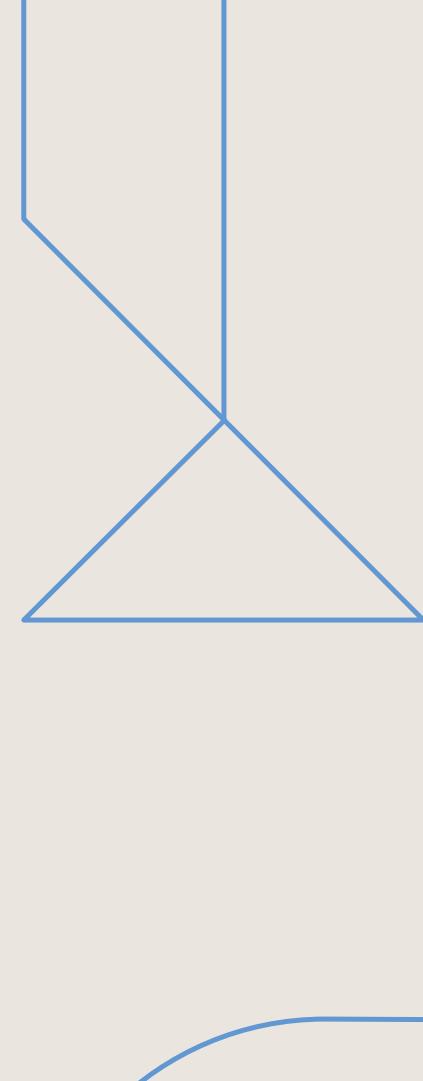




# Anderson Accelerated PMHSS for Complex-Symmetric Linear Systems

Måns I. Andersson, *Felix Liu, Stefano Markidis*  
March 7, 2024 — KTH Royal Institute of Technology



## Introduction

Linear Solvers

PMHSS Iteration

## Method

Anderson Acceleration

Test Cases

## Results

Rate of Convergence

Timings

## Conclusions and future work



### Complex-valued linear systems

$$(A + iB)x = b. \quad A, B \in \mathbb{R}^{n \times n} \text{ and } x, b \in \mathbb{C}^n \quad (1)$$

## Many PDEs

- Electromagnetics
- Waves
- etc.

There are many methods for solving linear systems

- Splitting methods: Jacobi, Gauss-Seidel, ...
- Projection methods
- Krylov subspace methods: CG, GMRES, ...
- Multigrid methods

## PMHSS-splitting

Preconditioned modified skew-Hermitian splitting (PMHSS), comes from a family of splitting: alternating direction iteration (ADI).

$$(\alpha V + A)\hat{x}_{k+1/2} = (\alpha V - iB)\hat{x}_k + b \quad (2)$$

$$(\alpha V + B)\hat{x}_{k+1} = (\alpha V + iA)\hat{x}_{k+1/2} - ib. \quad (3)$$

For  $V \in \mathbb{R}^{n \times n}$  and  $\alpha > 0$ , as  $k$  increase,  $\hat{x}$  tends to the solution of (1). For the choice  $\alpha = 1$  and  $V = A$  we have:

$$(A + B)\hat{x}_{k+1/2} = \frac{(1+i)}{2}(A - iB)x_k - \frac{(1-i)}{2}b \quad (4)$$

### the take away

This iteration is conditionally converging to  $x$  independently of  $\hat{x}_0$ . If the iteration matrix  $\Psi = (A + B)^{-1}\frac{(1+i)}{2}(A - iB)$  has a spectrum covered by the unit disk  $\rho(\Psi) < 1$ .

