

Multilevel Krylov Methods

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Deflation

Deflation, DD, MG

Multilevel Krylov
methods

Numerical
examples

MK methods for
Helmholtz
equation

AMK methods

Conclusion

Outline

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Deflated CG

Nicolaides 1987, Mansfield 1988, 1990, Kolotilina 1998,
Vuik, Segal, and Meijerink 1999, Morgan 1995, Saad,
Yeung, Erhel, and Guyomarch 2000, Frank and Vuik
2001, Blaheta 2006

Deflation and restarted GMRES

Morgan 1995, Erhel, Burrage, and Pohl 1996, Chapman
and Saad 1997, Eiermann, Ernst, and Schneider 2000,
Morgan 2002

Clemens et al. 2003,2004, de Sturler et al. 2006,
Aksoylu, H. Klie, and M.F. Wheeler 2007

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Deflation with general vectors

A symmetric positive definite

$$Z = [z_1, \dots, z_r] \quad \text{rank} Z = r \quad E = Z^T A Z$$

$$P_D = I - A Z E^{-1} Z^T, \quad Z \in \mathbb{R}^{n \times r},$$

$$P_D A Z = 0$$

$$\text{spectrum}(P_D A) = \{0, \dots, 0, \mu_{r+1}, \dots, \mu_n\}$$

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Deflation for linear systems

$$Z \in \mathbb{R}^{n \times r} \quad Z = [z_1, \dots, z_r] \quad \text{rank} Z = r$$

$$Ax = b \quad P_D = I - AZE^{-1}Z^T$$

We have: $x = (I - P_D^T)x + P_D^T x$ Compute both!

1. $(I - P_D^T)x = Z(Z^T AZ)^{-1}Z^T b$
2. Solve $P_D A \tilde{x} = P_D b$ preconditioner M^{-1} :
 $M^{-1} P_D A \tilde{x} = M^{-1} P_D b$
3. Build $P_D^T \tilde{x} = P_D^T x$

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Deflation

M^{-1} preconditioner, ILU Z approx. eigenvectors

$$ZE^{-1}Z^T$$

Domain decomposition

M^{-1} add. Schwarz Z grid transfer operator

$$ZE^{-1}Z^T \text{ coarse grid correction}$$

Multigrid

M^{-1} smoother Z grid transfer operator

$$ZE^{-1}Z^T \text{ coarse grid correction}$$

Name	Method	Operator
PREC	Traditional Preconditioned CG	M^{-1}
AD	Additive Coarse Grid Correc.	$M^{-1} + Q$
DEF1	Deflation Variant 1	$M^{-1}P_D$
DEF2	Deflation Variant 2	$P_D^T M^{-1}$
A-DEF1	Adapted Deflation Variant 1	$M^{-1}P_D + Q$
A-DEF2	Adapted Deflation Variant 2	$P_D^T M^{-1} + Q$
BNN	Abstract Balancing	$P_D^T M^{-1} P_D + Q$
R-BNN1	Reduced Balancing Variant 1	$P_D^T M^{-1} P_D$
R-BNN2	Reduced Balancing Variant 2	$P_D^T M^{-1}$

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$$Q = ZE^{-1}Z^T = Z(Z^T AZ)^{-1}Z^T$$

Name	Method	Operator
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$$Q = ZE^{-1}Z^T = Z(Z^T AZ)^{-1}Z^T$$

Nabben, Vuik 04, Nabben, Vuik 06, Nabben Vuik 08

Tang, Nabben, Vuik, Erlangga 07

Tang, MacLachlan, Nabben, Vuik 08

Non-symmetric Problems

- ▶ Erlangga, Nabben 06:

$$Z^T \rightarrow Y^T \quad E \rightarrow Y^T A Z$$

$$P_D = I - A Z E^{-1} Y^T$$

$$P_D^T \rightarrow Q_D = I - Z E^{-1} Y^T A$$

$$P_{BNN} = Q_D M^{-1} P_D + Z E^{-1} Y^T$$

$$\|M^{-1}(b - Au_{k,D})\|_2 \leq \|M^{-1}(b - Au_{k,BNN})\|_2.$$

Multilevel Krylov methods (MK methods)

$$Au = b, \quad A \in \mathbb{C}^{N \times N}, \quad u, b \in \mathbb{C}^N.$$

A is in general nonsymmetric, sparse and large

Problems:

