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Scalable Parallel Solver Strategies

> Two-Part Overall Strategy

- Physics-based preconditioning (PBP) is used to reduce the order of matrices and make them diagonally dominant (Chacón). Already implemented.
- Multigrid provides a scalable parallel method for solving diagonally-dominant matrices.

Coarsening and Refining Strategies

- Geometric (GMG): coarser and finer grids. Not suitable for spectral elements.
- Algebraic (AMG): choose largest matrix elements for coarsening. PETSc/Hypre/BoomerAMG. Tested for spectral elements; unsuccessful.
- **Spectral Element (SEMG):** Specifically designed to exploit spectral elements; uses nest spaces of higher and lower polynomial degrees within each grid cell.
- Smoothers and Solvers
 - Jacobi: Based on diagonal dominance of nodal basis representation. Used on all fine levels.
 - **Parallel Direct Solve:** Currently used on the coarsest level. Limits scalability.
 - **AMG:** Will be tested as a scalable replacement for Parallel Direct Solve.

> References

- E. M. Ronquist and A. T. Patera, "Spectral Element Multigrid I: Formulation and Numerical Results," J. Sci. Comput. 2, 4, 389-406 (1987).
- Y. Madera and R. Muñoz, "Spectral Element Multigrid II: Theoretical Justification," J. Sci. Comput. **3**, 4, 323-353 (1988).





Abstract Multigrid Algorithm

Abstract Elliptic Problem

Hilbert space \mathcal{H} , bilinear elliptic form a, linear form g.

Find $v \in \mathcal{H}$ such that $\forall u \in \mathcal{H}, \quad a(u, v) = g(u)$

Nested Finite-Dimensional Subspaces

 $\mathcal{M}_1 \subset \mathcal{M}_2 \subset \cdots \subset \mathcal{M}_j \subset \mathcal{H}$ Simplest case j = 1, 2. Find $v_j \in \mathcal{M}_j$ such that $\forall u \in \mathcal{M}_j, \quad a(u, v_j) = g(u)$

Smoother

 $b(u,v) \approx a(u,v)$, but easier to solve, *e.g.* Jacobi smoother, $b \sim \text{Diag}(a)$.

Multigrid V-Cycle

1. m/2 smoother iterations. Find $S\varphi \in \mathcal{M}_2$ such that

 $\forall u \in \mathcal{M}_2, \quad a(u, \mathcal{S}\varphi - \varphi) = g(u) - a(u, \varphi)$

2. Coarse correction. Find $\bar{\varphi} \in \mathcal{M}_1$ such that

$$\forall u \in \mathcal{M}_1, \quad a(u, \mathcal{S}\varphi - \bar{\varphi}) = g(u) - a(u, \varphi), \quad \mathcal{C}\varphi \equiv \varphi + \bar{\varphi}$$

3. m/2 smoother iterations.

Operator Expression

 $u^1 = \mathcal{S}^{m/2} \mathcal{C} \mathcal{S}^{m/2} u^0$

A COMPACTOR OF



Convergence Theorem

Generalized Eigenvalue Problem

$$\forall u \in \mathcal{M}_1, \quad a(u, \Psi_i) = \lambda_i b(u, \Psi_i)$$

Rescale $b \to \frac{b}{\lambda_{\max}}, \quad 0 < \lambda_1 \le \lambda_2 \le \dots \le \lambda_P = 1$

Nesting Hypothesis

The fine space is the span of all P eigenvectors, $\mathcal{M}_2 = \{\Psi_1, \Psi_2, \dots, \Psi_P\}$. The coarse space is the span of the lowest p eigenvectors, $\mathcal{M}_1 = \{\Psi_1, \Psi_2, \dots, \Psi_p\}$.

Convergence Theorem

Let $e^j \equiv u - u^j$, the error after the *j*th iteration.

 $a(e^1, e^1) \le (1 - \lambda_{p+1})^{2m} a(e^0, e^0)$

Interpretation

The largest "rough" eigenvalues, closest to 1, converge rapidly. The smallest "smooth" eigenvalues, closest to 0, converge slowly.

The coarse correction eliminates the smooth eigenvalues by transferring them to a coarse grid and solving exactly.





