PARALLEL IMPLICIT METHODS FOR AERODYNAMICS

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ABSTRACT. Domain decomposition (Krylov-Schwarz) iterative methods are natural for the parallel implicit solution of multidimensional systems of boundary value problems that arise in aerodynamics. They provide good data locality so that even a high-latency workstation network can be employed as a parallel machine. Matrix-free (Newton-Krylov) methods are natural when it is unreasonable to compute or store a true Jacobian. We call their combination Newton-Krylov-Schwarz and report experimental progress on two algorithmic aspects: the use of a coarse grid in additive Schwarz preconditioning and the use of mixed discretization schemes in the (implicitly defined) Jacobian and its preconditioner. Two model problems in two-dimensional compressible flow are considered: the full potential equation and the Euler equations.

1. Krylov-Schwarz Algorithms

Fully implicit linear solvers in aerodynamics allow more rapid asymptotic approach to steady states than time-explicit, approximate factorization, or relaxation solvers that hold the outer nonlinear iteration to small time steps. Nevertheless, the all-to-all data dependencies between the unknown fields in a fully implicit method have led to a resurgence of interest in less rapidly convergent methods in high-latency parallel environments. Resisting, we briefly overview two related efforts that lie along the route to parallel implicit computational aerodynamics. Though the governing equation formulations are mathematically very different – elliptic subsonic full potential and hyperbolic transonic Euler – a common implicit software core allows them to be treated together. Our ultimate interest is in applying Schwarzian domain decomposition techniques to industrial computations still being carried out in these (physically) primitive potential and Euler formulations, and in extending them to Navier-Stokes, for which implicit solvers are even more important. For a variety of reasons, industrial CFD groups are inclining towards the distributed network computing environment characterized by coarse to medium

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granularity, large memory per node, and very high latency, which creates a niche for domain decomposition methods.

Schwarz-preconditioned Krylov solvers for nonsingular linear systems, Ax = b, find the best approximation of the solution x in a small-dimensional subspace that is built up from successive powers of the preconditioned matrix on the initial residual. Such systems, in which A is a Jacobian matrix and b is the nonlinear residual, arise from discretization and linearization of the governing PDEs. A variety of parallel preconditioners, whose inverse action we denote by B^{-1} , can be induced by decomposing the domain of the underlying PDE, finding an approximate representation of A on each subdomain, inverting locally, and combining the results. Generically, we seek to approximate the inverse of A by a sum of local inverses:

$$B^{-1} = \sum_{k} R_{k}^{T} A_{k}^{-1} R_{k},$$

where R_k is a restriction operator that takes vectors spanning the entire space into the smaller dimensional subspace in which A_k is defined.

The simplest domain decomposition preconditioner is block Jacobi, which can be regarded as a zero-overlap form of additive Schwarz [2]. The convergence rate of block Jacobi can be improved, at the price of a higher cost per iteration, with subdomain overlap and (for many problems) by solving an additional judiciously chosen coarse grid system.

2. Newton-Krylov Methods

Evaluation of the discrete residuals of d-dimensional compressible flow formulations requires a large number of arithmetic operations. (For instance, a (d+2)-dimensional eigendecomposition may be required at each grid point.) Their Jacobians, though block-sparse, have dense blocks and are usually an order of magnitude even more complex to evaluate, whether by analytical or numerical means. Hence, matrix-free Newton-Krylov methods, in which the action of the Jacobian is required only on a set of given vectors, are natural in this context. To solve the nonlinear system f(u)=0, given u^0 , let $u^{l+1}=u^l+\lambda^l\delta u^l$, for $l=0,1,\ldots$, until the residual is sufficiently small, where δu^l approximately solves the Newton correction equation $J(u^l)\delta u^l=-f(u^l)$, and λ^l is a damping parameter. The action of Jacobian J on an arbitrary Krylov vector w can be approximated by

$$J(u^l)w pprox rac{1}{\epsilon} \left[f(u^l + \epsilon w) - f(u^l)
ight].$$

Finite-differencing with ϵ makes matrix-free methods potentially much more susceptible to finite word-length effects than ordinary Krylov methods [5]. Steady aerodynamics applications require the solution of linear systems that lack strong diagonal dominance, so a secondary goal of our investigation is to verify that properly-scaled matrix-free methods can be employed in this context. For brevity, details are deferred to a more comprehensive paper. We simply note here that GMRES may have an advantage over other Krylov methods in the matrix-free context in that the vectors w that arise in GMRES have unit two-norm, but may have widely varying scale in other Krylov methods.