- ▶ Diffusion problem (symmetric)
- ▶ Convection-diffusion equation (nonsymmetric)
- ▶ Helmholtz equation (symmetric, indefinite)

Preconditioned system:

$$M_1^{-1} A M_2^{-1} \tilde{u} = M_1^{-1} b, \quad \tilde{u} = M_2 u, \quad M_1, M_2 \text{ nonsingular.}$$

Here,

$$\hat{A} \hat{u} = \hat{b}, \quad \hat{A} := M^{-1} A, \quad \hat{u} := u, \quad \hat{b} := M^{-1} b.$$

Consider

$$P_N = P_D + \lambda_N Z \hat{E}^{-1} Y^T, \quad \hat{E} = Y^T \hat{A} Z,$$

where

$$P_D = I - \hat{A} Z \hat{E}^{-1} Y^T, \quad (\text{Deflation})$$

and solve the system

$$P_N \hat{A} \hat{u} = P_N \hat{b}.$$

- ▶ $Z, Y \in \mathbb{R}^{n \times r}$ are full rank
- ▶ \hat{E} : Galerkin product
- ▶ λ_N Approximation of largest eigenvalue of \hat{A} .

Properties of $P_N \hat{A}$

Spectral relation between $P_D \hat{A}$ and $P_N \hat{A}$.

Theorem

Z, Y are arbitrary rectangular matrices with rank r .

$$\begin{aligned}\sigma(P_D \hat{A}) &= \{0, \dots, 0, \mu_{r+1}, \dots, \mu_N\} \\ \implies \sigma(P_N \hat{A}) &= \{\lambda_N, \dots, \lambda_N, \mu_{r+1}, \dots, \mu_N\}.\end{aligned}$$

► $\sigma(P_N \hat{A})$ is similar to $\sigma(P_D \hat{A})$

Properties of $P_N \hat{A}$

Deflation:

- ▶ $P_D^2 = P_D$ (Projection)
- ▶ $P_D \hat{A} = \hat{A} Q_D$
- ▶ If \hat{A} is symmetric, then $P_D \hat{A}$ is also symmetric

In contrast:

- ▶ $P_N^2 \neq P_N$
- ▶ $P_N \hat{A} \neq \hat{A} Q_N$. However, $\sigma(P_N \hat{A}) = \sigma(\hat{A} Q_N)$
- ▶ $P_N \hat{A}$ is not symmetric even if \hat{A} is symmetric.

Multilevel Krylov method

$$P_N = P_D + \lambda_N Z \hat{E}^{-1} Y^T, \quad \hat{E} = Y^T \hat{A} Z,$$

Need to solve the coarse system with $\hat{A}_H := \hat{E}$.

- ▶ P_N is stable w.r.t. inexact solves.
- ▶ Applying P_N at the “second” level, i.e. use $P_{N,H}$
instead of $\hat{A}_H \hat{x}_H = b_H$
solve: $P_{N,H} \hat{A}_H \hat{x}_H = P_{N,H} b_H$
using a Krylov method
- ▶ With inner Krylov iterations, P_N is i.g. not constant
Use flexible Krylov subspace method (FGMRES,
FQMR, ...)

Multilevel Krylov method

- ▶ The choice of Z and Y

Sparsity of Z and Y ;

May be the same as prolongation and restriction matrices in multigrid

(piece-wise constant, bi-linear interpolation, etc.);

But not eigenvectors;

$$Y = Z;$$

- ▶ About λ_N

Expensive to compute, but an approximate is sufficient:

→ by Gerschgorin's theorem.

Numerical example: 2D Poisson equation

The 2D Poisson equation:

$$\begin{aligned} -\nabla \cdot \nabla u &= g, & \text{in } \Omega \in (0, 1)^2, \\ u &= 0, & \text{on } \Gamma = \partial\Omega. \end{aligned}$$

Discretization: finite differences.