## The fixed-point iteration

$$\Psi = \frac{(1+i)}{2}(A + B)^{-1}(A - iB), \quad c = \frac{(1-i)}{2}(A + B)^{-1}b. \quad (5)$$

Assuming an initial guess  $\hat{x}_0 = 0$  (and thus  $\hat{x}_1 = c$ ), repeated application of the update rule gives us the following relation:

$$\hat{x}_{k+1} = \Psi(\dots(\Psi c + c) + \dots) + c = (\Psi^k + \dots + \Psi + I)c. \quad (6)$$

The expression in the parentheses is a geometric sum (of matrices), which has the closed form  $(I - \Psi)^{-1}(I - \Psi^k)$ . Hence, we have

$$\hat{x}_{k+1} = (I - \Psi)^{-1}(I - \Psi^k)c. \quad (7)$$

When the spectral radius  $\rho(\Psi) < 1$  as in, we have that  $(I - \Psi)^{-1}(I - \Psi^k) \rightarrow (I - \Psi)^{-1}$  as  $k \rightarrow \infty$ . Thus, in the limit, the solution obtained satisfies:

$$(I - \Psi)\hat{x} = c. \quad (8)$$

## Solving the inner system in-exactly

The following linear system needs to be solved each iteration.

$$(A + B)\hat{x}_k = \frac{(1+i)}{2}(A - iB)x_k - \frac{(1-i)}{2}b \quad (9)$$

Evaluating the error

$$||e_0(k)|| = ||x_{k+1}^* - \hat{x}_k|| \quad (10)$$

$$\approx ||\hat{x}_{k+1} - \hat{x}_k|| \quad (11)$$

$$= ||(I - \Psi)^{-1}(\Psi^k - \Psi^{k+1})c|| \quad (12)$$

$$= ||(I - \Psi)^{-1}(I - \Psi)\Psi^k c|| \quad (13)$$

$$\leq ||\Psi^k|| \cdot ||c||. \quad (14)$$

## The take away

Each outer iteration generates a solution  $x$  that is a better approximation for the inner solver.

## As a preconditioner

The *iteration matrix* from the splitting has been used with Krylov solvers such as GMRES.

PMHSS preconditioning, where  $q$  is the current residual in the iterative method (GMRES in our case).

1: Solve  $(A + B)z = q$

2: Set  $x = \frac{(1-i)}{2}z$

## Applying the preconditioner

Note that the linear system that is solved as the preconditioner is not the same as the one in the fixed-point.

## AA-PMHSS

Anderson Acceleration is a method that is used to accelerate fixed-point iterations by recasting it as an optimization problem.

$$x = f(x), \quad g(x) := f(x) - x \quad (15)$$

With the minimization defined as:

$$\operatorname{argmin}_{\alpha_k \in A_k} \|G_k \alpha\|_2 \quad A_k = \{\alpha \in \mathbb{C}^k : \sum_i \alpha_i = 1\}, \quad (16)$$

where  $G_k = [g_0, \dots, g_k]$ . Then the current best solution is updated accordingly.

$$x_{k+1} = F_k \alpha_k \quad F_k = [f_0, \dots, f_k] \quad (17)$$

A single step in our algorithm

- 1: Solve :  $(A + B)x_{k+1} = \frac{(1+i)}{2}(A - iB)\bar{x}_k + \frac{(1-i)}{2}b$
- 2:  $X_k = [x_1 - x_0, \dots, x_{k-1} - x_k]$   
 $G_k = [g_1 - g_0, \dots, g_{k-1} - g_k]$   
 $g_k = x_{k+1} - \bar{x}_k$
- 3: Solve :  $\operatorname{argmin}_{\alpha_k \in A_k} \|g_k - G_k \alpha\|_2, A_k = \{\alpha \in \mathbb{C}^k\}$
- 4:  $\bar{x}_{k+1} = x_k + g_k - (X_k + G_k)\alpha_k$

## Anderson Acceleration and GMRES

fixed-point

$$x \in F(x_0, f(x_0), f(f(x_0)), f(f(f(x_0))), \dots)$$

AA

$$\operatorname{argmin}_{x \in F(x_0, f)} \|f(x) - x\|$$

Krylov

$$x \in K(r_0, Ar_0, A^2r_0, A^3r_0, \dots) \quad (18)$$

GMRES

$$\operatorname{argmin}_{x \in K(r_0, A)} \|r(x)\|$$

Essentially equivalent

Accelerating the fixed-point iteration defined by  $f(x) = Ax + b$  with AA is equivalent to solving  
 $(I - A)x = b$

