

A new analysis of block preconditioners for saddle point problems

Yvan Notay*

Université Libre de Bruxelles
Service de Métrologie Nucléaire

Delft, June 14, 2013

* Supported by the Belgian FNRS
<http://homepages.ulb.ac.be/~ynotay>

1. Introduction
2. Taxonomy of block preconditioners
3. Relations between the preconditioners
4. Eigenvalue analysis
5. Example of application
6. Conclusions

Linear systems

$$K \mathbf{u} = \mathbf{b}$$

where K has saddle point structure:

$$K = \begin{pmatrix} A & B^T \\ B & -C \end{pmatrix}$$

In this talk: A SPD, C sym. nonnegative definite
(all results compatible with $C = 0$)

Linear systems

$$K \mathbf{u} = \mathbf{b}$$

where K has **saddle point** structure:

$$K = \begin{pmatrix} A & B^T \\ B & -C \end{pmatrix}$$

In this talk: A SPD, C sym. nonnegative definite
(all results compatible with $C = 0$)

■ Applications

Stokes problems (stationnary or time dependent),
PDE-constrained optimization, etc

- All results presented here can be applied to any **context** (purely algebraic analysis)

- We are interested in iterative solution schemes combining a Krylov subspace method with **block preconditioners**
- A preconditioner is any procedure that provides a cheap approximation to the inverse of the system matrix K
- When the preconditioner is SPD, MINRES is a choice method

If the preconditioner is nonsymmetric or indefinite, GMRES or GCR has to be used instead

Block preconditioners

They are rooted in / equivalent to **fractional steps iterative methods**: approximately solve

$$\begin{pmatrix} A & B^T \\ B & -C \end{pmatrix} \begin{pmatrix} \mathbf{v} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{b}_v \\ \mathbf{b}_p \end{pmatrix}$$

alternating approximates solves of

$$A \delta \mathbf{v} = \mathbf{b}_v - A \hat{\mathbf{v}} - B^T \hat{\mathbf{p}} ;$$

i.e.:

$$\delta \mathbf{v} = M_A^{-1} (\mathbf{b}_v - A \hat{\mathbf{v}} - B^T \hat{\mathbf{p}})$$

and “pressure corrections”:

$$\delta \mathbf{p} = \pm M_S^{-1} (\mathbf{b}_p - B \bar{\mathbf{v}} + C \bar{\mathbf{p}})$$

These operations can be combined in a number of ways:

- different orders
- some may be repeated
- different possibilities to define \hat{v} , \hat{p} , \bar{v} , \bar{p} in function of previous and current approximations

These combinations give rise to the different types of block preconditioners

The schemes have in common that they need:

- A (good) preconditioner M_A for A (possibly defined only implicitly)
- An operator M_S governing the pressure correction

We'll see (well known fact) that it needs to be a (good) preconditioner for the Schur complement

$$S = C + B A^{-1} B^T$$

The schemes have in common that they need:

- A (good) preconditioner M_A for A (possibly defined only implicitly)
- An operator M_S governing the pressure correction

We'll see (well known fact) that it needs to be a (good) preconditioner for the Schur complement

$$S = C + B A^{-1} B^T$$

Types of contributions in the field

- Identify relevant M_A , M_S for a given application area
- Analyse one or several schemes assuming that relevant M_A and M_S have been obtained

2. Taxonomy of block preconditioners

Block Diagonal⁺:

$$M_+ = \begin{pmatrix} M_A & \\ & M_S \end{pmatrix}$$

Block Diagonal:

$$M_d = \begin{pmatrix} M_A & \\ & -M_S \end{pmatrix}$$

Block Triangular:

$$M_t = \begin{pmatrix} M_A & B^T \\ & -M_S \end{pmatrix}$$

Inexact Uzawa:

$$M_u = \begin{pmatrix} M_A & \\ B & -M_S \end{pmatrix}$$

Block Approximate Factorization:

$$M_f = \begin{pmatrix} M_A & \\ B & -M_S \end{pmatrix} \begin{pmatrix} I & M_A^{-1} B^T \\ & I \end{pmatrix}$$

2. Taxonomy of preconditioners (cont.)

\widetilde{M}_A such that:

$$I - \widetilde{M}_A^{-1}A = (I - M_A^{-1}A)^2$$



Block Triangular⁽²⁾:

$$M_{t_2} = \begin{pmatrix} \widetilde{M}_A & B^T \\ & -M_S \end{pmatrix}$$

Inexact Uzawa⁽²⁾:

$$M_{u_2} = \begin{pmatrix} \widetilde{M}_A & \\ B & -M_S \end{pmatrix}$$

Block SGS:

$$M_g = \begin{pmatrix} I & & \\ B & M_A^{-1} & I \end{pmatrix} \begin{pmatrix} \widetilde{M}_A & \\ & -M_S \end{pmatrix} \begin{pmatrix} I & M_A^{-1}B^T \\ & I \end{pmatrix}$$

3. Relations between the preconditioners.

Remark

The eigenvalues of $M_+^{-1}K$ are real (**positive and negative**)
For all other prec., the eigenvalues have positive real part

Theorem

(1)

$$\begin{aligned} \max_{\lambda \in \sigma(M_d^{-1}K)} |\lambda| &\leq \max_{\lambda \in \sigma(M_+^{-1}K)} |\lambda| \\ \min_{\lambda \in \sigma(M_d^{-1}K)} |\lambda| &\geq \min_{\lambda \in \sigma(M_+^{-1}K)} |\lambda| \end{aligned}$$

3. Relations between the preconditioners.

Remark

The eigenvalues of $M_+^{-1}K$ are real (positive and negative)
For all other prec., the eigenvalues have positive real part

Theorem

(1)

$$\begin{aligned} \max_{\lambda \in \sigma(M_d^{-1}K)} |\lambda| &\leq \max_{\lambda \in \sigma(M_+^{-1}K)} |\lambda| \\ \min_{\lambda \in \sigma(M_d^{-1}K)} |\lambda| &\geq \min_{\lambda \in \sigma(M_+^{-1}K)} |\lambda| \end{aligned}$$

(2) $M_t^{-1}K$ and $M_u^{-1}K$ have the same spectrum

3. Relations between the precondition.

Remark

The eigenvalues of $M_+^{-1}K$ are real (**positive and negative**)
For all other prec., the eigenvalues have positive real part

Theorem

(1)

$$\begin{aligned} \max_{\lambda \in \sigma(M_d^{-1}K)} |\lambda| &\leq \max_{\lambda \in \sigma(M_+^{-1}K)} |\lambda| \\ \min_{\lambda \in \sigma(M_d^{-1}K)} |\lambda| &\geq \min_{\lambda \in \sigma(M_+^{-1}K)} |\lambda| \end{aligned}$$

(2) $M_t^{-1}K$ and $M_u^{-1}K$ have the same spectrum

(3) $M_g^{-1}K$, $M_{t_2}^{-1}K$ and $M_{u_2}^{-1}K$ have the same spectrum

4. Eigenvalue analysis

The real eigenvalues λ of $M_*^{-1}K$ satisfy: $\underline{\xi} \leq \lambda \leq \bar{\xi}$

$$\underline{\mu} = \lambda_{\min}(M_A^{-1}A) \qquad \bar{\mu} = \lambda_{\max}(M_A^{-1}A)$$

$$\underline{\nu} = \lambda_{\min}(M_S^{-1}S) \qquad \bar{\nu} = \lambda_{\max}(M_S^{-1}S)$$

$\underline{\xi}$

$\bar{\xi}$

M_d	$\min(\underline{\mu}, \underline{\nu})$	$\max(\bar{\mu}, \bar{\nu})$	
M_u, M_t $(\bar{\mu} \leq 1)$	$\min(\underline{\mu}, \underline{\nu})$	$\max(1, \bar{\nu})$	$\eta \approx 1$
M_u, M_t $(\bar{\mu} > 1)$	$\min(\underline{\mu}, \eta^{-1} \underline{\nu})$	$\tilde{\nu}$	$\tilde{\nu} \geq \bar{\mu}, \bar{\nu}$
M_f	$\min(\underline{\mu}, \underline{\nu})$	$\max(\bar{\mu}, \bar{\nu})$	$\tilde{\nu} \leq \bar{\mu}(\bar{\nu} + 1)$
M_g M_{u_2}, M_{t_2}	$\min(1 - \rho_A^2, \underline{\nu})$	$\max(1, \bar{\nu})$	$\rho_A = \rho(I - M_A^{-1}A)$