An approximation to the Jacobian can be used to precondition the Krylov process. Natural examples are: (1) the Jacobian of a related discretization that allows

economical analytical evaluation of elements, (2) the Jacobian of a lower-order discretization, (3) a finite-differenced Jacobian computed with lagged values for expensive terms, and (4) domain-parallel preconditioners of the form

$$B^{-1} = R_0^T J_{0,u^l}^{-1} R_0 + \sum_{k=1}^K R_k^T J_{k,u^l}^{-1} R_k \text{ , where } J_{k,u^l} = \left\{ \frac{\partial f_i(u^l)}{\partial u_j} \right\}$$

is the Jacobian of f(u) for i and j in subdomain k, and subscript "0" corresponds to a possible coarse grid. The Newton-Krylov-Schwarz method (case 4) can be combined with any other split-discretization technique (cases 1–3), in principle. Right preconditioning of the Jacobian with an operator B^{-1} can be accommodated via

$$J(u^l)B^{-1}w \approx \frac{1}{\epsilon} \left[f((u^l + \epsilon B^{-1}w)) - f(u^l) \right].$$

3. Model Problems

For density ρ , velocity \mathbf{v} , specific internal energy e, and pressure p, the steady Euler equations of inviscid compressible flow are

$$\nabla \cdot (\rho \mathbf{v}) = 0$$
, $\nabla \cdot (\rho \mathbf{v} \mathbf{v} + pI) = 0$, and $\nabla \cdot ((\rho e + p)\mathbf{v}) = 0$.

The full potential equation for velocity potential Φ ,

$$\nabla \cdot (\rho(||\nabla \Phi||) \nabla \Phi) = 0,$$

follows from the additional assumptions of irrotationality, $\mathbf{v} \equiv \nabla \Phi$, and isentropy, $\nabla (p/\rho^{\gamma}) \equiv 0$. The density is given in terms of the potential by

$$ho =
ho_\infty \left(1 + rac{\gamma-1}{2} M_\infty^2 (1 - rac{q^2}{q_\infty^2})
ight)^{rac{1}{(\gamma-1)}},$$

where $q = ||\nabla \Phi||$ and $M_{\infty} = q_{\infty}/a_{\infty}$. Here, a is the sound speed, q the flow speed, and ∞ refers to the freestream. When the flow is everywhere subsonic the full potential formulation fits within the monotone nonlinear elliptic framework of additive Schwarz methods [1]. For a simple non-lifting model problem of an airfoil lying along the symmetry axis y = 0, we choose boundary conditions as follows:

- Upstream and Freestream: $\Phi = q_{\infty}x$ (zero angle of attack),
- Downstream: $\Phi_{,n} = q_{\infty}$,
- Symmetry: $\Phi_{,n} = 0$,
- On the parameterized airfoil with shape y = f(x): $\Phi_{,n} = -q_{\infty}f'(x)$.

The farfield boundary conditions lead to inaccuracies if applied too near the airfoil, but our interest is in algebraic convergence rates.

Coarse Grid	0 × 0	4×5	8 × 9	12×13	16×17	20×21
Analytical	177	35	28	27	24	21
Matrix-free	183	41	28	27	25	23

TABLE 1. Average number of GMRES steps per Newton step for full potential Newton-Krylov-Schwarz solver with varying coarse grid size.

Table 1 shows convergence performance for a fixed-size problem of 128×128 uniform cells with a fixed number of subdomains in an 8 × 8 array as the density of the unnested uniform coarse grid varies. Bilinear rectilinear elements are used for both coarse and fine grids, and bilinear interpolation for intergrid transfers. An overlap 2h is employed on each of the subdomains, which are solved exactly. M_{∞} is 0.1 and the airfoil is the scaled upper surface of a NACA0012. Nonlinear convergence is declared following a 10⁻³ relative reduction in the steady-state residual, which requires only three Newton steps independent of inner linear method. Inner iteration convergence is a relative residual reduction of 10^{-4} . The Krylov solver used throughout this paper is GMRES [7], because of previous comparisons [3] with other modern Krylov solvers on the same problem class that showed CPU cost differences to be small and unsystematic when well-enough preconditioned that any of the methods were practical. Here, we restart GMRES every 20 iterations and precondition on the right, in order to keep the preconditioner out of the residual norm estimates used in the convergence test. Key observations from this example are: (1) even a modest coarse grid makes a significant improvement in an additive Schwarz preconditioner; (2) a law of diminishing returns sets in at roughly one point per subdomain; and (3) matrix-free "matvecs" degrade convergence as much as 15-20% in the less well-conditioned cases.