## Spectrum

Assume  $A$  is SPD and  $B$  is SPSD so that the eigenvalues  $\mu$  of  $A^{-1}B$  are real and non-negative. we have that  $(A + B)^{-1}(A + iB) = (I + (i - 1)(I + A^{-1}B)^{-1}(A^{-1}B))$  thus the preconditioned eigenvalues are:

$$\lambda_{\text{PMHSS-GMRES}} = 1 + (i - 1) \frac{\mu}{\mu + 1}. \quad (19)$$

We then have, if  $\mu_{\min} \leq \mu \leq \mu_{\max}$ ,

$$\frac{1}{1 + \mu_{\max}} \leq \Re(\lambda) \leq \frac{1}{1 + \mu_{\min}}, \quad (20)$$

$$\frac{\mu_{\min}}{1 + \mu_{\min}} \leq \Im(\lambda) \leq \frac{\mu_{\max}}{1 + \mu_{\max}}. \quad (21)$$

Modelling the AA-PMHSS as GMRES with  $(I - \Psi)$  and similar analysis as before, gives us the preconditioned eigenvalues for the AA-PMHSS iteration as

$$\lambda_{\text{AA-PMHSS}} = 1 - \bar{\lambda}_{\text{PMHSS-GMRES}} \quad (22)$$

$$= 1 + (i + 1) \frac{\mu}{\mu + 1}. \quad (23)$$

## Conclusions

- This is the same spectrum but mirrored around the real axis.
- The spectrum is discretization independent

## Test Systems

### Padé approximation

$$\left[ \left( L + \frac{3 - \sqrt{3}}{h} I \right) + i \left( L + \frac{3 + \sqrt{3}}{h} I \right) \right] x = b \quad (24)$$

### Inhomogeneous Helmholtz equation or Shifted $\omega$ -system

$$\left[ L + (\mu + i\omega)I \right] x = b. \quad (25)$$

Here we use the coefficients  $\mu = 0$  and  $\omega = 0.01$ .

### Equations of Motion

$$\left[ (K - \omega^2 M) + i(\omega C_V + C_H) \right] x = b. \quad (26)$$

Where  $M$  and  $K$  are the inertia- and the stiffness matrices,  $C_V$  and  $C_H$  are the viscous- and hysteretic dampening matrices.

### Time-harmonic Eddy Current simulation

High-order FEM system in three sizes.

With a conditioning parameter  $\omega$

$$\begin{bmatrix} \underline{A} & \underline{B}^T \\ \underline{B} & \underline{C} \end{bmatrix} \mathbf{x} = \mathbf{b} \quad (27)$$

Where  $\underline{B}$  is a complex matrix,  $\underline{A}$  and  $\underline{C}$  are complex-symmetric matrices.

