

Algebraic Multigrid Methods for Complex-Valued Systems

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Complex-Valued Systems

Many applications lead to complex-valued linear systems

- Fourier-domain wave propagation

$$u_{tt} + \alpha u_t = c^2 \Delta u \Rightarrow -c^2 \Delta \hat{u} + i\alpha k \hat{u} - k^2 \hat{u} = 0$$

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- Schrödinger equation
- Quantum dynamics
- Maxwell equations
- Conformal mapping
- Structural dynamics

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How do we solve these problems efficiently?

Krylov Techniques

Core numerical linear algebra easily extends to complex case

- Orthogonality conditions still apply
- Hermitian PD matrices give inner products and norms

Major Krylov methods extend trivially

- GMRES
- BiCG/CGS/BiCGStab
- CG (for Hermitian PD systems)

Other techniques specifically aimed at complex systems

- QMR

What About Preconditioning?

Simple preconditioners extend easily

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- ILU

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- MG & DD for Helmholtz equation
- MG for Maxwell's equations

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What about AMG?

Equivalent Real Forms

Rewrite the complex system $A\mathbf{u} = \mathbf{b}$ as

$$\begin{bmatrix} A^{(R)} & -A^{(I)} \\ A^{(I)} & A^{(R)} \end{bmatrix} \begin{bmatrix} \mathbf{u}^{(R)} \\ \mathbf{u}^{(I)} \end{bmatrix} = \begin{bmatrix} \mathbf{b}^{(R)} \\ \mathbf{b}^{(I)} \end{bmatrix}.$$

Now apply standard algebraic preconditioner to real form

- ILU
- Smoothed Aggregation

Day & Heroux, SISC 2001, **23**:480-498

Vaněk, Mandel, Brezina, Contemp. Math. 1998, **218**:349-356

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Disadvantages:

- Extra cost (double dimension and nnz)
- Lose structure

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Real AMG For Complex-Valued Problems

For many problems, real-valued part is dominant

$$-\Delta u + ik^2 u = f$$

Question: Why not ignore complex part?

- Apply AMG coarsening to $\Re(A)$
- Build interpolation based on $\Re(A)$
- Coarsen A as before, $A_c = P^T A P$

If a dominating real operator can be found, then
preconditioning can be effective

Lahaye et al., IEEE Trans. Magn. 2000, **36**:1535-1538

Lahaye Ph.D. Thesis, KU-Leuven, 2001

Reitzinger et al., J. Comput. Appl. Math. 2003, **155**:405-421

Inherently Complex Operators

What if no dominating real matrix can be found?

Covariant Derivatives:

$$D_{\mu}\psi(\mathbf{x}) = e^{i\omega(\mathbf{x})}\partial_{\mu}\left(e^{-i\omega(\mathbf{x})}\psi(\mathbf{x})\right)$$

$$D_{\mu}^2\psi(\mathbf{x}) = e^{i\omega(\mathbf{x})}\partial_{\mu}^2\left(e^{-i\omega(\mathbf{x})}\psi(\mathbf{x})\right)$$

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Extend individual AMG components to **naturally** handle complex-valued systems

Interpolation

Expanding $A\mathbf{u} = \mathbf{b}$,

$$\begin{bmatrix} A^{(R)} & -A^{(I)} \\ A^{(I)} & A^{(R)} \end{bmatrix} \begin{bmatrix} \mathbf{u}^{(R)} \\ \mathbf{u}^{(I)} \end{bmatrix} = \begin{bmatrix} \mathbf{b}^{(R)} \\ \mathbf{b}^{(I)} \end{bmatrix}.$$

Apply heuristic for real matrices:

Choose P based on symmetric part of \hat{A}

- If A is Hermitian, \hat{A} is symmetric
- If A is complex-symmetric, base P on $A^{(R)}$

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- If A is Hermitian, \hat{A} is symmetric
- If A is complex-symmetric, base P on $A^{(R)}$
- If $B = \imath A$, interpolate differently for A and B

Another Point of View

Form of coarse-grid correction doesn't change

$$\mathbf{e}^{(\text{new})} = (I - PB_c^{-1}RA)\mathbf{e}^{(\text{old})}$$

Still need

- Complementary relaxation and coarse-grid correction
- Algebraically smooth errors in $\text{Range}(P)$

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Use classical AMG interpolation, just with complex values

Restriction

Choice of $R = P^T$ (or $R = P^*$) no longer automatic

- Hermitian problems $\Rightarrow R = P^*$
- Complex-symmetric problems $\Rightarrow R = P^T$
- Complex non-symmetric problems $\Rightarrow R = ???$

Can justify many things...

