

## Convergence properties of preconditioned iterative solvers for saddle point linear systems

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That is, whether we can predict that



#### Motivation

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Other reasons for studying this problem:

Let  $x_k$  be an approximate solution, and  $r_k = b - Ax_k$ .

- Partial stagnation phases occur more frequently (staircase slope)
- Bounds of the type

 $||r_{k+1}|| \le c ||r_k||, \quad 0 < c < 1$ 

important whenever  $\boldsymbol{c}$  independent of problem parameters

 $\Rightarrow$  convergence behavior is not influenced by other model components:  $||r_k|| \le c^k ||r_0||$ 

 $\Rightarrow$  Crucial to design preconditioning techniques

#### Elman bound (PhD thesis, 1982)

Let  $H = (A + A^T)/2$ 

If H is positive definite (i.e.  $\lambda_{\min}(H) > 0$ ), then

$$||r_k|| \le \left(1 - \frac{\lambda_{\min}^2(H)}{\|A\|^2}\right)^{\frac{1}{2}} ||r_{k-1}|| < ||r_{k-1}||$$

so that

$$||r_k|| \le \left(1 - \frac{\lambda_{\min}^2(H)}{||A||^2}\right)^{\frac{k}{2}} ||r_0||$$

Non-Stagnation and Parameter independence

$$||r_k|| \le \left(1 - \frac{\lambda_{\min}^2(H)}{||A||^2}\right)^{\frac{k}{2}} ||r_0||$$

If  $\lambda_{\min}(H)$ , ||A|| independent of parameters (viscosity, meshsize, etc.):

Number of iterations to converge is independent of parameters

- Bound *per se* is not sharp
- Very much used in certain contexts

(e.g. Domain Decomposition methods, cf. Toselli & Widlund 2005)

Related and unrelated bounds

After one iteration of a minimal residual method:

$$||r_1|| = \sqrt{1 - \frac{(r_0^T A r_0)^2}{||Ar_0||^2 \, ||r_0||^2}} ||r_0||$$

...true stagnation is *very* unlikely !

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- Characterization of matrices which lead to complete stagnation (Zavorin etal. 2003)
- Some improvements over this bound for diag.ble/nondiag.ble matrices

(Eisenstat etal. '83, Greenbaum '97, Saad '03, Liesen '00, Freund '90, ...)

• Different bounds, using  $\mathcal{F}(A) \subset \mathbb{C}^+$ 

(Eiermann & Ernst '01, Greenbaum '97, Starke '97)

• Additional results for A normal (s.t.  $AA^T = A^T A$ )

The new non-stagnation condition

Grcar tr'89:

Let  $q_k$  be polynomial with  $q_k(0) = 0$ . If  $\frac{1}{2}(q_k(A) + q_k(A)^T) > 0$  then

$$\|r_k\| \le \left(1 - \frac{\theta_{\min}^2}{\|q_k(A)\|^2}\right)^{\frac{1}{2}} \|r_0\| \quad \theta_{\min} = \lambda_{\min}(\frac{1}{2}(q_k(A) + q_k(A)^T))$$

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We reverse the problem:

We fix  $q_k(t) = t^k$ , k = 2, 4 and determine conditions on A such that Grear's result can be applied

#### Sufficient condition

For  $q_k(t) = t^k$ , k = 2:

If A is such that Grear's result holds, then GMRES cannot stagnate for more than k - 1 = 1 consecutive iterations

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(Similar for k = 4)
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Note: Also relevant for restarted GMRES

DEF. M is positive definite if  $\frac{1}{2}(M + M^T) > 0$ 

Restatement of the problem:

Find conditions on A so that  $q_2(A) = A^2$  is positive definite

The new conditions

- Let  $H = \frac{1}{2}(A + A^T)$ ,  $S = \frac{1}{2}(A A^T)$ .
  - 1. If H is nonsingular, then  $A^2$  is positive definite if and only if  $\|SH^{-1}\| < 1$
  - 2. If S is nonsingular, then  $A^2$  is negative definite if and only if