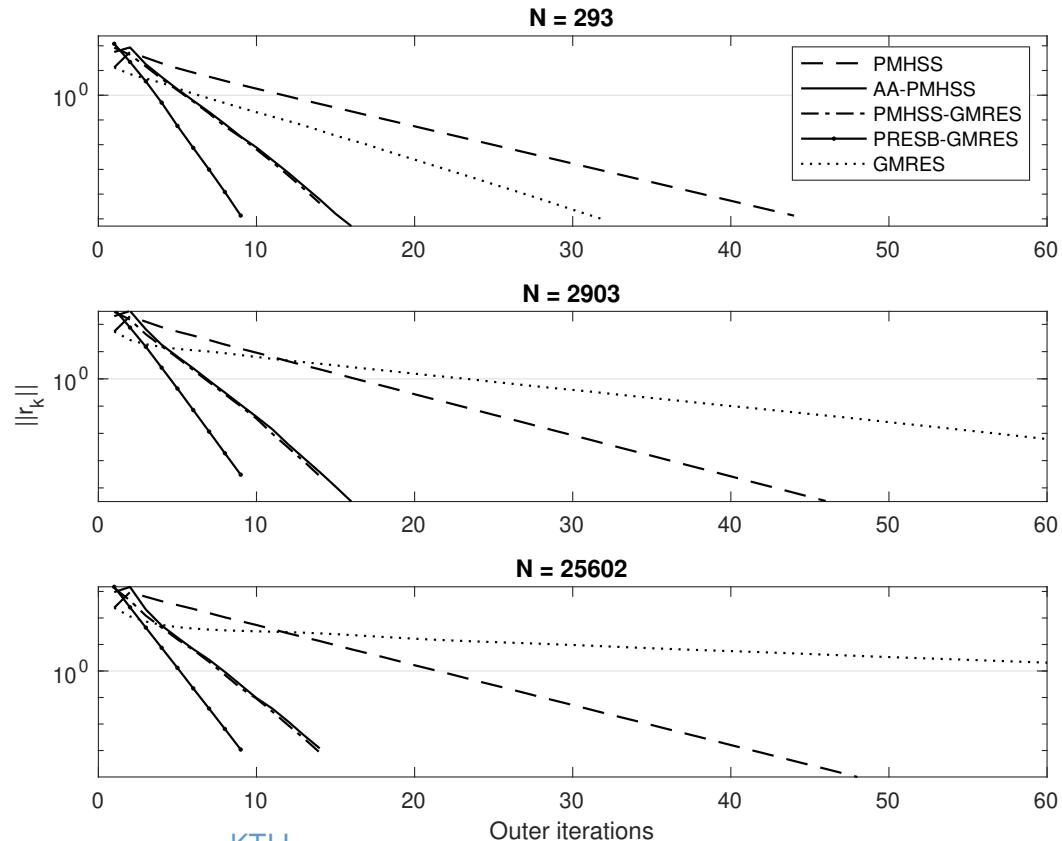
## Results

### Rate of Convergence

- Unpreconditioned GMRES only converge for the smallest system.
- PMHSS-GMRES and AA-PMHSS converges at essentially the same rate.