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Can justify many things...

But don't want to stray too far from AMG

Adjoins

A may not define a norm, but A^*A does

Take $T = (I - M_2^{-1}A)(I - PB_c^{-1}RA)(I - M_1^{-1}A)$, then

$$\|T\|_{A^*A} = \|(A^*A)^{-1}T^*(A^*A)\|_{A^*A}$$

Define the cycle for A^* by

$$\begin{aligned}\bar{T} &= (I - (M_1^{-1})^*A^*) (I - R^*(B_c^{-1})^*P^*A^*) (I - (M_2^{-1})^*A^*) \\ &= (A^*)^{-1} T^* A^*\end{aligned}$$

Then $\|T\|_{A^*A} = \|\bar{T}\|_2$

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R^* must be an effective interpolation operator for A^*

Special Cases

- If $A = A^*$, then can choose $R = P^*$

\Rightarrow Variational Condition

- If $A = A^T$, and interpolation preserves complex conjugation

$\Rightarrow R = P^T$, also Variational

Otherwise, compute restriction separately from interpolation

We always use $R = P^*(A^*)$

Coarse-Grid Selection

Still use strength-of-connection measure

$$S_i = \{j : |a_{ij}| \geq \theta \max_{k \neq i} |a_{ik}|\}$$

Then

- Independent set over graph of strong connections
- Second pass to ensure good AMG interpolation possible

Easy to extend many coarsening schemes, but what makes most sense?

Relaxation

A is an **H-matrix** if $\mathcal{M}(A)$ is an M-matrix,

$$(\mathcal{M}(A))_{ij} = \begin{cases} |a_{ii}| & \text{if } i = j \\ -|a_{ij}| & \text{if } i \neq j \end{cases},$$

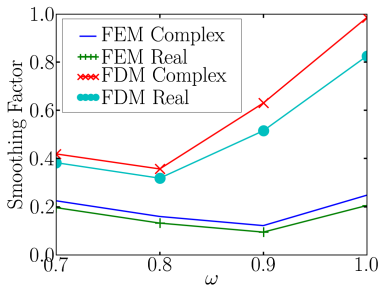
If A is an H-matrix, then

- Jacobi converges for A at least as fast as for $\mathcal{M}(A)$
- Weighted Jacobi converges for all $\omega \leq 1$
- SOR converges for all $\omega \leq 1$

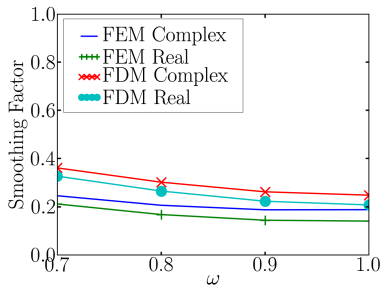
What About Smoothing?

Convergence theory says nothing about smoothing properties

Use local Fourier analysis (LFA) for $-\Delta u + \alpha u$, $\alpha = k^2, k^2 i$



Jacobi



Gauss-Seidel

Time-Harmonic Maxwell Equations

Reduce Maxwell's equations by assuming:

- linear constitutive laws
- low-frequency excitation
- 2D cross-section

$$\Rightarrow -\nabla \cdot \left(\frac{1}{\mu} \nabla \hat{A}_z \right) + i\omega\sigma \hat{A}_z = \hat{J}_{s,z}$$