4. Eigenvalue analysis

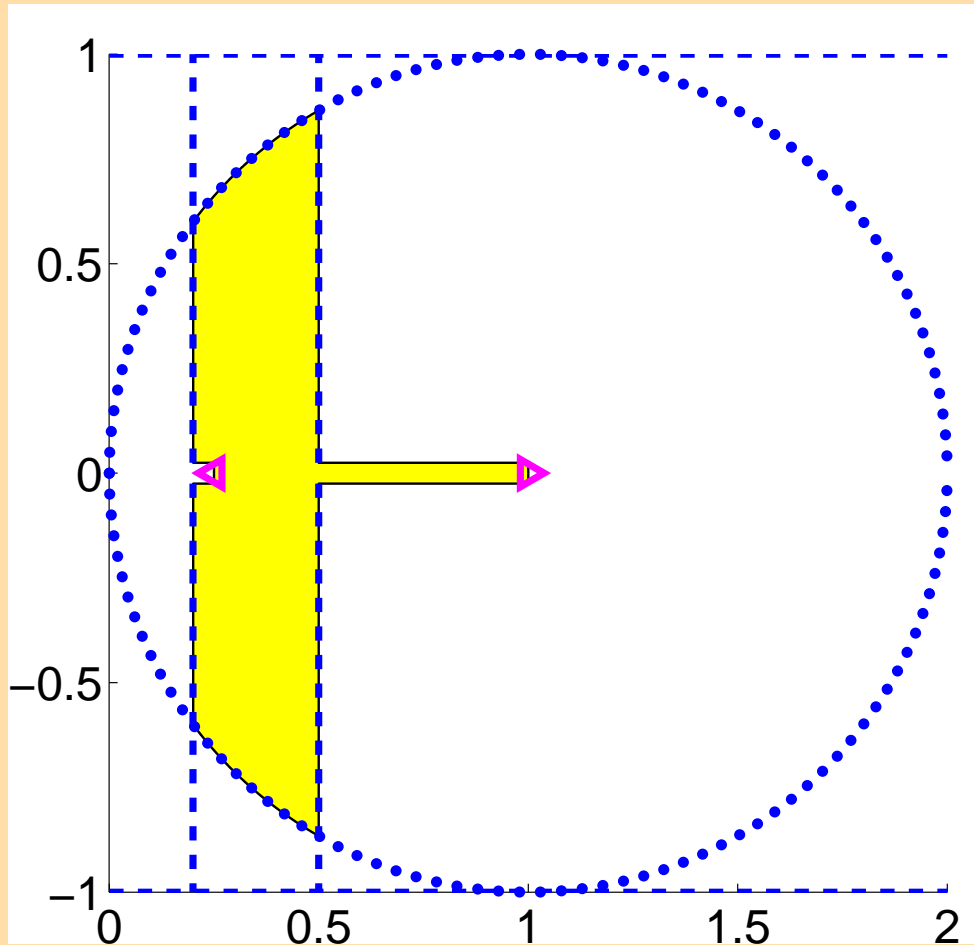
The eigenvalues with nonzero imaginary part satisfy

$$\underline{\chi} \leq \Re(\lambda) \leq \bar{\chi}, \quad |\Im(\lambda)| \leq \delta, \quad |\lambda - \zeta| \leq \zeta$$

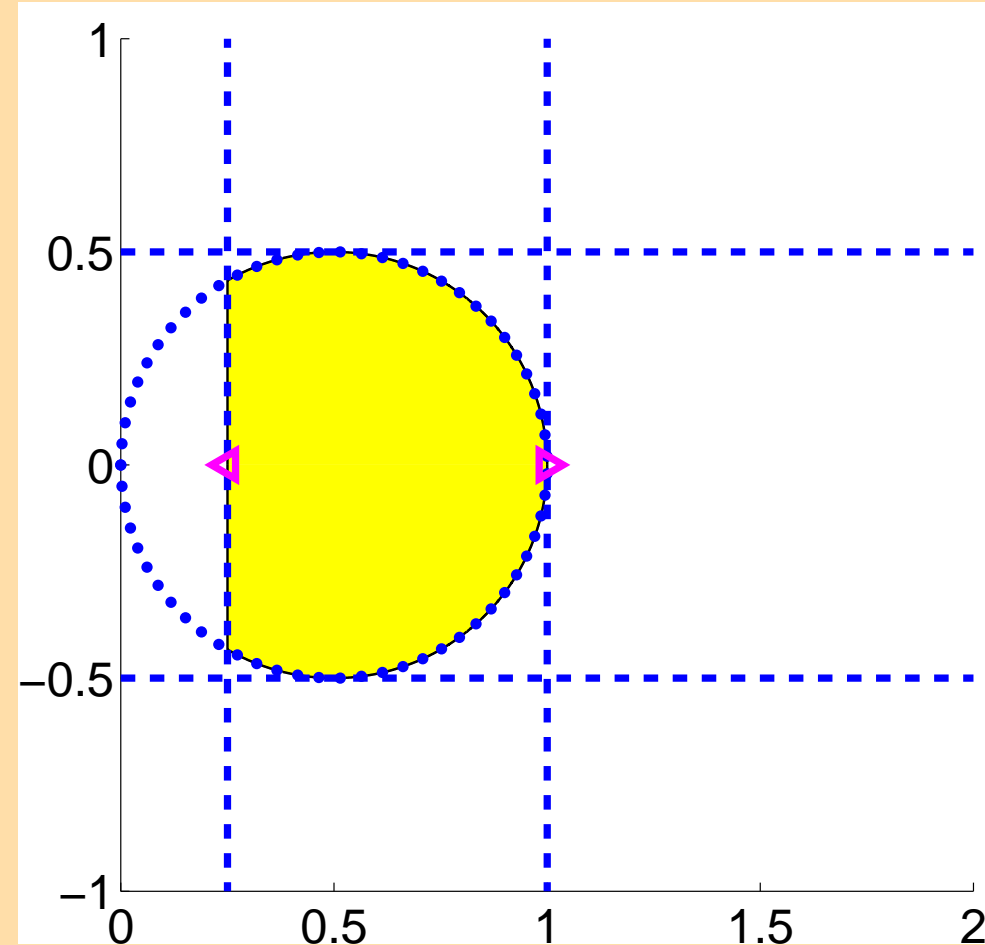
4. Eigenvalue analysis

The eigenvalues with nonzero imaginary part satisfy

$$\underline{\chi} \leq \Re(\lambda) \leq \bar{\chi}, \quad |\Im(\lambda)| \leq \delta, \quad |\lambda - \zeta| \leq \zeta$$