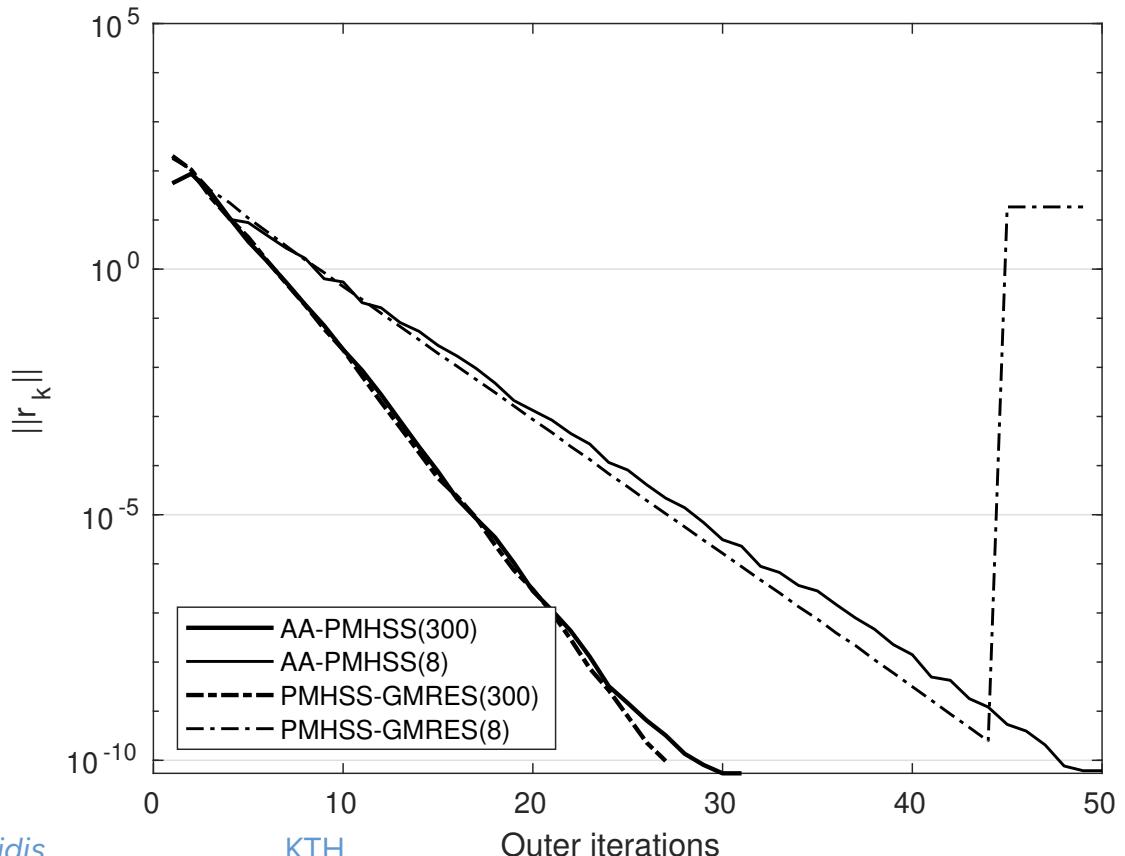
### PRESB preconditioner

The PRESB preconditioner needs to solve **two** systems with  $(A + B)$  for each GMRES iteration.



## Outer iterations

- Number of outer iterations
- PMHSS-GMRES is more prone to break down due to accuracy of inner solver

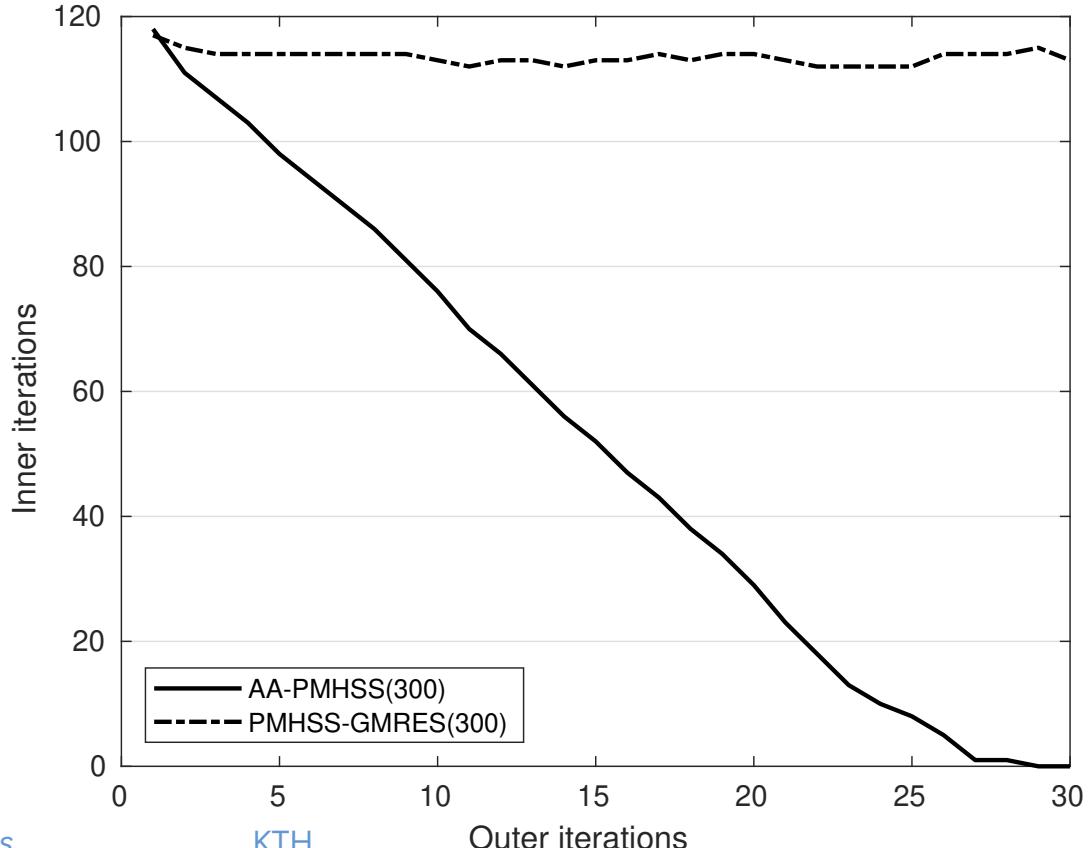


## Inner iterations

- Inner/Outer
- 300 iterations was pessimistic
- We see a linear decay in the number of needed inner iterations for AA-PMHSS.