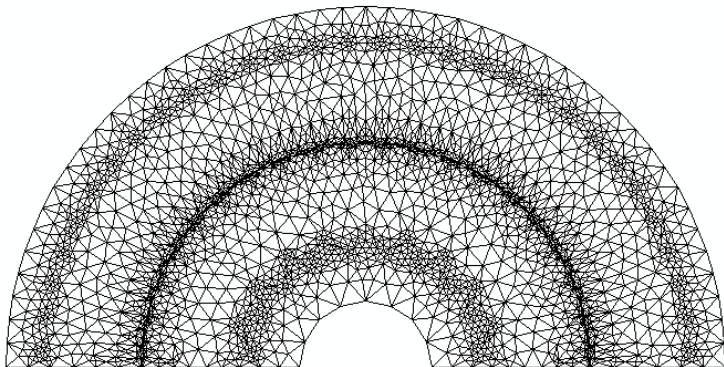
for Fourier-domain potential, $\hat{A} = (0, 0, \hat{A}_z)^T$

Induction Motor

Solve

$$-\nabla \cdot \left(\frac{1}{\mu} \nabla \hat{A}_z \right) + i\omega\sigma \hat{A}_z = \hat{J}_{s,z}$$

on annular geometry



AMG Performance

Problem	Solver	C_A	t_{setup}	t_{solve}	# Iters.
15302 nodes $nnz = 104926$	real AMG	2.86	0.1	0.6	29
	complex AMG	2.85	0.2	0.7	32
	AMG-BiCGStab	2.86	0.1	0.4	9
	cAMG-BiCGStab	2.85	0.2	0.3	8
34555 nodes $nnz = 239661$	real AMG	2.91	0.4	1.7	31
	complex AMG	2.91	0.4	1.7	30
	AMG-BiCGStab	2.91	0.4	1.0	8.5
	cAMG-BiCGStab	2.91	0.4	1.0	8.5
75951 nodes $nnz = 529317$	real AMG	2.87	1.0	4.5	31
	complex AMG	2.87	1.1	4.2	29
	AMG-BiCGStab	2.87	1.0	2.6	8.5
	cAMG-BiCGStab	2.87	1.1	2.5	8

Lattice Gauge Theory

Gauge theories model physics at quantum scales

Lattice gauge theory is discrete form of **standard model**

- Model of interactions between elementary particles
- Includes electromagnetism, weak force, strong force
- Consistent with known particle accelerator experiments

Goal: Use simulation to predict behavior out of reach of experiment

Covariant Laplacian

Conservation laws play important role in gauge theory

- Derivatives are always covariant

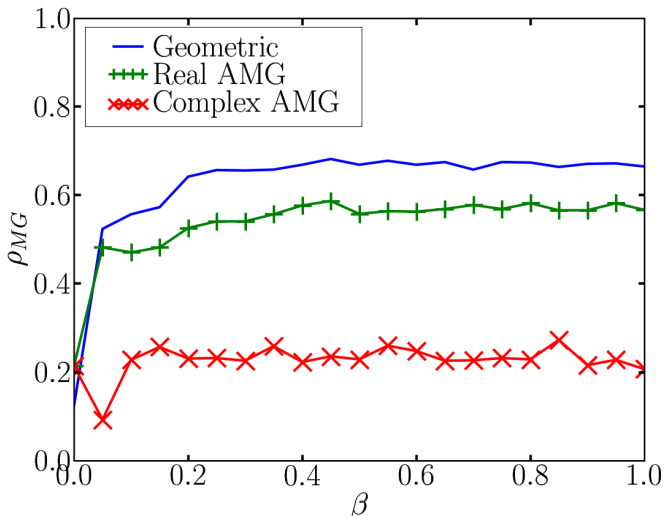
$$D_\mu \psi(\mathbf{x}) = e^{i\omega(\mathbf{x})} \partial_\mu (e^{-i\omega(\mathbf{x})} \psi(\mathbf{x}))$$