Block Diagonal



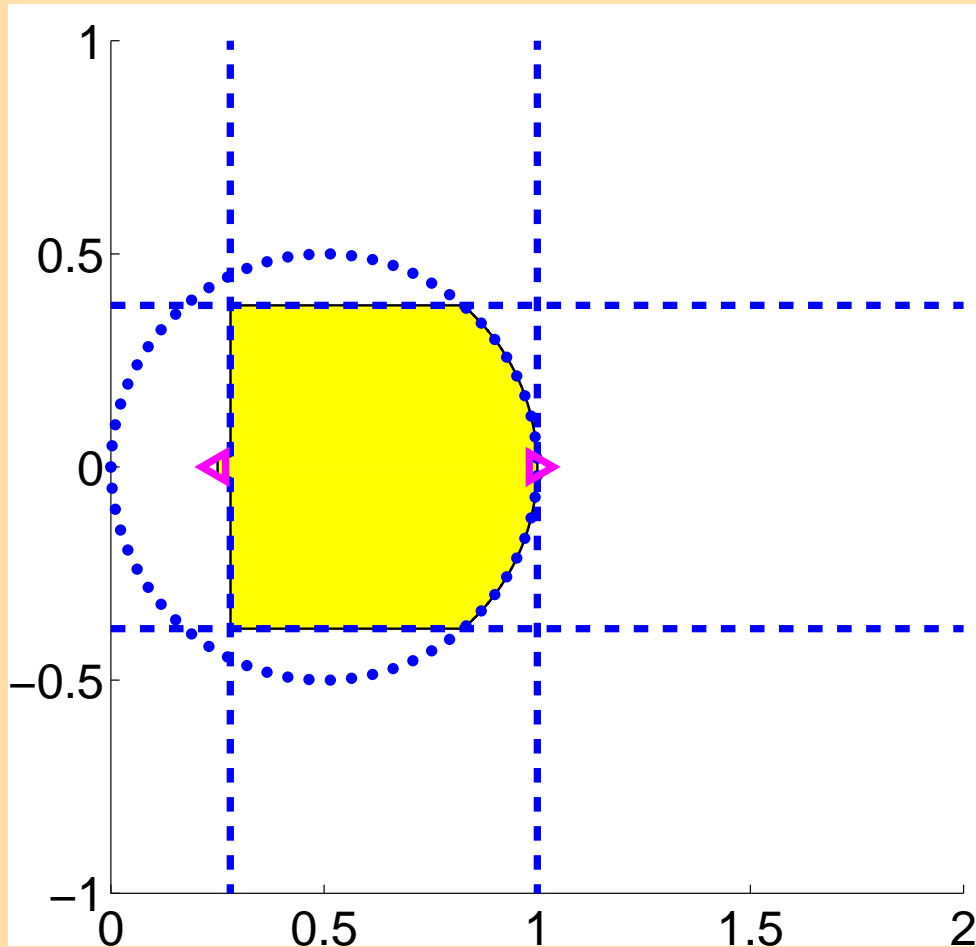
Block Triang. & Uzawa

$$\underline{\mu} = 0.4, \quad \underline{\nu} = 0.25, \quad \bar{\mu} = \bar{\nu} = 1$$

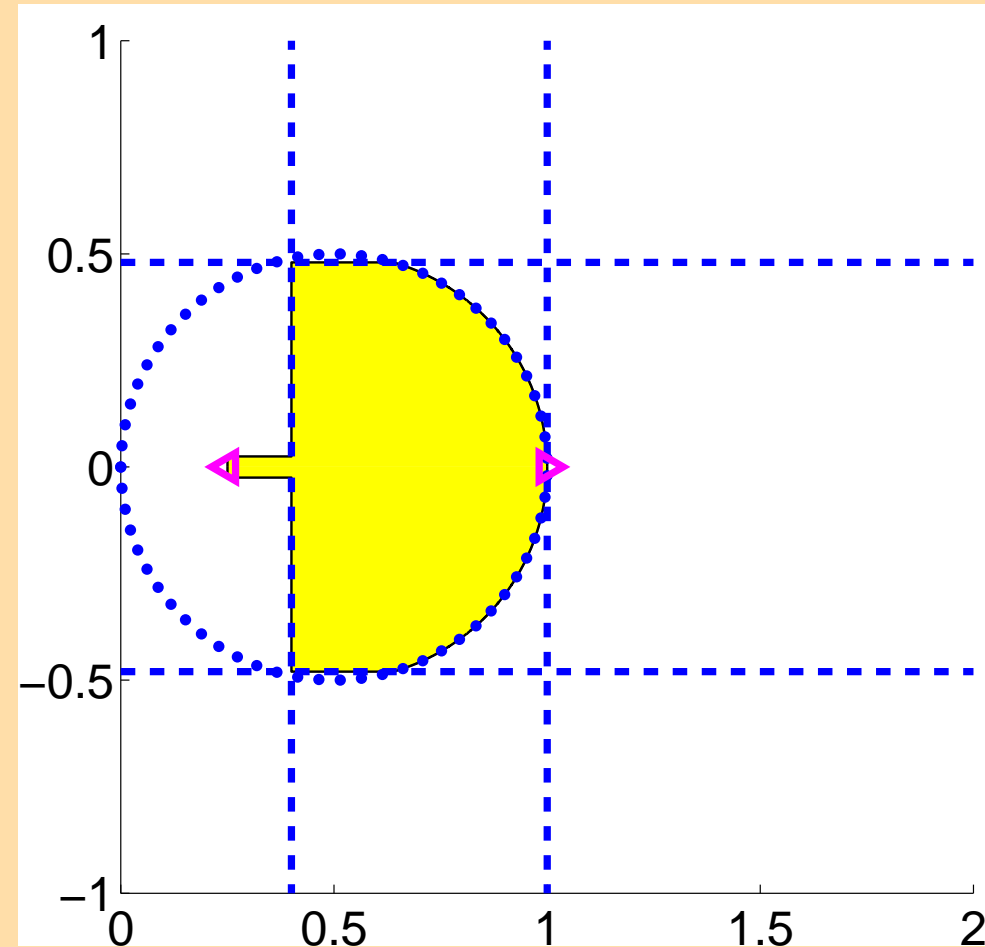
4. Eigenvalue analysis

The eigenvalues with nonzero imaginary part satisfy

$$\underline{\chi} \leq \Re(\lambda) \leq \bar{\chi}, \quad |\Im(\lambda)| \leq \delta, \quad |\lambda - \zeta| \leq \zeta$$



Block Approx. Fact.



Block SGS

$$\underline{\mu} = 0.4, \quad \underline{\nu} = 0.25, \quad \bar{\mu} = \bar{\nu} = 1$$

4. Eigenvalue analysis

Comparison with bounds appeared previously:

- Our estimates for real eigenvalues are sharper (see below)

4. Eigenvalue analysis

Comparison with bounds appeared previously:

- Our estimates for real eigenvalues are sharper (see below)
- The bound in [Simoncini (2004)] for Block Triang.:

$$\Im_m(\lambda) \neq 0 \quad \Rightarrow \quad |\lambda - 1| \leq \sqrt{1 - \underline{\mu}}$$

offers a useful complement.

4. Eigenvalue analysis

Comparison with bounds appeared previously:

- Our estimates for real eigenvalues are sharper (see below)
- The bound in [Simoncini (2004)] for Block Triang.:

$$\Im_m(\lambda) \neq 0 \quad \Rightarrow \quad |\lambda - 1| \leq \sqrt{1 - \underline{\mu}}$$

offers a useful complement.

- Ours analysis extends this result to Uzawa, and further to Block SGS, for which we get:

$$\Im_m(\lambda) \neq 0 \quad \Rightarrow \quad |\lambda - 1| \leq \rho_A$$

(via the equivalence between Block SGS and Block Triangular⁽²⁾)

4. Eigenvalue analysis

Combining this with our bounds for real eigenvalues:

Corollary

Let

$$\rho_A = \rho(I - M_A^{-1}A) , \quad \rho_S = \rho(I - M_S^{-1}S)$$

- For Block Triang. & Uzawa, if

$$\bar{\mu} = \lambda_{\max}(M_A^{-1}A) \leq 1$$

then:

$$\rho(I - M_t^{-1}K) = \rho(I - M_u^{-1}K) \leq \max(\sqrt{\rho_A}, \rho_S)$$

- For block SGS:

$$\rho(I - M_g^{-1}K) \leq \max(\rho_A, \rho_S)$$

5. Example

Stationnary Stokes problem on the unit square Ω

Find velocity vector \mathbf{v} and the kinematic pressure field p satisfying

$$\begin{aligned} -\Delta \mathbf{u} + \nabla p &= \mathbf{f} && \text{in } \Omega \\ \nabla \cdot \mathbf{u} &= 0 && \text{in } \Omega \end{aligned}$$

Dirichlet BC for \mathbf{v} & FD MAC scheme

→

$$K = \begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix}$$

Further, with $M_S = I$,

$$\underline{\nu} \approx 0.25 \quad \text{and} \quad \bar{\nu} = 1$$

Preconditioner for A

- A block diagonal with each diag. block being a discrete Laplace operator



well approximated by any (algebraic) multigrid method

Preconditioner for A

- A block diagonal with each diag. block being a discrete Laplace operator



well approximated by any (algebraic) multigrid method

- AGMG efficient and as simple as (in Matlab):

% Set up the preconditioner:

```
>> agmg(A, [], [], [], [], [], [], 1);
```

% Apply prec. to x, result in y:

```
>> y=agmg([], x, [], [], [], [], [], 3);
```