## Fewer CG calls

A 1/3 as many CG iterations in total



## Results

Method/size	outer iter.	inner iter.	$T \pm \sigma$ [s]	$\ r\ $	$\ x^* - x\ $	$\ (A + B)^{-1}r\ $
<b>Padé</b>						
PMHSS						
10000	33	4172	$0.677228 \pm 5.79\%$	3.44893e-16	6.24067e-15	4.99637e-15
40000	34	6134	$2.63969 \pm 0.8817\%$	1.71351e-17	6.20946e-16	4.99868e-16
90000	34	7531	$6.13483 \pm 0.3547\%$	4.10001e-18	2.16516e-16	1.74593e-16
GMRES						
10000	133	-	$0.638218 \pm 1.767\%$	7.715e-15	6.87689e-14	-
40000	188	-	$3.4308 \pm 1.579\%$	9.54629e-16	2.35029e-14	-
90000	227	-	$9.44645 \pm 2.2\%$	2.9451e-16	1.0854e-14	-
AA-PMHSS						
10000	10	1300	<b><math>0.238979 \pm 5.973\%</math></b>	1.11205e-15	6.81956e-16	4.95078e-16
40000	11	1963	<b><math>1.00926 \pm 1.607\%</math></b>	3.94656e-17	3.33882e-17	2.47074e-17
90000	11	2432	<b><math>2.37411 \pm 0.4164\%</math></b>	1.31937e-17	1.36944e-17	1.01987e-17
PMHSS-GMRES						
10000	9	2204	$0.345996 \pm 0.02171\%$	1.96557e-14	1.37958e-14	1.00285e-14
40000	10	3203	$1.36888 \pm 0.009568\%$	5.67274e-16	5.10832e-16	3.77199e-16
90000	10	3995	$3.27446 \pm 0.01784\%$	1.92527e-16	1.97563e-16	1.46642e-16
PRESB-GMRES						
10000	8	3464	$0.258084 \pm 0.008148\%$	3.35243e-15	4.61427e-15	2.46771e-15
40000	8	5637	$1.28754 \pm 0.02749\%$	1.1019e-15	2.12654e-15	1.18669e-15
90000	8	7030	$3.05243 \pm 0.02684\%$	3.76795e-16	8.44006e-16	4.7808e-16