Ω with index set $\mathcal{I} = \{i | u_i \in \Omega\}$.

Ω is partitioned into non-overlapping subdomain Ω_j ,
 $j = 1, \dots, l$, with respective index $\mathcal{I}_j = \{i \in \mathcal{I} | u_i \in \Omega_j\}$.

Then, $Z = [z_{ij}]$:

$$z_{ij} = \begin{cases} 1, & i \in \mathcal{I}_j, \\ 0, & i \notin \mathcal{I}_j. \end{cases}$$

$$Y = Z$$

Numerical example: 2D Poisson equation

Convergence results: relative residual $\leq 10^{-6}$
Gerschgorin estimate for λ_N

N	MK(2,2,2)	MK(4,2,2)	MK(6,2,2)	MK(4,3,3)	MG
32^2	15	14	14	14	11
64^2	16	14	14	14	11
128^2	16	14	14	14	11
256^2	16	14	14	14	11

- ▶ MK(4,2,2,): Multilevel Projection with 4,2,2 FGMRES iterations at level no. 2,3 and 4. etc.
- ▶ MG: Multi Grid (here, V-cycle, one pre- and post RB-GS smoothing, bilinear interpolation)

Observation:

- ▶ h -independent convergence
- ▶ Convergence of MK is comparable with MG.

2D Convection-diffusion equation

The 2D convection-diffusion equation with vertical winds:

$$\begin{aligned}\frac{\partial u}{\partial y} - \frac{1}{Pe} \nabla \cdot \nabla u &= 0, \quad \text{in } \Omega = (-1, 1)^2, \\ u(-1, y) &\approx -1, \quad u(1, y) \approx 1, \\ u(x, -1) &= x, \quad u(x, 1) = 0.\end{aligned}$$

Discretization: Finite volume, upwind discretization for convective term

Z: piece-wise constant interpolation, $Y = Z$

$$\hat{A} = M^{-1}A, \quad M = \text{diag}(A)$$

2D Convection-diffusion equation

Convergence results: relative residual $\leq 10^{-6}$
MK(4,2,2,2), Gerschgorin estimate for λ_N

Grid	<i>Pe</i> :			
	20	50	100	200
128^2	16	16	18	24
256^2	16	16	16	17
512^2	15	16	16	15

- ▶ In MK, FGMRES is used
- ▶ MG (with V-cycle, one pre- and post RB-GS smoothing and bilinear interpolation) does not converge

Observation:

- ▶ Almost h - and Pe -independent convergence

Conclusion - so far

Multilevel Krylov method (MK method)

- ▶ preconditioner M
- ▶ flexible Krylov method
- ▶ multilevel structure (subspace systems)
- ▶ restrictions, prolongations, deflation vectors etc.
- ▶ estimates for λ_N .

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- ▶ h - and Pe -independent convergence

Next:

- ▶ Helmholtz equation
- ▶ algebraic construction of restrictions, prolongations
algebraic MK methods, AMK methods

MK for Helmholtz equation: MKMG method

$$\mathcal{A}u(\mathbf{x}) := - \left(\nabla \cdot \nabla - (1 - \hat{i}\alpha) \left(\frac{\omega}{c} \right)^2 \right) u(\mathbf{x}) = f(\mathbf{x}) \quad (1)$$
$$(\mathbf{x} \in \Omega = (0, L)^2),$$

equipped with radiation condition:

$$\frac{du}{dn} - \hat{i} \frac{\omega}{c} u = 0 \quad (\Gamma = \partial\Omega).$$

- ▶ $\hat{i} = \sqrt{-1}$
- ▶ $0 \leq \alpha \leq 0.1$, attenuative (damping) factor
- ▶ $\omega = 2\pi f$ the angular frequency, with f the temporal frequency
- ▶ $c = c(\mathbf{x})$ the velocity data

Applications: aero- and marineacoustics,
electromagnetics, seismics, etc.