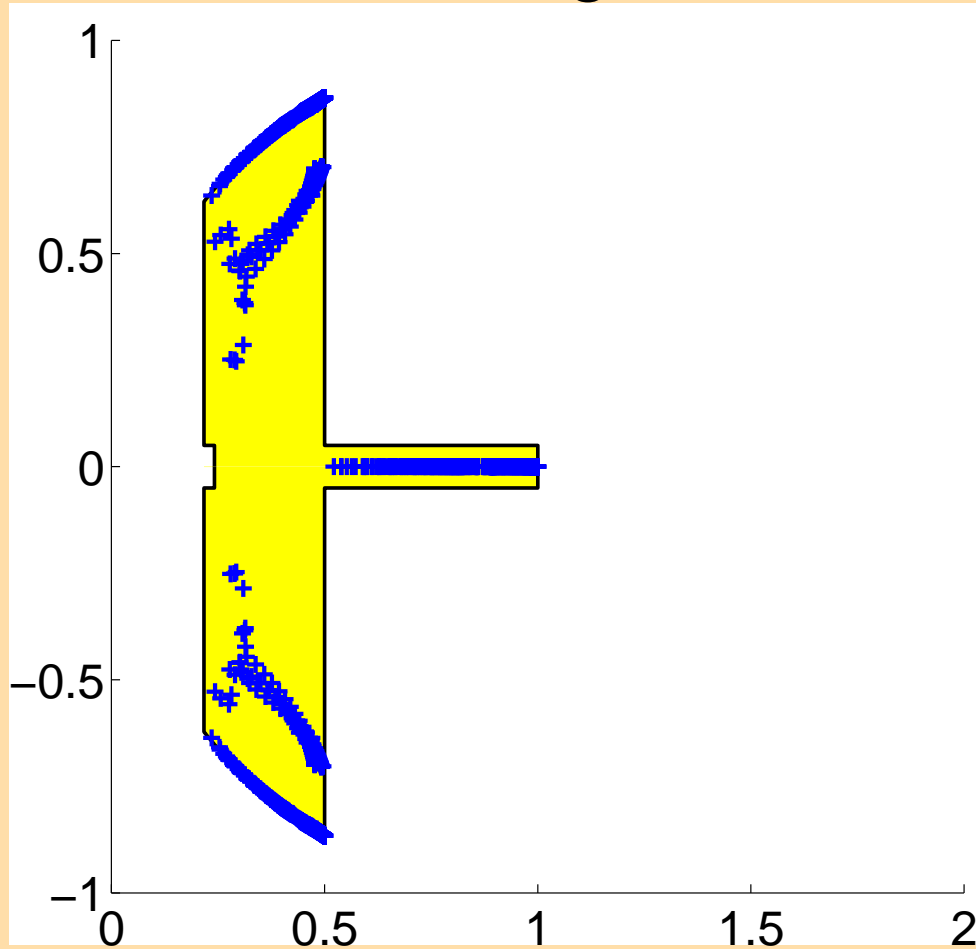
- Then:

