Scalable Algorithms for High-Order Solutions on Common Research Model (CR1)

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Introduction

- □ Finite-element methods (FEM)
 - Compact stencils
 - Clear path to exact linearization
 - Reaching machine convergence is common and expected
 - High number of computation operations per memory fetch
 - More suitable for emerging hardware architectures
 - Seeking to increase accuracy, robustness, and efficiency over more established discretizations
- □ Standardized test cases (TMR, DPW, HLPW, HiOCFD, ...)
- Code verification
- Performance assessment (memory consumption and computational cost)
 - Mesh convergence study
 - Linear and nonlinear convergence study (less studied)
- **Gamma** Focus of this talk: Solution strategies and convergence histories
 - More details on mesh convergence results in Marshall Galbraith's talk

Description of Code

HOMA Solver (High-Order Multilevel Adaptive Solver)

- □ SUPG, RANS, neg-SA, strong and weak implementation of BCs.
- □ Fully implicit with exact linearization through automatic diff.
- □ Non-Linear Strategies:
- Pseudo Transient Continuation (PTC)
- P-multigrid (PMG) solver based on Full Approximation Scheme (FAS)
- Principal Linear Solver: Flexible GMRes (FGMRes)
- Built-ins:
 - Local ILU(k)
 - Implicit Line Relaxation (with Double-CFL Strategy)
 - Additive Schwarz (Restrictive)
- External Packages: PETSc (Used only for comparing with home-developed solvers.)

Description of Code

□ References:

- Ahrabi, B. R. and Mavriplis D. J., "Scalable Solution Strategies for Stabilized Finite-Element Flow Solvers on Unstructured Meshes", 55th AIAA Aerospace Sciences Meeting, AIAA Paper 2017-0517, Dallas, TX, January 2017.
- Ahrabi, B. R. and Mavriplis D. J., "Scalable Solution Strategies for Stabilized Finite-Element Flow Solvers on Unstructured Meshes, Part II", 23rd AIAA Computational Fluid Dynamics Conference, AIAA AVIATION Forum, AIAA Paper 2017-4275, Denver, CO, June 2017.
- Ahrabi, B. R., Brazell, M. J., and Mavriplis D. J., "An Investigation of Continuous and Discontinuous Finite-Element Discretizations on Benchmark 3D Turbulent Flows (Invited)", 2018 AIAA Aerospace Sciences Meeting, Kissimmee, FL, January 2018.

Transonic and Subsonic Turbulent Flow over DPW-6 Configuration HiOCFD5 Coarse Mesh

433,893 DoFs for P1, and 3,401,021 DoFs for P2



Transonic Turbulent Flow over DPW-6 Configuration Flow Visualization

• Free stream conditions: M = 0.85, Re = 5e+6, Alpha = 2.75°



Good shock resolution even on the coarse mesh

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Good shock resolution even on the coarse mesh

Transonic Turbulent Flow over DPW-6 Configuration Aerodynamic Forces

• Free stream conditions: M = 0.85, Re = 5e+6, Alpha = 2.75°



- Meshes are coarse for these simulations
 - Large difference between P1 and P2 solutions

Subsonic Turbulent Flow over DPW-6 Configuration Aerodynamic Forces

• Free stream conditions: M = 0.3, Re = 5e+6, Alpha = 2.75°



- Meshes are coarse for these simulations
 - Large difference between P1 and P2 solutions

Transonic Turbulent Flow over DPW-6 Configuration Convergence Histories

Free stream conditions:

- M = 0.85
- Re = 5e + 6
- Alpha = 2.75°

Nonlinear solver:

• PTC

Linear solver:

- FGMRes
- Relative tol. = 10^{-4}

Linear preconditioner:

- Dual-CFL line solver (PILJ) See Refs. [1,2]
- Maximum number of sweeps per line = 200



Transonic Turbulent Flow over DPW-6 Configuration Convergence Histories



Slight dependency on the mesh resolution,
 But, remember that finer grids are not much finer!