MKMG method

Preconditioner for the Helmholtz equation:
Erlangga, Vuik, Oosterlee, 2004:

$$\mathcal{M} := -\nabla \cdot \nabla - (1 - \beta \hat{i}) \left(\frac{\omega}{c} \right)^2, \quad \beta = (0, 1].$$

Discretization of $\mathcal{M} \rightarrow M$.

M is inverted approximately by one multigrid iteration

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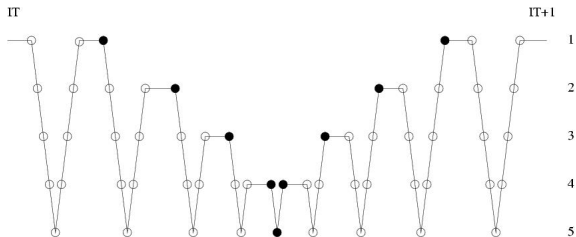
Discretization of $\mathcal{M} \rightarrow M$.

M is inverted approximately by one multigrid iteration

$$\hat{A}_2 := RA_1 M_1^{-1} R^T \approx RA_1 R^T (RM_1 R^T)^{-1} RR^T.$$

$$R = Z^T = Y \quad R^T = P = Z$$

Multilevel Krylov-Multigrid (MKMG) Cycle



●: projection steps, ○: multigrid cycle

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MKMG method

constant $k := \omega L/c$.

g/w means “#grid points per wavelength”.

MKMG(4,2,1)

g/w	k:							
	20	40	60	80	100	120	200	300
15	11	14	15	17	20	22	39	64
20	12	13	15	16	18	21	30	45
30	11	12	12	13	13	15	24	39

MKMG(8,2,1)

g/w	k:							
	20	40	60	80	100	120	200	300
15	11	14	14	17	18	21	27	39
20	12	13	15	14	15	16	20	28
30	11	12	12	12	13	14	15	19

- convergence almost independent of k (with MK(8,2,1))

AMK methods

So far geometric restrictions, prolongations, coarse grids use AMG techniques in the MK method

algebraic MK methods, AMK methods

We used

- ▶ Ruge-Stüben technique
- ▶ agglomeration-based technique

to build R and P , coarse grid matrix : RAP

AMK for 2D Convection-diffusion equation

2D Convection-diffusion equation with rotating flow

$$\nabla \cdot (\vec{c}(x, y)u(x, y)) - \Delta u(x, y) = f(x, y), \quad \Omega \in (0, 1)^2$$

homogeneous Dirichlet boundary conditions, $\vec{c}(x, y)$ is the prescribed velocity vector field

$$c_1(x, y) = -Cxy(1 - x), \quad c_2(x, y) = Cxy(1 - y),$$

where $C = 80$.

Discretization: cell-centered finite volumes, uniform mesh

AMK for 2D Convection-diffusion equation

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	# iterations for mesh:			
	16^2	32^2	64^2	128^2
AMK(4,2,1)-AG	16	18	19	21
AMK(4,2,1)-RS	11	13	20	37
AMK(4,2,1)-AG-M	16	18	19	21
AMK(4,2,1)-RS-M	11	13	19	35
AMG-RS	16	39	87	154

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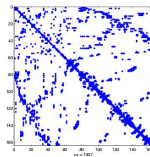
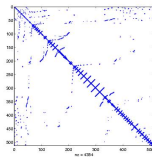
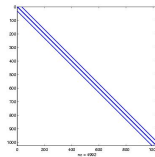
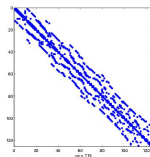
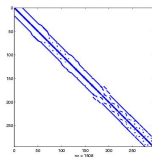
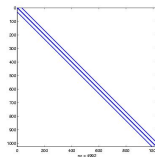
AMK methods

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$$M = \text{diag}(A_l)$$

AMK for 2D Convection-diffusion equation

Aggregation and Ruge-Stüben techniques



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Algebraic Multilevel Krylov methods (AMK methods)

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<http://www.math.tu-berlin.de/~nabben>