- Discretizations are always consistent

Model Problem: Covariant Laplacian

$$\sum_{\mu} D_{\mu}^2 \psi(\mathbf{x}) \Rightarrow \begin{bmatrix} & -e^{i\beta\phi(\mathbf{x})} & \\ -e^{-i\beta\theta(\mathbf{x} - \begin{pmatrix} 1 \\ 0 \end{pmatrix})} & 4 & -e^{i\beta\theta(\mathbf{x})} \\ & -e^{-i\beta\phi(\mathbf{x} - \begin{pmatrix} 0 \\ 1 \end{pmatrix})} & \end{bmatrix}$$

Convergence



Shifted Covariant Laplacian

Two reasons to consider shifting

- As β increases, discrete problem becomes better conditioned
- Physical operators always appear with negative-definite shift

$$\sum_{\mu} D_{\mu}^2 \psi - m^2 \psi \Rightarrow \begin{bmatrix} & -e^{i\beta\phi(\mathbf{x})} & \\ -e^{-i\beta\theta(\mathbf{x} - \begin{pmatrix} 1 \\ 0 \end{pmatrix})} & 4 - m^2 & -e^{i\beta\theta(\mathbf{x})} \\ & -e^{-i\beta\phi(\mathbf{x} - \begin{pmatrix} 0 \\ 1 \end{pmatrix})} & \end{bmatrix}$$

Shifted Covariant Laplacian

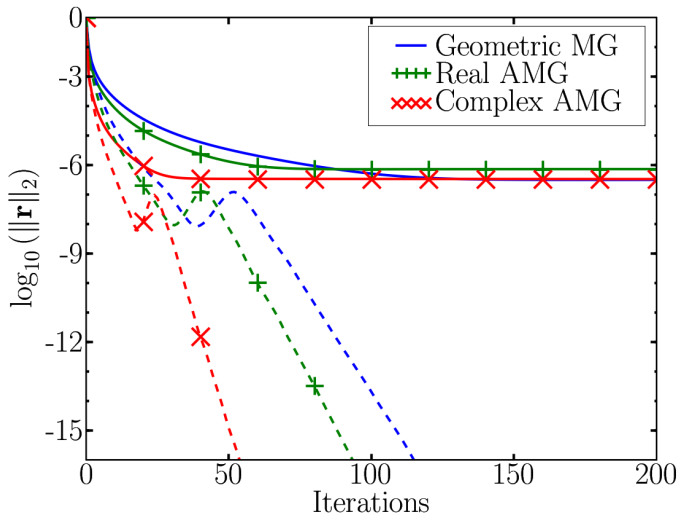
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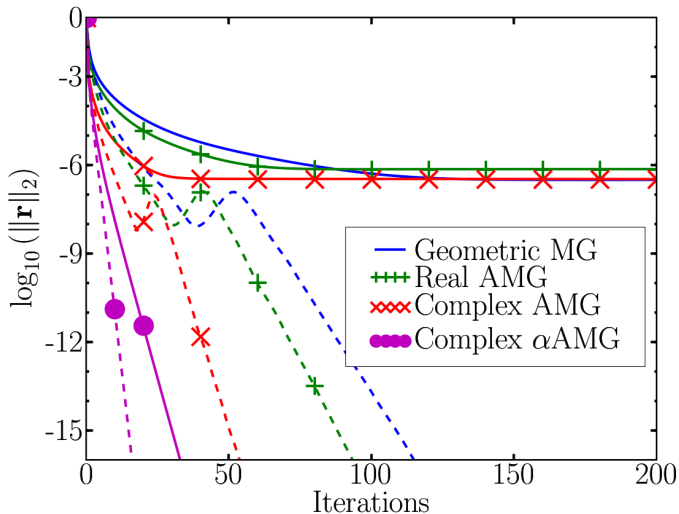
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Shifting **changes energy but not form** of algebraically smooth errors

Convergence histories



Convergence histories



Summary

- Natural extension of AMG to complex arithmetic
- Consistent choice of restriction for special cases
- Local Fourier analysis confirms algorithmic choices
- Performance similar to real AMG for complex problems

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Future Work

- Better understand coarse-grid selection (real and complex)
- Extend to systems, distributed relaxation
- Provide solvers for quantum dynamical simulation