- □ But why line preconditioner? Why not ILU(k)?
- We go with lines because...
 - 1. Strong scalability
 - Linear and nonlinear convergence are independent of number of partitions
 - 2. Significantly less memory (hundreds of times)
 - More suitable for emerging HPC architectures
 - 3. More tunability
 - Increased effectiveness using more iterations in preconditioning
 - 4. More computational efficiency

Slide From AIAA Paper 2017-4275

Implicit Line Preconditioner

- Identify and solve implicitly along strong connections.
- Attempt to reproduce success of line solver observed in FV.
- Works well on 1st order Jacobian matrix BUT not on 2nd order. (Diagonal dominancy)

 $\sim \sim$

To solve

$$[A]x = b$$

$$[A] = [T] + [O]$$
for $k = 1, n_i$

$$r^{k-1} = b - [A]x^{k-1}$$
Solve $[T]dx^k = r^{k-1}$

$$x^k = x^{k-1} + \omega \cdot dx^k$$
end for
$$T$$

Connectivity pattern for a P1 Discretization

Slide From AIAA Paper 2017-4275

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Implicit Line Preconditioner

- Identify and solve implicitly along strong connections.
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• Gauss elimination does not produce any fill-ins

Slide From AIAA Paper 2017-0517

Dual CFL Strategy for Implicit Line Solver

- Newton-Krylov system: $[\mathbf{A}_{CFL_{NK}}]\mathbf{x} = \mathbf{b}$ Solved using FGMRes
- Preconditioner system:

$$[\mathbf{A}_{CFL_{NK}}]\mathbf{x} = \mathbf{b}$$
 Solved using tridiagonal solver

Reformulate the preconditioner system as a defect correction method:

$$\mathbf{r} = \mathbf{b} - [\mathbf{A}_{CFL_{NK}}]\mathbf{x} \qquad n_o \text{ outer iterations}$$
Dual CFL
Strategy
$$\begin{bmatrix} \mathbf{A}_{CFL_{NK}} \end{bmatrix} \Delta \mathbf{x} = \mathbf{r} \\ \begin{bmatrix} \mathbf{A}_{CFL_{Line}} \end{bmatrix} \Delta \mathbf{x} = \mathbf{r} \qquad n_i \text{ inner iterations}$$

$$CFL_{Line} = \min(CFL_{NK}, CFL_{Cap})$$

Solved using block tridiagonal/pentadiagonal solver:

$$[\mathbf{T}]\Delta(\Delta \mathbf{x}) = \mathbf{r} - \left[\mathbf{A}_{\mathrm{CFL}_{\mathbf{Line}}}\right]\Delta \mathbf{x}$$

Explained as a Preconditioned Implicit Line Jacobi (PILJ) method

Comparison of line and ILU(k) preconditioners on the Subsonic Turbulent Flow over DPW-6 Configuration

- Memory efficiency
- Computational efficiency

- Free stream conditions: M = 0.3, Re = 5e+6, Alpha = 4.0°
- HiOCFD5 coarse mesh
- Preconditioner: Line (PILJ)



- Free stream conditions: M = 0.3, Re = 5e+6, Alpha = 4.0°
- HiOCFD5 coarse mesh
- Preconditioner: ILU(k)



 \Box ILU(2) is sufficient to solve the P1 problem)

- Free stream conditions: M = 0.3, Re = 5e+6, Alpha = 4.0°
- HiOCFD5 coarse mesh
- Preconditioner: ILU(k) + Restrictive Additive Schwartz (RAS)



Subsonic Turbulent Flow over DPW-6 Configuration Comparison of Memory Consumption

Fill = ratio of the number of non-zero entries in the factorized matrix over the number of non-zero entries in the original matrix



And this is just for k=3

| PI | PILJ (Line Preconditioner) (Fill = 0.066) | | | | | | | | | | | |
|--------------------------|---|---------------|------------------|---------------------|--|--|--|--|--|--|--|--|
| | | Setting IV | | | | | | | | | | |
| n_p | n_{NL} | $n_{L,tot}$ | k _{max} | t_{tot} | | | | | | | | |
| 180 | 32 | 1269 | 99 | 4.075E4 | | | | | | | | |
| 360 | 32 | 1269 | 99 | 2.067E4 | | | | | | | | |
| 720 | 32 | 1269 | 99 | 1.059E4 | | | | | | | | |
| Setting IV: | | | | | | | | | | | | |
| CFL_{c} | eap = 25 | $50, n_o = 1$ | $0, n_i = 2$ | $0, \omega_L = 0.1$ | | | | | | | | |

| | | | | CM o | ordering | | | | | | |
|---------|-------|----------|-------------|------------------|-----------|----------------------|--------------------|------------------|-----------|--|--|
| | |] | LU (2) | (Fill=12 | .36) | ILU (3) (Fill=22.46) | | | | | |
| Overlap | n_p | n_{NL} | $n_{L,tot}$ | k _{max} | t_{tot} | n_{NL} | n _{L,tot} | k _{max} | t_{tot} | | |
| | 180 | NM | NM | NM | NM | NM | NM | NM | NM | | |
| 0 | 360 | SC | SC | 200 | SC | NM | NM | NM | NM | | |
| | 720 | SC | SC | 200 | SC | 38 | 2721 | 200 | 2.38E4 | | |
| 1 | 180 | NM | NM | NM | NM | NM | NM | NM | NM | | |
| | 360 | SC | SC | 200 | SC | NM | NM | NM | NM | | |
| | 720 | SC | SC | 200 | SC | 32 | 2988 | 200 | 2.44E4 | | |