$$\underline{\mu} \approx 0.4 \quad \text{and} \quad \bar{\mu} = 1$$

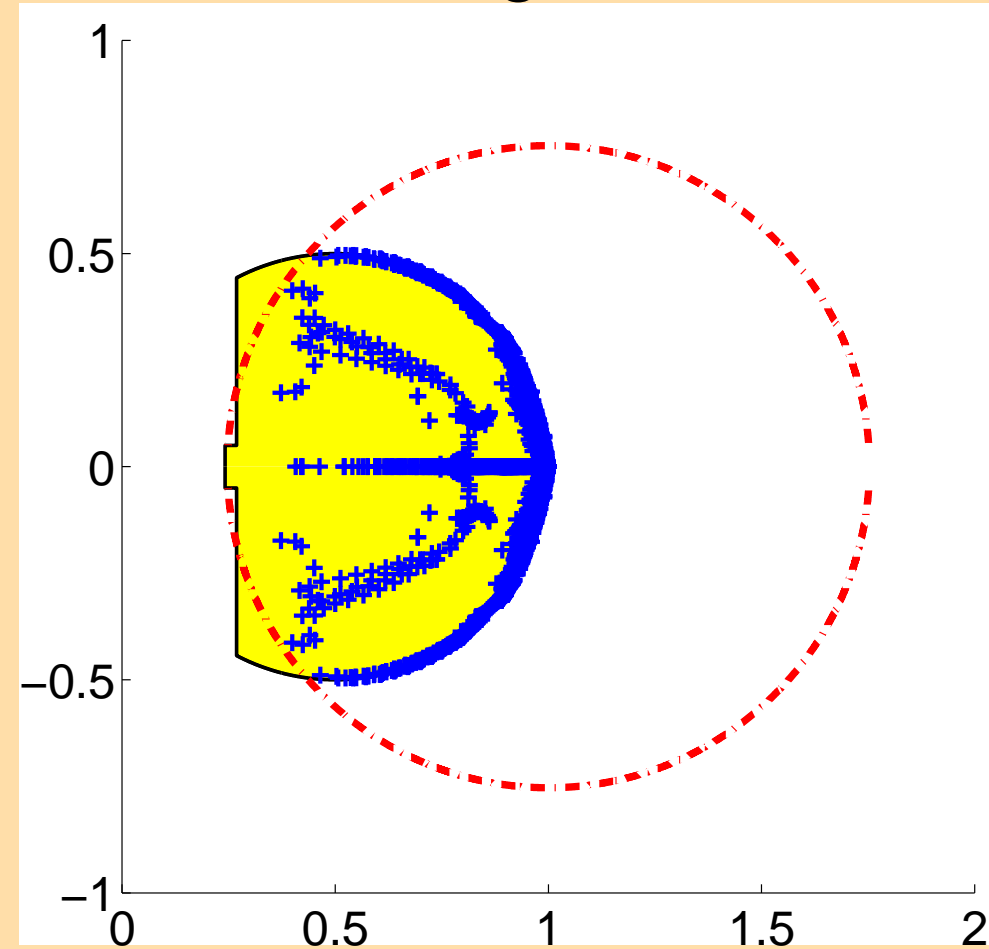
5. Example

Eigenvalue plot

Block Diagonal



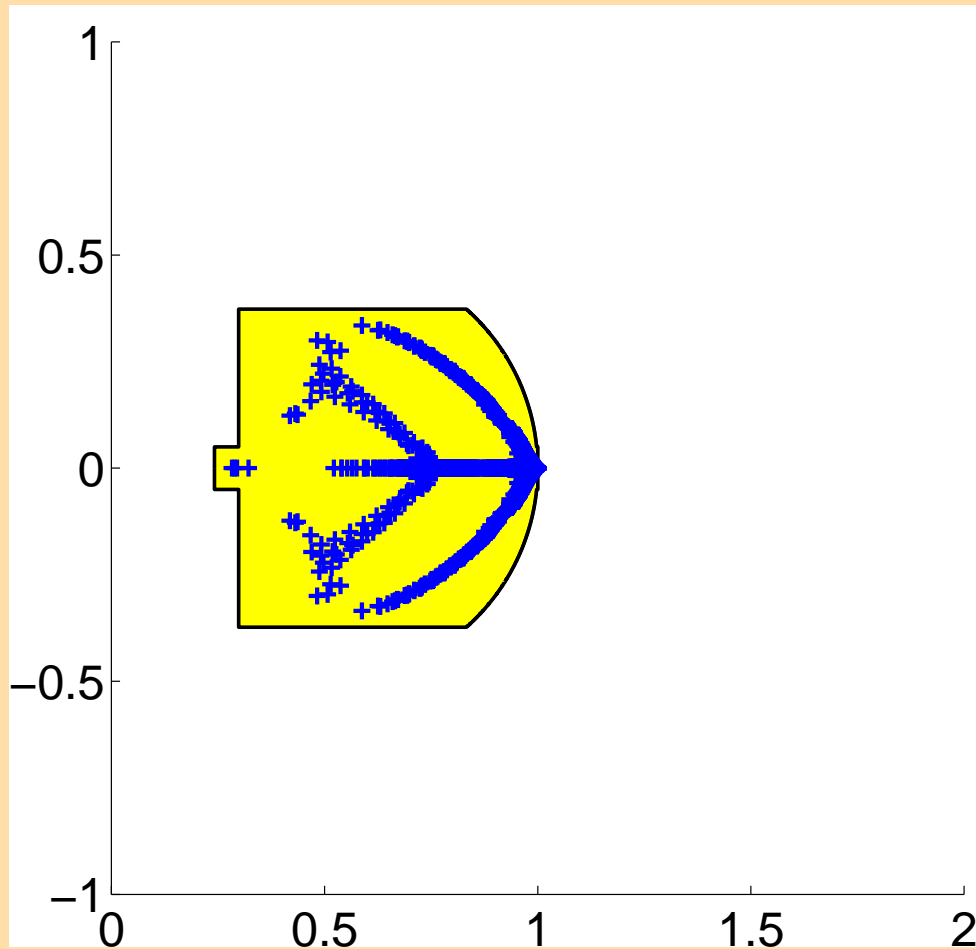
Block Triang. & Uzawa



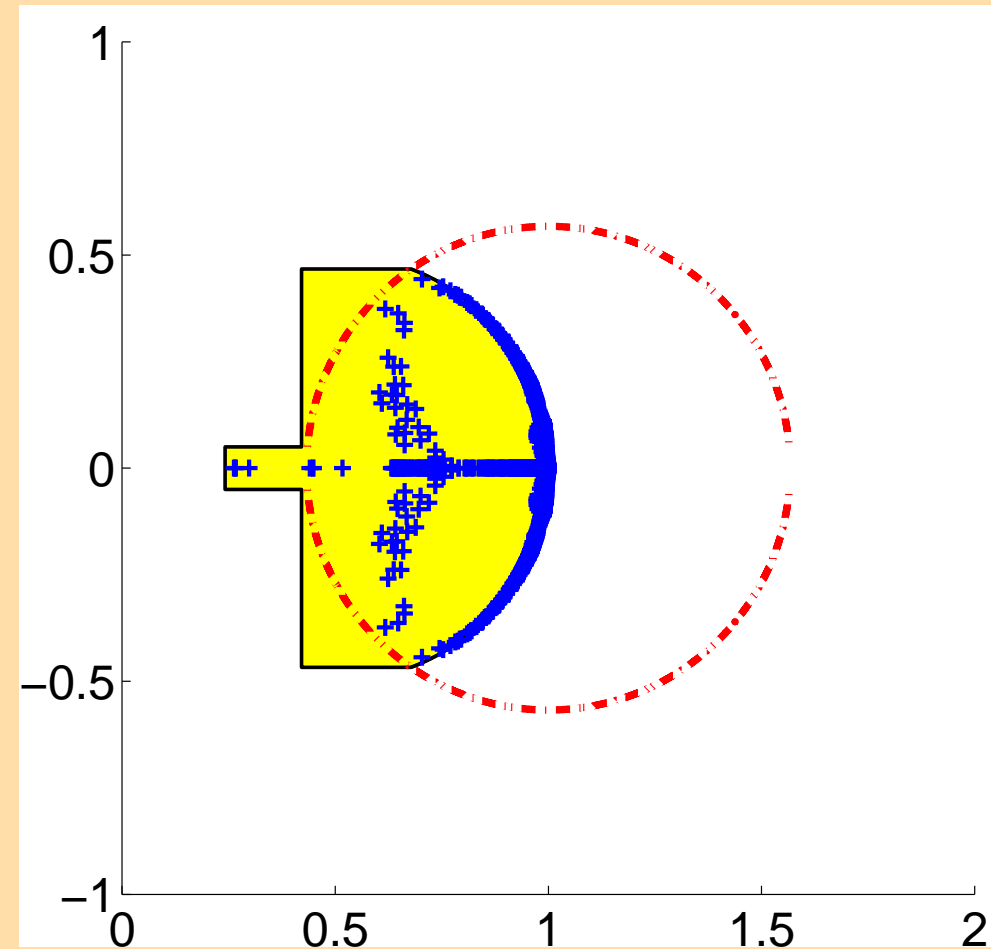
5. Example

Eigenvalue plot

Block Approx. Fact.



Block SGS



5. Example

Rescaling the preconditioner for A :

$$\underline{\mu} \approx 0.4\alpha \quad \text{and} \quad \bar{\mu} = \alpha$$

Makes sense for Block Triang. & Uzawa

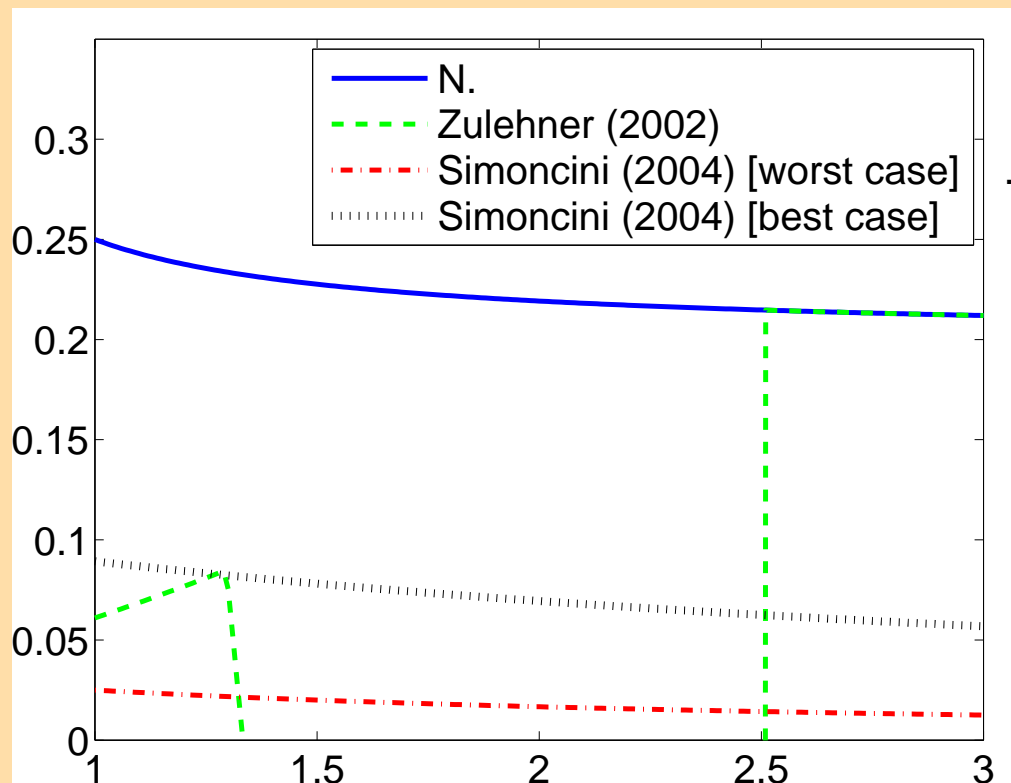
5. Example

Rescaling the preconditioner for A :

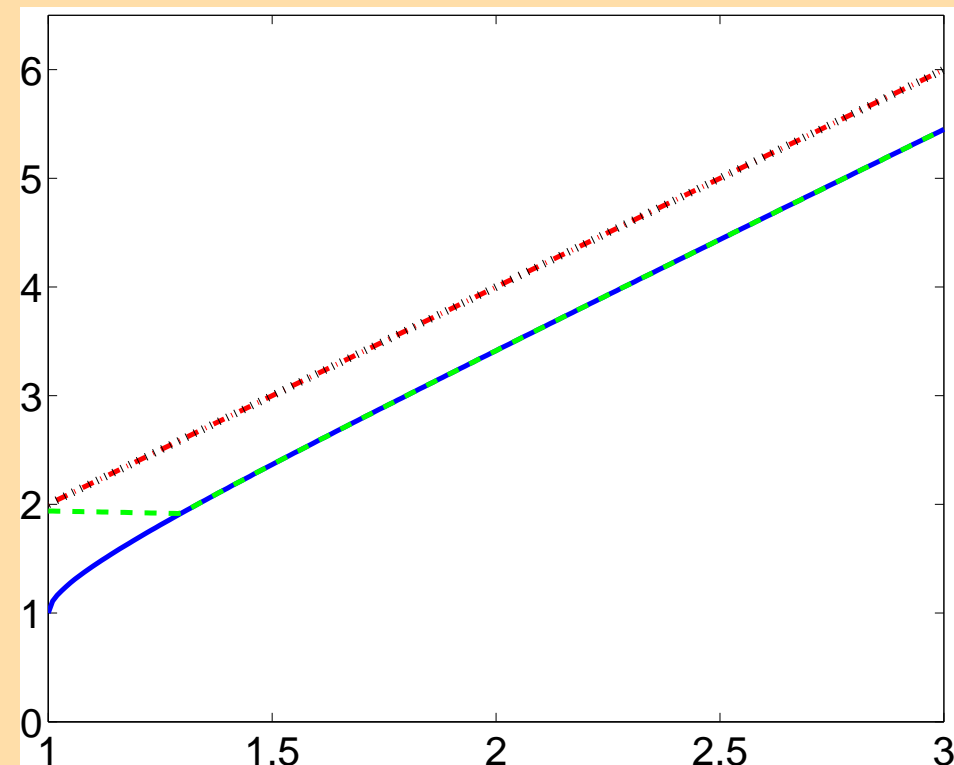
$$\underline{\mu} \approx 0.4\alpha \quad \text{and} \quad \bar{\mu} = \alpha$$