Precond.	Block Jacobi		Add. Schwarz		Mult. Schwarz	
CFL No.	1	10^{2}	1	102	1	-10^{2}
1 × 1	1	1	1	1	1	1
2×2	4	14	7	14	2	7
4×4	4	18	7	17	3	8
8×8	5	28	10	23	3	8

Table 2. Iteration counts for transonic flow Jacobians at local CFL numbers of 1 and 10^2 , for various preconditioners and decomposition into 4, 16, or 64 subdomains.

Our Euler example is a two-dimensional transonic airfoil flow modeled using an EAGLE-derivative code [6] that employs a finite volume discretization over a body-fitted coordinate grid. Only C-grids of 128×16 or 128×32 cells (from [3]) around a NACA0012 airfoil at an angle of attack of 1.25° and an M_{∞} of 0.8 are considered herein. To obtain a representative matrix/RHS pair on which to test the behavior of Euler Jacobians under Krylov-Schwarz, we first ran a demonstration case from [6] partway to convergence and linearized about the resulting flow state. Following the defect correction practice of [8], a flux vector split scheme is employed for the implicit operators, and f(u) itself is discretized by a flux difference split scheme. Characteristic variable boundary conditions are employed at farfield boundaries using an explicit, first-order accurate formulation. For a given granularity of decomposition, curvilinear "box" decompositions are generally better than curvilinear "strip" decompositions for this problem. Table 2 shows that the zerooverlap results are only slightly less convergent than the corresponding h-overlapped additive Schwarz results at high Courant-Friedrichs-Lewy (CFL) number, and that h-overlapped multiplicative Schwarz is significantly better, though the latter is a much less parallel algorithm. Though we have not yet experimented with a coarse grid in the Euler context, [9] shows that even a piecewise constant coarse grid operator substantially improves Krylov-Schwarz convergence rates in unstructured problems.

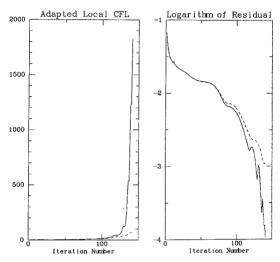


FIGURE 1. CFL and steady-state residual versus iteration count for defect correction and Newton-Krylov solvers.

To test the nonlinear matrix-free approach in a situation with four differently scaled components per gridpoint, we started over and approached the steady solution via a pseudo-transient continuation with an adaptively chosen local CFL number, as described in [3]. Use of the baseline approximate factorization defect correction algorithm produces the dashed curves in Fig. 1. To obtain the solid curves, the explicitly available (Van Leer) flux vector split Jacobian (J_{VL}) is used to precondition the implicitly defined (Roe) flux difference split Jacobian (J_R) at each implicit time step. In matrix terms, the corrections u are obtained as the approximate solutions of, respectively,

$$J_{VL}u = -f_R$$
 and $(J_{VL})^{-1}J_Ru = -(J_{VL})^{-1}f_R$.

Unfortunately, in the retrofit of the existing code, transition to a full Newton method (CFL number approaching infinity) is precluded by explicit boundary conditions, but CFL number can be advanced, as shown in the figure, to $\mathcal{O}(10^3)$ with advantage.

Though space does not permit a meaningful discussion, we mention that both codes have been executed on an ethernet network of workstations using a package of distributed sparse linear system routines developed at Argonne National Laboratory by Gropp and Smith [4], with p4 as the data exchange layer. When exact solvers are used on each subdomain, speedups on a per iteration basis are seen on up to 16 processors, but exact solvers are an extreme case. As a serial preconditioner, global incomplete LU is superior to a Schwarz method using exact subdomain solvers.

4. Conclusions

By concentrating data dependencies locally, domain decomposition preconditioners exploit the two-level memory hierarchy of high-latency distributed-memory architectures. Low-communication zero or small overlaps between the preconditioner blocks are feasible with small convergence rate penalty, at least for intermediate granularities. The addition of a coarse grid has been shown to lead to major iteration count improvements, for a fixed problem size and algebraic residual reduction. Demonstrating the applicability of elliptic-based domain decomposition preconditioners to full potential and Euler problems is only a beginning. Further research will explore the limits of inconsistent preconditioners in matrix-free contexts, the cost versus benefits of the coarse grid solve in parallel contexts, and the relative tuning of inner and outer iteration convergence tolerances.

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