Subsonic Turbulent Flow over DPW-6 Configuration Comparison of Run Time

• Free stream conditions: M = 0.3, Re = 5e+6, Alpha = 4.0°

HiOCFD5 coarse mesh



Conclusions

- □ Feasibility of a high-order SUPG solver was demonstrated for the transonic flow over common research model.
- □ The effectiveness of the implicit line preconditioner was demonstrated for high-order continuous finite-element methods.
- Nonlinear convergence showed slight dependency on the mesh resolution.
- Memory and computational efficiency of the line preconditioner was compared with ILU(k)

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Thank you! Any questions?

Backup Slide:

- More details on the line preconditioner
- Demonstration of the strong scalability
- Effect of different node orderings on ILU(k)

Lines on P1 and DPW-4 Configuration (Transonic)

- A transonic turbulent test case
- Flow conditions: M = 0.85, Re = 5 million, $alpha = 2.0^{\circ}$
- Comparison of PILJ and ILU(k) on a P1 problem

Transonic Turbulent Flow over DPW-4 Configuration Custom Unstructured Mesh 6,861,035 DOFs for P1



Matrix-based lines shown in red

Flow conditions: M = 0.85, Re = 5 million, $alpha = 2.0^{\circ}$

| | PILJ (Line Preconditioner) | | | | | | | | | | | | |
|-------|----------------------------|--------------------|------------------|-----------|----------|--------------------|------------------|------------------|-------------|-------------|-----------|------------------|--|
| | | Setti | ing I | | | Set | ting II | | Setting III | | | | |
| n_p | n_{NL} | n _{L,tot} | k _{max} | t_{tot} | n_{NL} | n _{L,tot} | k _{max} | t _{tot} | n_{NL} | $n_{L,tot}$ | k_{max} | t _{tot} | |
| 90 | Not tested | | | | | 2801 | 120 | 1.439E5 | 100 | 710 | 49 | 9.709E4 | |
| 180 | | Not t | ested | | 125 | 2801 | 120 | 7.380E4 | 100 | 710 | 49 | 4.979E4 | |
| 360 | | Not t | ested | | 125 | 2801 | 120 | 3.837E4 | 100 | 710 | 49 | 2.589E4 | |
| 720 | SC | SC | 200 | SC | 125 | 2801 | 120 | 2.013E4 | 100 | 710 | 49 | 1.358E4 | |

Nomenclature:

 n_p = Number of processors (and partitions)

 n_{NL} = Number of non-linear steps to reach the full convergence (up to machine precision)

 k_{max} = Maximum number of Krylov vectors used at the linear solutions during the nonlinear steps

 t_{tot} = Total elapsed time in seconds

NC = No convergence (the nonlinear solver was very slow and the simulation was not continued)

SC = Slow convergence (due to frequent failure of the linear solver the nonlinear proceeds with low CFL)

Setting I: CFL_{cap} = 10^8 , $n_o = 5$, $n_i = 10$, $\omega_L = 0.3$ (see Algorithms 1 and 2)

Setting II: CFL_{cap} = 10^3 , $n_o = 5$, $n_i = 10$, $\omega_L = 0.3$ (see Algorithms 1 and 2)

Setting III: $\text{CFL}_{cap} = 10^3, n_o = 10, n_i = 20, \omega_L = 0.3$ (see Algorithms 1 and 2)

Flow conditions: M = 0.85, Re = 5 million, $alpha = 2.0^{\circ}$



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Flow conditions: M = 0.85, Re = 5 million, $alpha = 2.0^{\circ}$



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Flow conditions: M = 0.85, Re = 5 million, $alpha = 2.0^{\circ}$



Flow conditions: M = 0.85, Re = 5 million, $alpha = 2.0^{\circ}$



Convergence behavior is independent of number of partitions → Strong scalability

Transonic Turbulent Flow over DPW-4 Configuration Strong Scaling using PILJ



Can we get the same behavior with ILU(k)? Not always...