 $\left\|HS^{-1}\right\| < 1$ 

The new conditions

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 $\|HS^{-1}\| < 1$ 

$$||r_2|| \le \left(1 - \frac{\theta_{\min}^2}{||A^2||^2}\right)^{\frac{1}{2}} ||r_0|| \quad \theta_{\min} = \lambda_{\min}(\frac{1}{2}(A^2 + (A^2)^T)) > 0$$

The same relation holds at every other iteration

#### A simple Sufficient condition

H "dominates" S:

If  $\min_i |\lambda_i(H)| > \max_j |\lambda_j(S)|$ , then  $A^2$  is positive definite

(A corresponding result for  $A^2$  negative definite)

The 
$$k = 4$$
 case

Let  $H = \frac{1}{2}(A + A^T)$ ,  $S = \frac{1}{2}(A - A^T)$ .

- 1. If  $H^2+S^2$  is nonsingular, then  $A^4$  is positive definite if and only if  $\|(HS+SH)(H^2+S^2)^{-1}\|<1$
- 2. If HS + SH is nonsingular, then  $A^4$  is negative definite if and only if

$$||(H^2 + S^2)(HS + SH)^{-1}|| < 1$$

- $\star$  One could continue with higher powers, but ....
- $\star$  There may be other polynomials  $q_k(t)$  such that Grcar's result applies

#### Some Examples

FD discretization of:

$$L(u) = -(\alpha u_{x_1})_{x_1} - (\beta u_{x_2})_{x_2} + \gamma u_{x_1} + \delta u_{x_2} - \eta u$$

size(A) =1600.  $\eta = 100$ .

α	eta	$\gamma$	δ	$\lambda_{\min}(H)$	$\ SH^{-1}\ $
$\exp(-x_1x_2)$	$\exp(x_1x_2)$	-1	-1	-0.04719	0.6194
1	1	$-1/(.1x_1+100x_2)$	0	-0.04775	0.1577
1	1	$1/10(x_1 - x_2)$	0	-0.04772	0.1838
1	1	$1/10(x_1+x_2)$	0	-0.04772	0.5819
1	1	0.2	0	-0.04781	0.5811

Navier-Stokes problem. Flow over a backward facing step IFISS Package (Elman, Ramage, Silvester) Oseen Problem. Uniform grid, Q1-P0 elements, F nonsymmetric

Augmentation block diagonal preconditioning:

$$\mathcal{A} = \begin{bmatrix} F & B^T \\ B & -\beta C \end{bmatrix} \quad \mathcal{P} = \begin{bmatrix} F + B^T \tilde{C}^{-1} B \\ & \tilde{C} \end{bmatrix}$$

Spectrum of  $\mathcal{AP}^{-1}$  tends to cluster around  $\lambda = 1$ ,  $\lambda = -1$  (Cao, 2008)



#### Stokes Problem. Channel domain

IFISS Package (Elman, Ramage, Silvester) uniform grid, Q1-P0 elements, *M* symmetric

Nonsymmetric Preconditioning (cf. Elman, Silvester & Wathen '05):

$$\mathcal{A} = \begin{bmatrix} M & B^T \\ B & \end{bmatrix}, \quad \mathcal{P} = \begin{bmatrix} M & B^T \\ & G \end{bmatrix}, \quad G \approx BM^{-1}B$$

Spectrum of  $\mathcal{AP}^{-1}$  tends to cluster around  $\lambda = 1, -1$ 



#### Symmetric Saddle-Point type Problem

Nonsymmetric version (cf. survey: Benzi, Golub & Liesen '05):

$$\mathcal{A}_{-} = \begin{bmatrix} \mu I & B^T \\ -B & 0 \end{bmatrix}, \quad \mu > 0$$

Spectrum of  $\mathcal{A}_{-}$  is in  $\mathbb{C}^{+}$ , but  $\frac{1}{2}(\mathcal{A}_{-} + \mathcal{A}_{-}^{T}) \geq 0$ 



#### Conclusions

New conditions for non-stagnation:
Useful to establish parameter independence

• Possibility to extend the result

Reference

V. Simoncini and Daniel B. Szyld

New conditions for non-stagnation of minimal residual methods

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