## Results

Method/size	outer iter.	inner iter.	$T \pm \sigma$ [s]	$\ r\ $	$\ x^* - x\ $	$\ (A + B)^{-1}r\ $
<b>Shifted-<math>w</math> system</b>						
PMHSS						
10000	49	6281	$1.085 \pm 0.2419\%$	2.58145e-10	7.72275e-11	7.63703e-11
40000	50	7126	$3.157 \pm 0.2576\%$	9.04324e-11	2.94851e-11	2.90853e-11
90000	50	7187	$6.058 \pm 0.6971\%$	6.02718e-11	1.97633e-11	1.9518e-11
GMRES						
10000	212	-	$1.677 \pm 3.722\%$	7.91011e-11	2.32285e-09	-
40000	232	-	$5.214 \pm 0.508\%$	3.87483e-11	1.35325e-09	-
90000	233	-	$10.05 \pm 1.007\%$	2.6577e-11	9.42004e-10	-
AA-PMHSS						
10000	18	2751	$0.5119 \pm 5.645\%$	1.61867e-11	1.2482e-11	1.03118e-11
40000	21	3498	$1.783 \pm 1.185\%$	2.50785e-11	2.48957e-11	2.05558e-11
90000	22	3648	$3.639 \pm 0.7017\%$	1.28299e-11	1.19634e-11	1.00378e-11
PMHSS-GMRES						
10000	18	4340	$0.7649 \pm 0.004377\%$	1.06948e-10	1.1078e-10	8.60833e-11
40000	22	5572	$2.461 \pm 0.03121\%$	4.01809e-11	4.64125e-11	3.75068e-11
90000	22	5906	$5.013 \pm 0.03158\%$	4.72434e-11	5.3194e-11	4.32722e-11
PRESB-GMRES						
10000	12	5616	<b><math>0.4379 \pm 0.009439\%</math></b>	5.23277e-11	7.00731e-11	4.42568e-11
40000	12	6793	<b><math>1.561 \pm 0.02302\%</math></b>	5.95516e-11	7.75276e-11	4.89482e-11
90000	12	6544	<b><math>2.866 \pm 0.03836\%</math></b>	6.26606e-12	7.18707e-11	4.42628e-11
AA-PMHSS(50)						
10000	21	1049	<b><math>0.2244 \pm 3.837\%</math></b>	4.67345e-12	4.88027e-11	4.2427e-11
40000	25	1248	<b><math>0.8066 \pm 1.546\%</math></b>	1.16763e-12	1.0343e-11	8.89348e-12
90000	26	1299	<b><math>1.78 \pm 1.487\%</math></b>	1.12012e-12	7.91642e-12	6.65226e-12

## Results

Method/size	outer iter.	inner iter.	$T \pm \sigma$ [s]	$\ r\ $	$\ x^* - x\ $	$\ (A + B)^{-1}r\ $
<b>Eq. of Motion</b>						
PMHSS						
10000	49	10358	$1.786 \pm 1.65\%$	1.25823e-10	7.30042e-11	7.09021e-11
40000	51	20638	$8.945 \pm 1.766\%$	3.13699e-11	3.25443e-11	3.16235e-11
90000	52	30185	$24.9 \pm 0.1285\%$	1.45818e-11	2.19632e-11	2.13458e-11
GMRES						
10000	280	-	$3.057 \pm 0.8814\%$	7.83319e-11	2.34518e-09	-
40000	548	-	$28.78 \pm 1.626\%$	3.97398e-11	4.39533e-09	-
90000	807	-	$117.2 \pm 1.591\%$	2.63963e-11	6.403e-09	-
AA-PMHSS						
10000	12	2732	<b>0.5376</b> $\pm 3.711\%$	4.22335e-13	6.91976e-13	5.76201e-13
40000	12	5369	<b>2.679</b> $\pm 1.429\%$	5.5309e-12	4.84056e-12	4.36451e-12
90000	12	7855	<b>7.421</b> $\pm 0.6799\%$	5.67128e-12	4.05111e-12	3.65619e-12
PMHSS-GMRES						
10000	11	4096	$0.7076 \pm 0.002372\%$	2.31419e-10	2.14003e-10	1.91861e-10
40000	11	8092	$3.542 \pm 0.007765\%$	1.2527e-10	1.82148e-10	1.63178e-10
90000	11	12055	$9.983 \pm 0.05302\%$	7.93328e-11	1.53629e-10	1.38611e-10
PRESB-GMRES						
10000	11	8210	$0.6628 \pm 0.05686\%$	1.59327e-10	2.3776e-10	1.60705e-10
40000	11	16261	$3.716 \pm 0.04519\%$	1.33754e-10	2.88983e-10	2.06747e-10
90000	11	24320	$10.34 \pm 0.0956\%$	4.37123e-11	1.14813e-10	8.5615e-11