Coarsening and Refining

Lobatto Nodal Basis Functions

$$u(x) = u_i \alpha_i(x), \quad x \in (-1, 1), \quad i = 0, \dots, n$$

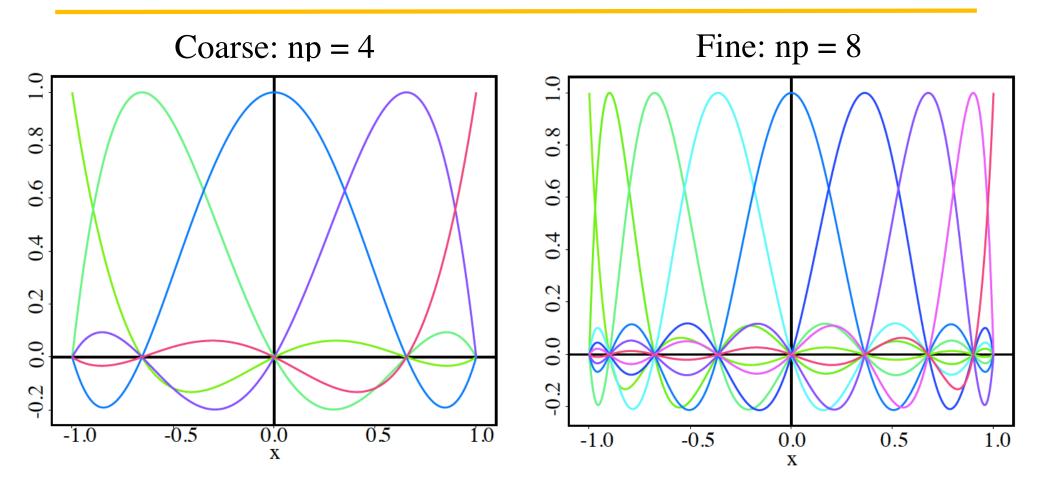
$$\alpha_i(x) \equiv \prod_{j \neq i} \left(\frac{x - \xi_j}{\xi_i - \xi_j} \right), \quad (1 - \xi_i^2) P_n^{(0,0)'}(\xi_i) = 0$$

Coarsening and Refining Operations

$$\begin{split} u^m(x) &= \sum_{i=0}^m u_i^m \alpha_i^m(x), \quad u^n(x) = \sum_{i=0}^n u_i^n \alpha_i^n(x), \quad m < n \\ C_{ij}^{mn} &\equiv \alpha_i(\xi_j^n), \quad u_i^m = C_{ij}^{mn} u_j^n, \quad i = 0, \dots, m, \quad j = 0, \dots, n \\ \mathbf{C}^{mn} &= \{C_{ij}^{mn}\}, \quad \mathbf{u}^m = \mathbf{C}^{mn} \mathbf{u}^n \\ \mathbf{R}^{nm} &= (\mathbf{C}^{mn})^T, \quad \mathbf{u}^n = \mathbf{R}^{nm} \mathbf{u}^n \end{split}$$







Coarsener: evaluating fine basis functions at coarse nodes. Refiner is the transpose of the coarsener





1D Stiffness Matrix

Coarse and Fine Stiffness Matrices

$$u, Lv) \equiv -\int_{-1}^{1} dx u \frac{\partial^2 v}{\partial x^2} = \int_{-1}^{1} dx \frac{\partial u}{\partial x} \frac{\partial v}{\partial x} = u_i A_{ij} v_j$$
$$A_{ij}^m \equiv \int_{-1}^{1} dx \frac{\partial \alpha_i^m}{\partial x} \frac{\partial \alpha_j^m}{\partial x}, \quad i, j = 1, \dots, m$$
$$A_{kl}^n \equiv \int_{-1}^{1} dx \frac{\partial \alpha_k^n}{\partial x} \frac{\partial \alpha_l^n}{\partial x}, \quad k, l = 1, \dots, n$$

$$m < n, m =$$
coarse, $n =$ fine.

Coarsening the Stiffness Matrix

Given the fine stiffness matrix, we can coarsen it with the refining matrix R.

 $u_i^m A_{ij}^m v_j^m = R_{ki}^{nm} u_i^m A_{kl}^n R_{lj}^{nm} v_j^m$

$$A_{ij}^m = R_{ki}^{nm} A_{kl}^n R_{lj}^{nm}, \quad \mathbf{A}^m = (\mathbf{R}^{nm})^T \mathbf{A}^n \mathbf{R}^{nm}$$





Analytical Test Case

1D Poisson Equation

$$-u''(x) = g(x), \quad u(x) = 0 \text{ at } x = \pm 1$$

Convergence Theorem

Using the Lobatto nodal basis functions of degree N, with the scaled diagonal matrix as the smoother, convergence for one element and 2 levels is given by

$$a(u - u^1, u - u^1) = \mu a(u - u^0, u - u^0)$$

$$\mu = \left[1 - \frac{N+2}{2(N-1)}\right]^{2m}$$
$$\mu < 1 \text{ for } N > 2$$