Flow conditions: M = 0.85, Re = 5 million, $alpha = 2.0^{\circ}$



Do overlapping or reordering methods solve this issue? Not always...

| CM ordering | | | | | | | | | | | | | |
|---------------------|-------|----------|--------------------|-----------|---------------------|----------|--------------------|-----------|----------------------|----------|--------------------|-----------|------------------|
| ILU (2) (Fill=4.57) | | | | | ILU (3) (Fill=8.08) | | | | ILU (4) (Fill=12.86) | | | | |
| Overlap | n_p | n_{NL} | n _{L,tot} | k_{max} | t _{tot} | n_{NL} | n _{L,tot} | k_{max} | t_{tot} | n_{NL} | n _{L,tot} | k_{max} | t _{tot} |
| 0 | 90 | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM |
| | 180 | 250 | 18078 | 200 | 9.13E4 | NM | NM | NM | NM | NM | NM | NM | NM |
| | 360 | NC | NC | 200 | NC | 115 | 6161 | 200 | 2.32E4 | NM | NM | NM | NM |
| | 720 | NC | NC | 200 | NC | 136 | 8132 | 200 | 1.44E4 | 118 | 7399 | 200 | 1.76E4 |
| 1 | 90 | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM |
| | 180 | NC | NC | 200 | NC | NM | NM | NM | NM | NM | NM | NM | NM |
| | 360 | NC | NC | 200 | NC | 105 | 4499 | 200 | 2.43E4 | NM | NM | NM | NM |
| | 720 | 151 | 7046 | 200 | 1.57E4 | 128 | 6554 | 200 | 1.64E4 | NC | NC | 200 | NC |

Nomenclature:

 n_p = Number of processors (and partitions)

 n_{NL} = Number of non-linear steps to reach the full convergence (up to machine precision)

 k_{max} = Maximum number of Krylov vectors used at the linear solutions during the nonlinear steps

 t_{tot} = Total elapsed time in seconds

NM = Not enough memory to run the case

NC = No convergence (the nonlinear solver stalled before the getting close to the final solution)

Fill = The ratio of the number of non-zero entries in the factorized matrixed over the number of non-zero entries in the original matrix (averaged over all partition numbers)

Overlap = overlap level for the Restrictive Additive Schwarz (RAS) method

| RCM ordering | | | | | | | | | | | | | |
|--------------|-------|---------------------|--------------------|------------------|------------------|---------------------|--------------------|-----------|------------------|----------------------|--------------------|-----------|------------------|
| | | | ILU (2) | (Fill=4.4 | 47) | ILU (3) (Fill=8.04) | | | | ILU (4) (Fill=12.68) | | | |
| Overlap | n_p | n_{NL} | n _{L,tot} | k _{max} | t _{tot} | n_{NL} | n _{L,tot} | k_{max} | t _{tot} | n_{NL} | n _{L,tot} | k_{max} | t _{tot} |
| | 90 | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM |
| 0 | 180 | NC | NC | 200 | NC | NM | NM | NM | NM | NM | NM | NM | NM |
| 0 | 360 | 171 | 15811 | 200 | 3.07E4 | NC | NC | 200 | NC | NM | NM | NM | NM |
| | 720 | NC | NC | 200 | NC | NC | NC | 200 | NC | 128 | 8160 | 200 | 1.48E4 |
| | 90 | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM |
| 1 | 180 | 138 | 7936 | 200 | 5.54E4 | NM | NM | NM | NM | NM | NM | NM | NM |
| 1 | 360 | NC | NC | 200 | NC | NC | NC | 200 | NC | NM | NM | NM | NM |
| | 720 | NC | NC | 200 | NC | NC | NC | 200 | NC | NC | NC | 200 | NC |
| | | | | | | Line of | ordering | | | | | | |
| | | ILU (2) (Fill=4.83) | | | | | ILU (3) | (Fill=8. | 54) | ILU (4) (Fill=13.53) | | | |
| Overlap | n_p | n_{NL} | $n_{L,tot}$ | k_{max} | t _{tot} | n_{NL} | $n_{L,tot}$ | k_{max} | t _{tot} | n_{NL} | $n_{L,tot}$ | k_{max} | t _{tot} |
| | 90 | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM |
| 0 | 180 | NC | NC | 200 | NC | NM | NM | NM | NM | NM | NM | NM | NM |
| 0 | 360 | NC | NC | 200 | NC | NC | NC | 200 | NC | NM | NM | NM | NM |
| | 720 | NC | NC | 200 | NC | NC | NC | 200 | NC | 127 | 9948 | 200 | 2.21E4 |
| | 90 | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM | NM |
| 1 | 180 | NC | NC | 200 | NC | NM | NM | NM | NM | NM | NM | NM | NM |
| 1 | 360 | NC | NC | 200 | NC | 165 | 8148 | 200 | 4.45E4 | NM | NM | NM | NM |
| | 720 | NC | NC | 200 | NC | 193 | 9509 | 200 | 2.78E5 | 171 | 8583 | 200 | 4.05E4 |