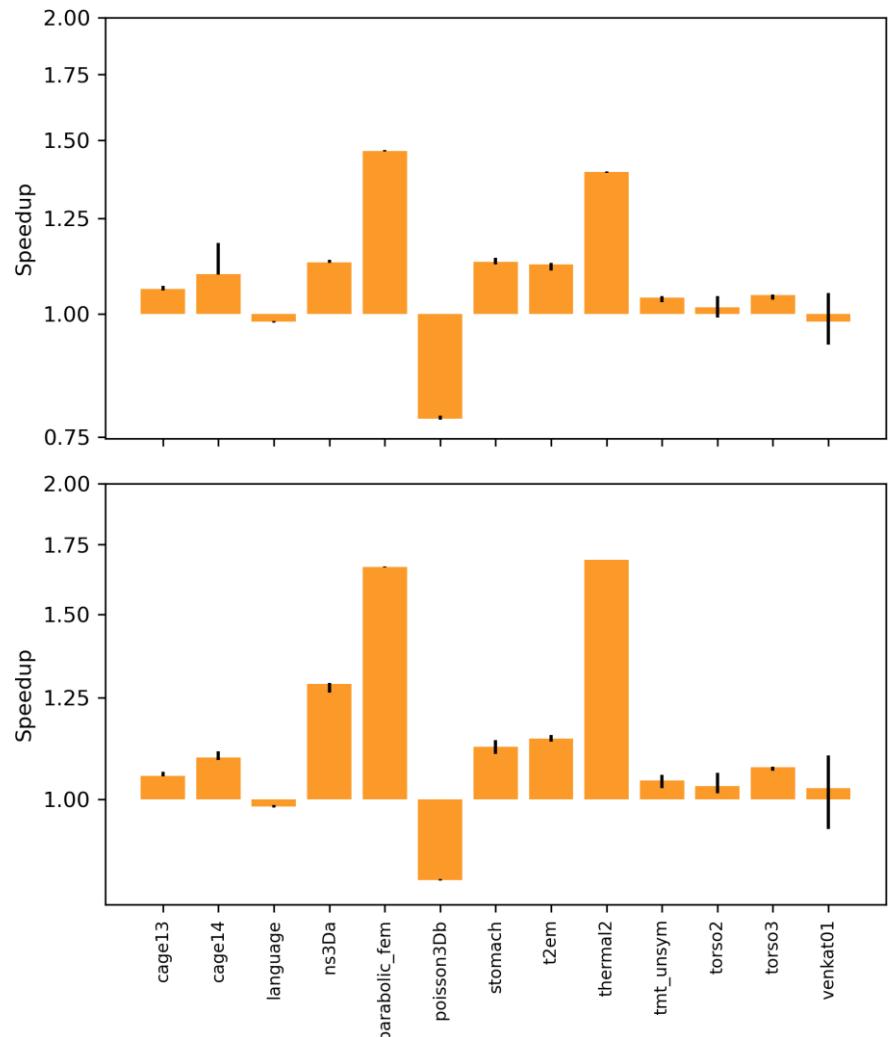
- Speedups
 - Median time of 3 run
 - 3 runs
 - Error bars: mins and maxes
- Geometric mean of speedup
 - MGS: 14%
 - CGSR: 54%

Performance – ILU(0)



- Speedups
 - Median time of 3 run
 - 3 runs
 - Error bars: mins and maxes
- Geometric mean of speedup
 - MGS: -7%
 - CGSR: -4%

Performance – ILU(0) with Jacobi Solves



- ILU(0) w/ 5 Jacobi iterations for each triangular solve
- Speedups
 - Median time of 3 run
 - 3 runs
 - Error bars: mins and maxes
- Geometric mean of speedup
 - MGS: 8%
 - CGSR: 14%

Future Directions

- Choice of low-precision
 - Half, Bfloat16
 - Compression
- Distributed systems
- Other Krylov methods
- Applications

Conclusions

- When restarted, mixed-precision GMRES often outperforms double-precision GMRES

Extra Slides

Test Configuration Details

- CUDA 10.2.199, Kokkos 3.1.01, GCC 7.3.0
- <https://bitbucket.org/icl/mixed-precision-gmres>
 - tag [TPDS-perf](#)

Publications

- N. Lindquist, P. Luszczek, and J. Dongarra, “Improving the performance of the GMRES method using mixed-precision techniques,” in Driving Scientific and Engineering Discoveries through the Convergence of HPC, Big Data and AI. DOI: [10.1007/978-3-030-63393-6_4](https://doi.org/10.1007/978-3-030-63393-6_4)
- [Submitted] N. Lindquist, P. Luszczek, and J. Dongarra, “Accelerating restarted GMRES with mixed precision arithmetic,” in Transactions on Parallel and Distributed Systems.