Makes sense for Block Triang. & Uzawa

Lower bounds = $f(\alpha)$

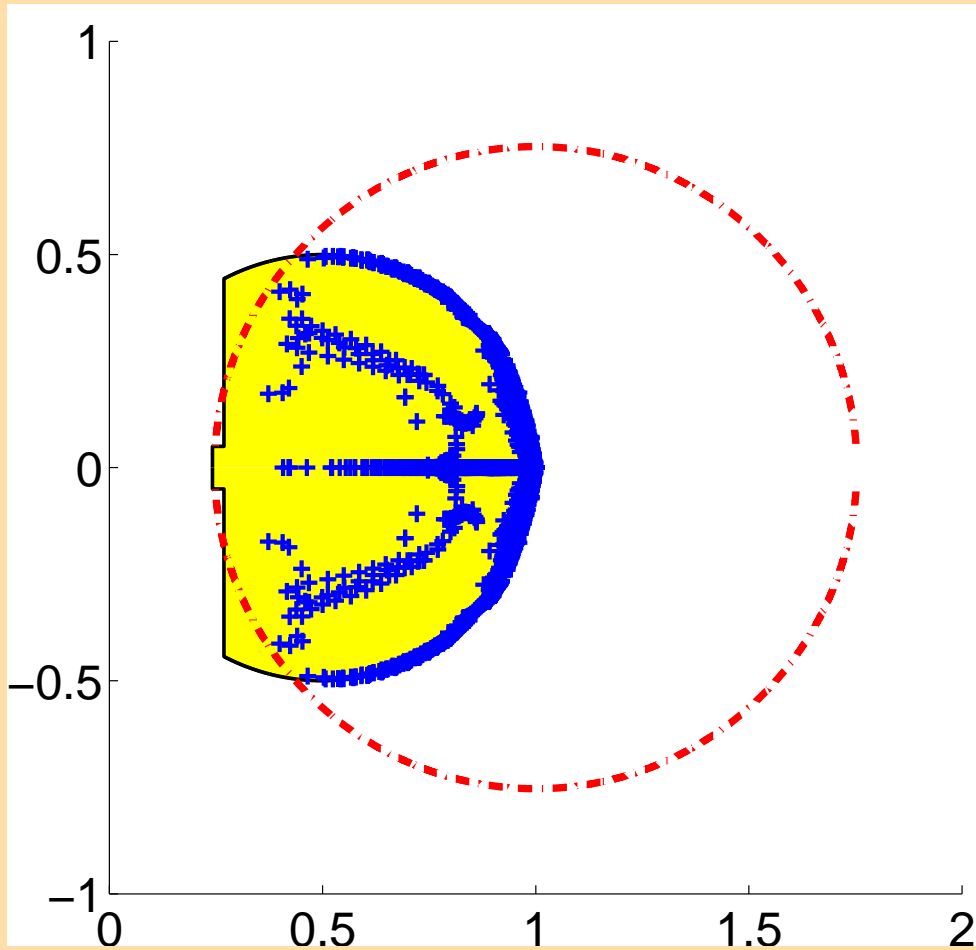


Upper bounds = $f(\alpha)$

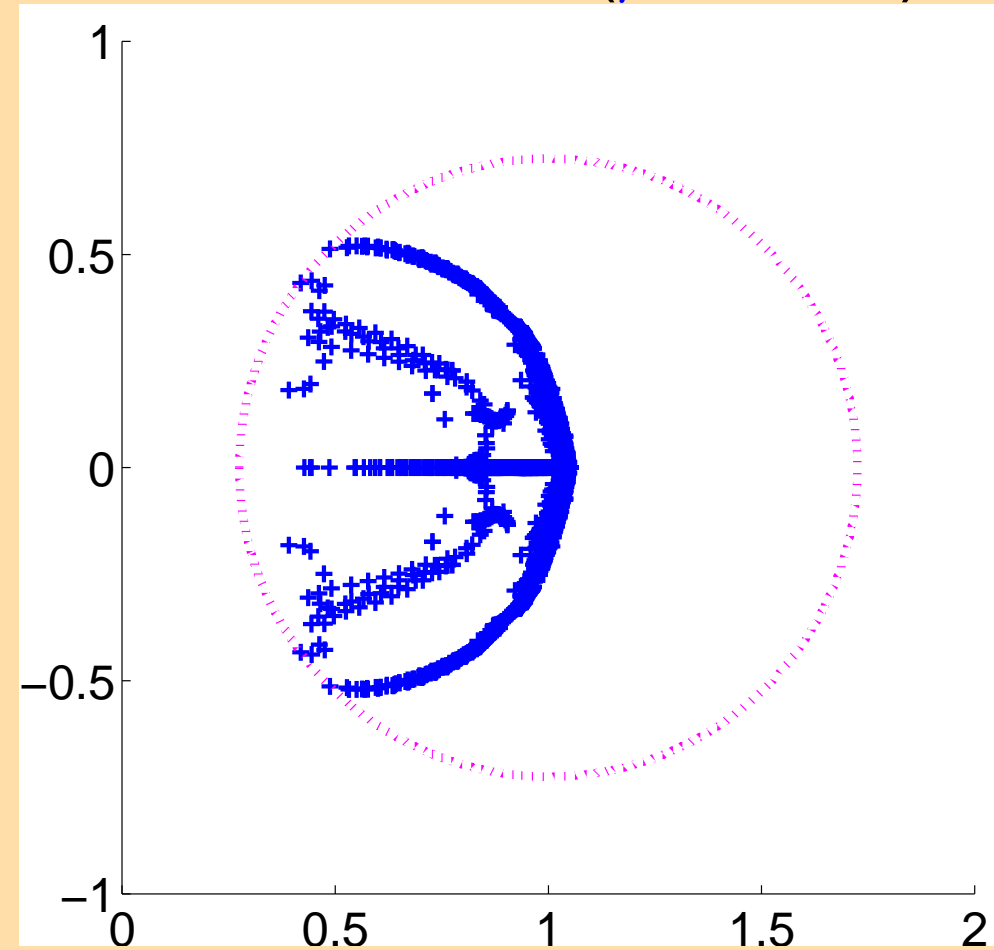


5. Example

$\alpha = 1$