## Results

Method $\omega = 10^4$	outer iter.	inner iter.	$T \pm \sigma$ [s]	$\ r\ $	$\ x^* - x\ $	$\ (A + B)^{-1}r\ $
<hr/>						
PMHSS						
LU	67	-	$25.43 \pm 0.592\%$	$2.73197e-09$	$1.03058e-09$	$1.02994e-09$
CG	67	2491	$1.162 \pm 2.973\%$	$2.73819e-09$	$1.03532e-09$	$1.03377e-09$
<hr/>						
AA-PMHSS						
LU	22	-	$8.733 \pm 0.4441\%$	$1.71475e-09$	$6.70805e-10$	$5.38084e-10$
CG	22	824	$0.5365 \pm 0.447\%$	$1.93437e-09$	$7.51107e-10$	$6.1096e-10$
CG(50)	23	765	$0.514 \pm 0.3678\%$	$9.89885e-10$	$3.953e-10$	$3.25222e-10$
CG(25)	23	496	$0.3881 \pm 0.6916\%$	$1.66947e-09$	$6.44053e-10$	$5.23233e-10$
<hr/>						
PMHSS-GMRES						
LU	23	-	$9.009 \pm 0.03202\%$	$1.00433e-09$	$3.90251e-10$	$2.1615e-10$
CG	23	1679	$0.8161 \pm 0.0251\%$	$1.01339e-09$	$3.9347e-10$	$3.05682e-10$
<hr/>						
PRESB-GMRES						
LU	14	-	$10.95 \pm 0.06775\%$	$1.88122e-09$	$9.2189e-10$	$5.98921e-10$
CG	14	2262	$0.6489 \pm 0.01114\%$	$1.8851e-09$	$9.22169e-10$	$5.99389e-10$
<hr/>						
$\omega = 10^{-4}$						
PMHSS						
LU	80	-	$30.25 \pm 0.6644\%$	$5.62865e-11$	$1.13009e-09$	$1.12844e-09$
CG	78	12506	$5.491 \pm 1.285\%$	$2.06198e-10$	$3.7422e-09$	$3.79037e-09$
<hr/>						
AA-PMHSS						
LU	25	-	$9.861 \pm 0.1814\%$	$3.58754e-10$	$6.31284e-10$	$5.02025e-10$
CG	26	4061	$2.099 \pm 0.5552\%$	$5.69164e-10$	$4.03492e-09$	$3.88251e-09$
CG(50)	32	1489	$0.9801 \pm 1.043\%$	$5.93055e-10$	$4.06493e-09$	$3.82198e-09$
CG(25)	43	1012	$0.9603 \pm 1.684\%$	$2.29485e-10$	$5.43464e-09$	$5.05516e-09$
<hr/>						
PMHSS-GMRES						
LU	23	-	$9.572 \pm 0.08122\%$	$4.73915e-09$	$1.18588e-08$	$6.48492e-09$
CG	23	7500	$3.33 \pm 0.03415\%$	$5.09979e-09$	$1.35183e-08$	$1.11445e-08$
<hr/>						
PRESB-GMRES						
LU	15	-	$11.67 \pm 0.07115\%$	$3.23933e-09$	$6.54481e-09$	$4.27435e-09$
CG	15	10184	$2.686 \pm 0.02967\%$	$4.00906e-09$	$8.49476e-09$	$7.52229e-09$

## Conclusions

- AA-PMHSS converges in a mesh-independently with respect to the number of outer iterations under assumptions on  $A$  and  $B$ . As PMHSS, PMHSS-GMRES, and PRESB-GMRES.
- AA-PMHSS' convergence behavior can be explained by the relationship between AA and GMRES.
- AA-PMHSS has improved the numerical properties of the inner solver compared with previous methods.
- AA-PMHSS is more flexible, as it is possible to tune the inner solver relax the tolerances.



## Conclusions and future work

### Future Work

- Restarted methods
- Orthogonalization
- General study on splittings

PSS splitting for saddle-point problems



## Conclusions and future work

Thank you!

Registration and travel support for this presentation was provided by **Professor W. Randolph Franklin**.

and by KTH Jubileumsanslaget.

Please join us at the MS on **Modern Preconditioners and Linear Solvers in Scientific Applications**. In this room tomorrow at 11.

- [1] M. I. Andersson, F. Liu, and S. Markidis, *Anderson accelerated pmhss for complex-symmetric linear systems*, in Proceedings of the 2024 SIAM Conference on Parallel Processing for Scientific Computing (PP), SIAM, 2024, pp. 39–52.