Effect on Convergence: Configuration

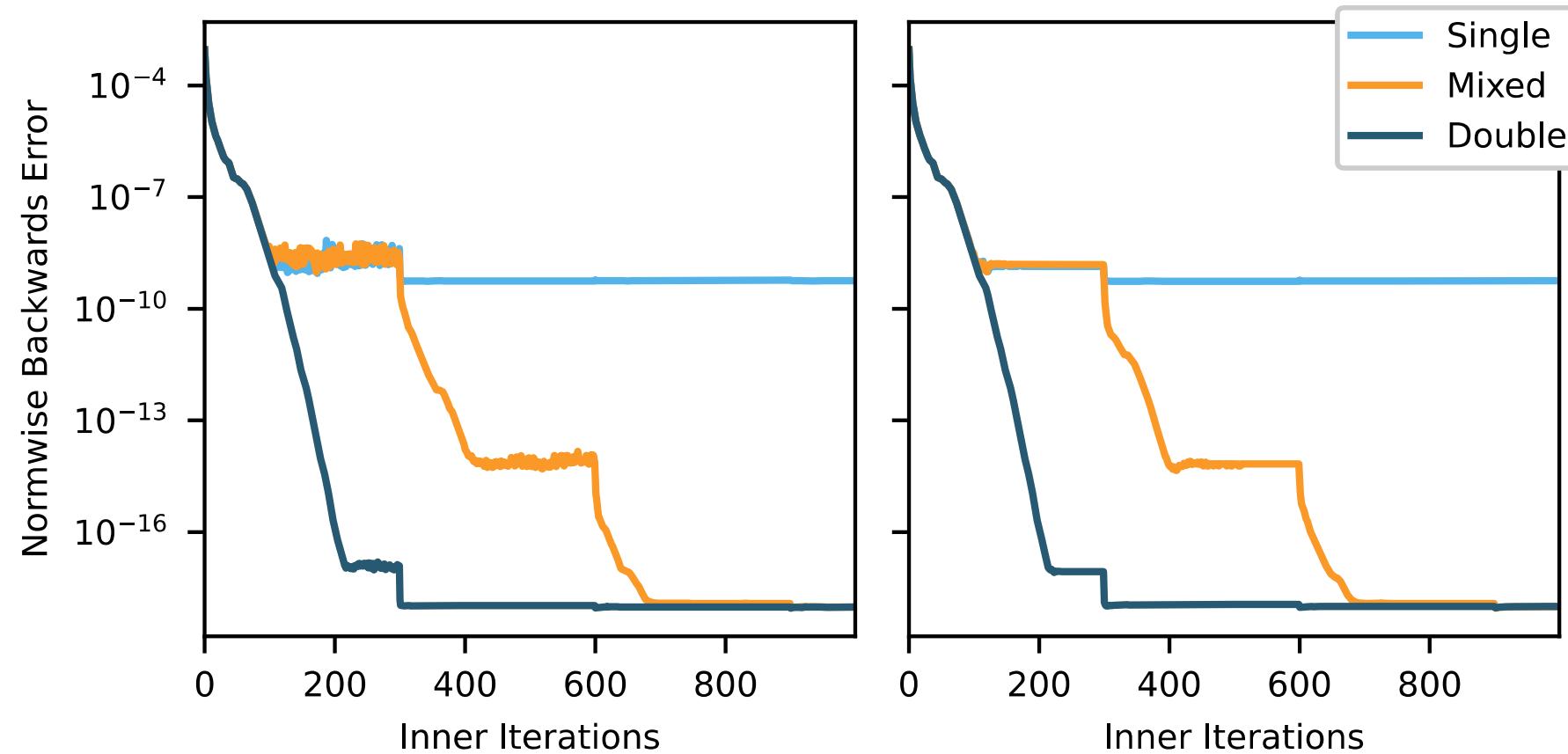
- ILU(0) preconditioner (M^{-1})
- CSR matrix format
- Custom, mixed precision kernels w/ Kokkos
- 20-core Haswell node
 - 2x Intel® Xeon® E5-2650 v3 processors

Effect on Convergence: Configuration

- airfoil_2d from SuiteSparse collection
 - $n = 14,214$
 - $nnz = 259,688$
 - $\kappa_2 = 1.8 \cdot 10^6$
- Error if GMRES stopped

$$\frac{\|b - Ax\|_2}{\|A\|_F \|x\|_2 + \|b\|_2}$$

Accuracy results



Modified Gram-Schmidt
Orthogonalization (MGS)

Classical Gram-Schmidt with
Reorthogonalization (CGSR)