Implementation of SEMG in the 2D HiFi Code

- SEMG is implemented in a Fortran 95 module, 3300 lines of new code written and tested.
- Uses PETSc library for distributed parallel operation and high-level matrix operations.
- Fortran 95 derived types are defined for multigrid levels, diagonal blocks matrices, scatter contexts between native and PETSc vectors, and a full SEMG problem.
- > Coarsening and refining use low-order (np+1) matrices C and R.
- Scaled diagonal Jacobian smoother uses low-order (nqty) matrix **D**.
- ➤ Jacobian scale factor, largest eigenvalue, is computed by power iteration.
- Recursive Fortran 95 subroutine calls itself at successively coarser levels, with a full parallel direct solve on the coarsest level.
- Highly modular code structure, facilitating easy modification, testing, and improvement.



Weak Parallel Scaling Study: Test Case

2D Poisson Equation

 $-\nabla^2 u(x,y) = \rho(x,y)$

Homogeneous Dirichlet Boundary Conditions

$$u(0, y) = u(1, y) = u(x, 0) = u(x, 1) = 0$$

Random Source Term and Initial Conditions

$$\rho(x,y) = \sum_{m=1}^{m_{\max}} \sum_{n=1}^{n_{\max}} \rho_{mn} \sin\left(\frac{mx}{\pi}\right) \sin\left(\frac{ny}{\pi}\right)$$
$$u(x,y) = \sum_{m=1}^{m_{\max}} \sum_{n=1}^{n_{\max}} u_{mn} \sin\left(\frac{mx}{\pi}\right) \sin\left(\frac{ny}{\pi}\right)$$

 ρ_{mn} and u_{mn} random

Weak Scaling Study

Hold the work per processor fixed.

Successively double n_x , n_y , m_{max} , and n_{max} and quadruple the number of parallel cores.

V-cycle: polynomial degree $n_p = 8, 4, 2, 1, 2, 4, 8$

Jacobi iterations per multigrid level = 3 Multigrid cycles to 10^{-5} convergence = 27





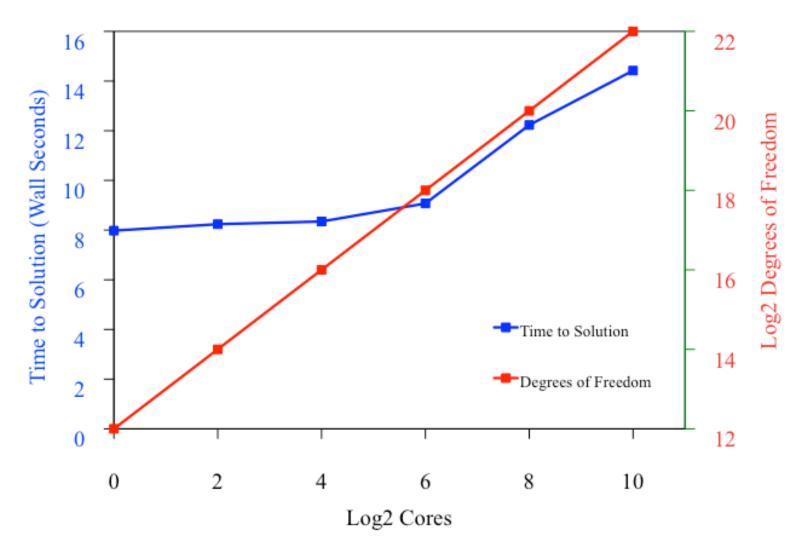
Weak Parallel Scaling Study: Random Solutions

Initial State **Final State** ∞ ∞ $\mathbf{\Sigma}$ > 4Х Х



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2010 APS/DPP Meeting, Glasser & Lukin, Slide 11



Summary and Future Work

- ➢ Formulation and implementation of SEMG are general.
- ➤ Tested for scalability on the 2D Poisson equation.
- ➤ Improvements to be studied:
 - More scalable alternative to the direct coarse solve, *e.g.* AMG.
 - More general boundary conditions.
- ≻ Following this, it will be tested on:
 - Simple linear 2D wave equation.
 - Ideal MHD waves in a periodic plane.
 - GEM challenge problem.
- After completion of 2D development, it will be ported to the 3D HiFi and M3D-C1 codes. The latter will require the development of a different coarsener.