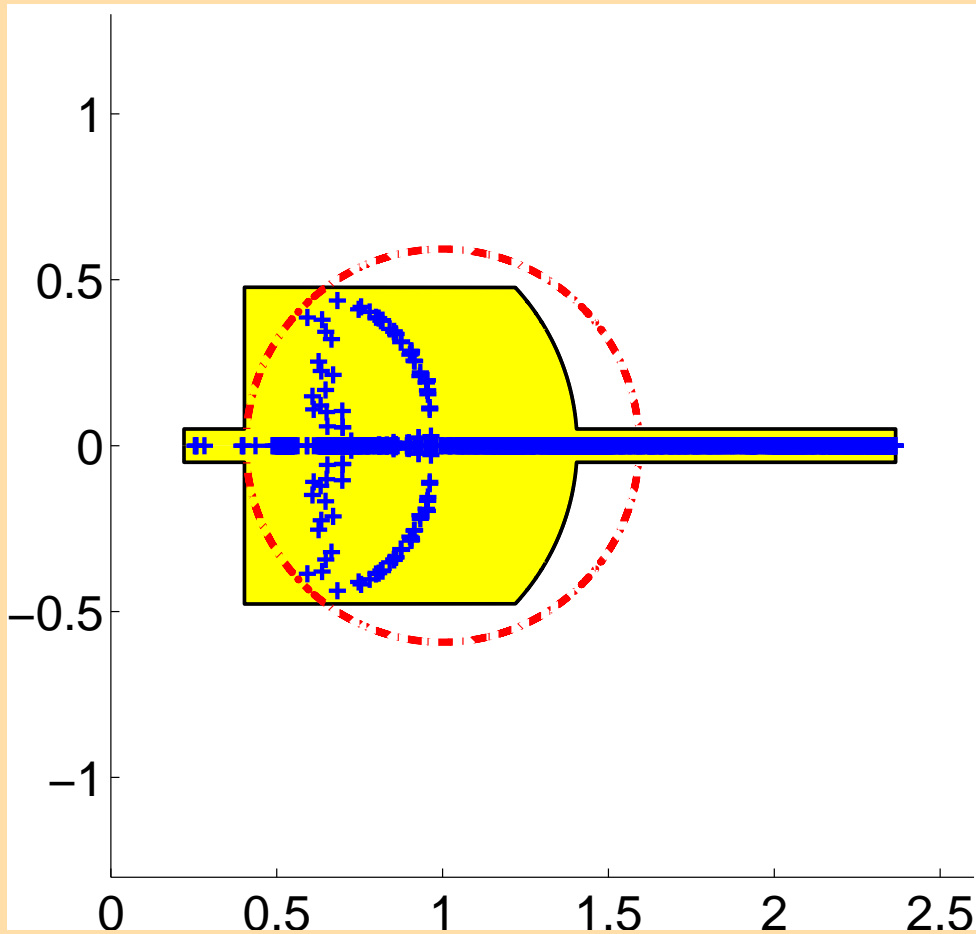


$\alpha = 1$, scaled ($\rho = 0.73$)

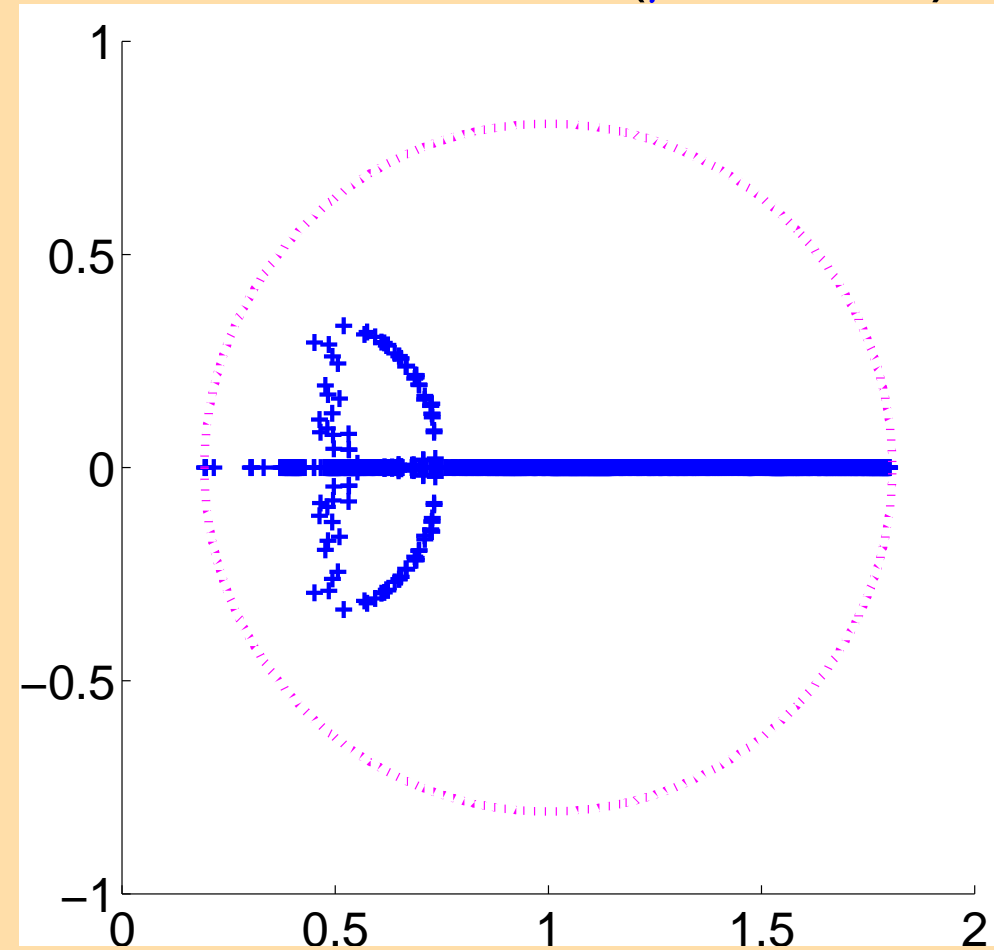


5. Example

$\alpha = 1.5$

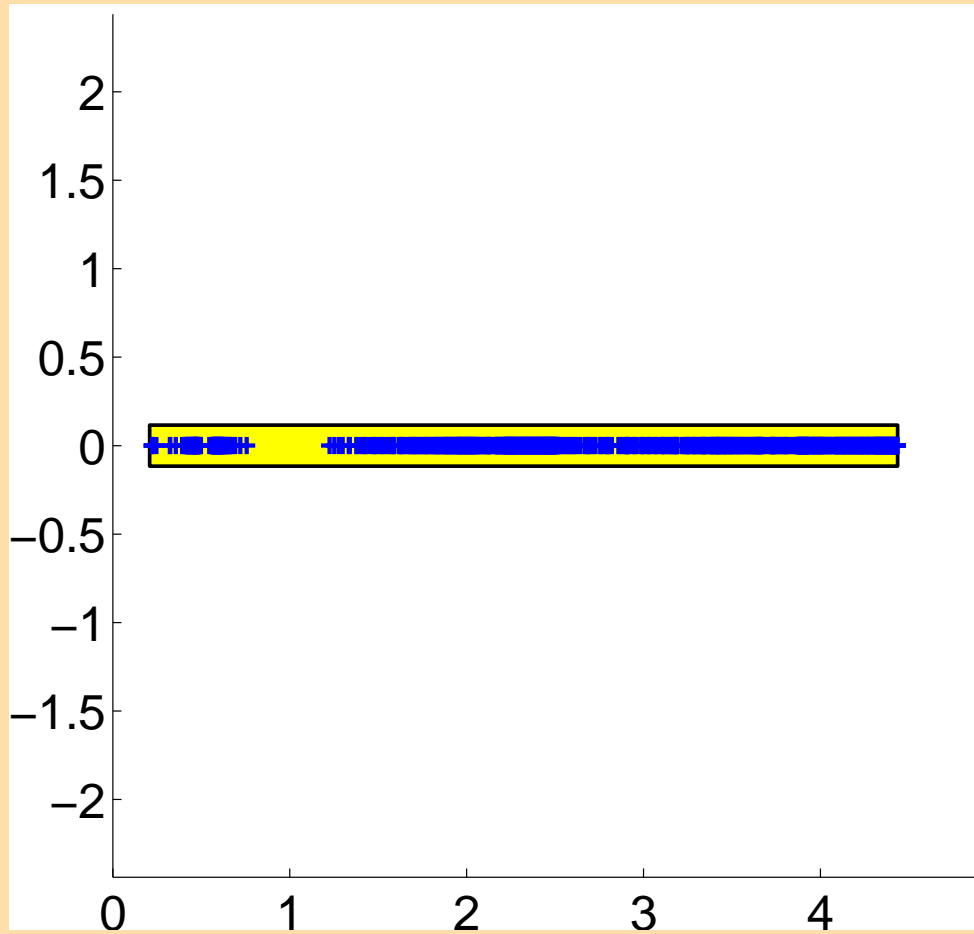


$\alpha = 1.5$, scaled ($\rho = 0.81$)

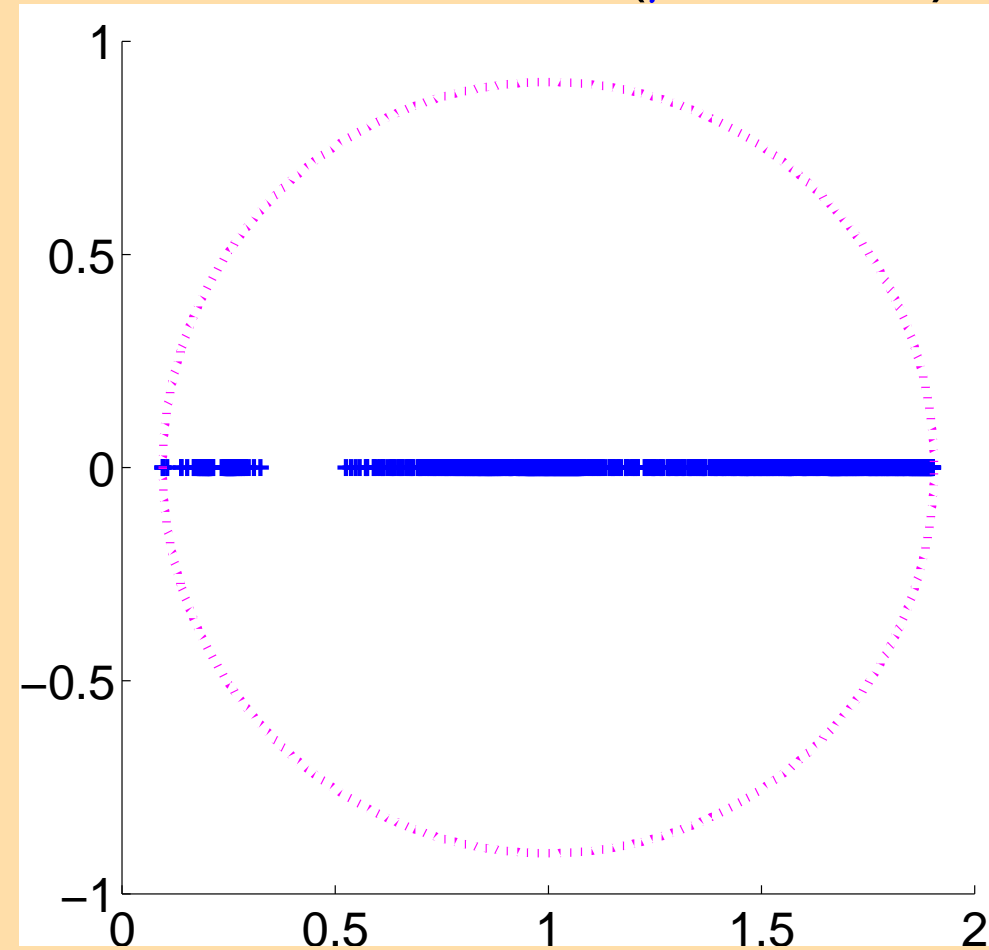


5. Example

$\alpha = 2.5$



$\alpha = 2.5$, scaled ($\rho = 0.90$)



5. Example

Number of iterations to solve the system ($\alpha = 1.5$)

h^{-1}	32	512	1024
Block Diag. ⁽⁺⁾ (M_+)	43	59	62
Block Diag. (M_d)	58	89	150
Block Triang. (M_t)	28	47	58
Inexact Uzawa (M_u)	30	61	57
Block Fact. (M_f)	21	29	37
Block SGS (M_g)	19	23	26
Block Triang. ⁽²⁾ (M_{t_2})	19	23	26
Inexact Uzawa ⁽²⁾ (M_{u_2})	20	26	28

Krylov:

MINRES for M_+

GCR(15) in all other cases

- AGMG becomes a variable preconditioner for $h^{-1} > 32$
- M_d performs similarly to M_+ if larger restart is chosen

5. Example

Number of iterations to solve the system ($\alpha = 1.5$)

Preconditioner for A	AGMG			amg(ifiss 3.2)
	h^{-1}	32	512	1024
Block Diag. ⁽⁺⁾ (M_+)	43	59	62	37
Block Diag. (M_d)	58	89	150	54
Block Triang. (M_t)	28	47	58	28
Inexact Uzawa (M_u)	30	61	57	28
Block Fact. (M_f)	21	29	37	25
Block SGS (M_g)	19	23	26	25
Block Triang. ⁽²⁾ (M_{t_2})	19	23	26	23
Inexact Uzawa ⁽²⁾ (M_{u_2})	20	26	28	25

5. Example

Time to solve the system ($\alpha = 1.5$)

Preconditioner for A	AGMG		amg(ifiss 3.2)
	h^{-1}	512	1024
Block Diag. ⁽⁺⁾ (M_+)	16.0	17.2	28.0
Block Diag. (M_d)	26.8	49.0	40.8
Block Triang. (M_t)	15.0	19.5	21.5
Inexact Uzawa (M_u)	19.7	19.5	21.7
Block Fact. (M_f)	15.3	20.6	35.6
Block SGS (M_g)	11.8	14.0	35.3
Block Triang. ⁽²⁾ (M_{t_2})	11.2	13.5	32.0
Inexact Uzawa ⁽²⁾ (M_{u_2})	12.9	15.2	34.9

Elapsed time excluding set up in
seconds per 10^6 unknowns

5. Example

Time to solve the system ($\alpha = 1.5$)

Preconditioner for A	AGMG		amg(ifiss 3.2)
	h^{-1}	512	1024
Block Diag. ⁽⁺⁾ (M_+)	16.4	17.7	5717.
Block Diag. (M_d)	27.2	49.5	5730.
Block Triang. (M_t)	15.4	19.9	5711.
Inexact Uzawa (M_u)	20.1	20.0	5711.
Block Fact. (M_f)	15.7	21.1	5725.
Block SGS (M_g)	12.2	14.4	5725.
Block Triang. ⁽²⁾ (M_{t_2})	11.6	13.9	5721.
Inexact Uzawa ⁽²⁾ (M_{u_2})	13.3	15.7	5724.

Total Elapsed time (including set up) in
seconds per 10^6 unknowns

- All block preconditioners work well except perhaps M_d

- All block preconditioners work well except perhaps M_d
- We are left with essentially three options:
 - positive definite block diagonal
 - block upper or lower triangular
 - block SGS

- All block preconditioners work well except perhaps M_d
- We are left with essentially three options:
 - positive definite block diagonal
 - block upper or lower triangular
 - block SGS
- What is the best scheme depends on the relative quality and cost of the preconditioners for A and S (Block SGS iterates twice on A)

- All block preconditioners work well except perhaps M_d
- We are left with essentially three options:
 - positive definite block diagonal
 - block upper or lower triangular
 - block SGS
- What is the best scheme depends on the relative quality and cost of the preconditioners for A and S (Block SGS iterates twice on A)
- There is in fact more to gain or loose in the choice of the preconditioner for the subproblems (M_A and M_S)

- All block preconditioners work well except perhaps M_d
- We are left with essentially three options:
 - positive definite block diagonal
 - block upper or lower triangular
 - block SGS
- What is the best scheme depends on the relative quality and cost of the preconditioners for A and S (Block SGS iterates twice on A)
- There is in fact more to gain or loose in the choice of the preconditioner for the subproblems (M_A and M_S)
- Be cautious with numerical results involving multigrid (especially: algebraic multigrid) and displaying only iteration counts

More details: reports available

Y. Notay, A new eigenvalue analysis of block preconditioners for saddle point problems

Francisco J. Gaspar, Y. Notay, Cornelis W. Oosterlee and Carmen Rodrigo, A simple and efficient segregated smoother for the discrete Stokes equations

More details: reports available

Y. Notay, A new eigenvalue analysis of block preconditioners for saddle point problems

Francisco J. Gaspar, Y. Notay, Cornelis W. Oosterlee and Carmen Rodrigo, A simple and efficient segregated smoother for the discrete Stokes equations

Thank you for your attention !