## Nested Krylov Methods for Solving the Time-Harmonic Elastic Wave Equation at Multiple Frequencies

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## Context of this work (1/3)

#### Why are we here?

Definition: Shifted linear systems

We want to efficiently solve

$$(\mathcal{A} - \sigma_k I)\mathbf{x}_k = \mathbf{b}$$

for multiple shifts  $\sigma_1, ..., \sigma_{N_{\sigma}}, N_{\sigma} \in \mathbb{N}$ .

The *efficiency* of our numerical method heavily relies on the shift-invariance property of Krylov subspaces,

$$\mathcal{K}_m(\mathcal{A}, \mathbf{b}) \equiv \operatorname{span}\{\mathbf{b}, \mathcal{A}\mathbf{b}, ..., \mathcal{A}^{m-1}\mathbf{b}\} = \mathcal{K}_m(\mathcal{A} - \sigma I, \mathbf{b}).$$



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## Context of this work (2/3)

#### Who works in this area?

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#### Recent work

- A. Saibaba, T. Bakhos, and P. Kitanidis. A flexible Krylov solver for shifted systems with application to oscillatory hydraulic tomography. SIAM J. Sci. Comput. 35:3001-3023 (2013).
- M. Ahmad, D. Szyld, and M. van Gijzen. Preconditioned multishift BiCG for H<sub>2</sub>-optimal model reduction. Report 12-06-15, Temple University (2013).

#### Very recent work

- T. Bakhos, P. Kitanidis, S. Ladenheim, A. Saibaba, and D. Szyld. *Multipreconditioned GMRES for Shifted Systems.* Report 16-03-31, Temple University (2016).
- K. Soodhalter. Block Krylov Subspace Recycling for Shifted Systems with Unrelated Right-Hand Sides. SIAM J. Sci. Comput., 38(1):A302-A324 (2016).

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## Context of this work (3/3)

What does the present work contribute?

We present a nested Krylov solver for shifted linear systems:

- Use a multi-shift Krylov method as an inner method (the preconditioner),
- use a flexible multi-shift Krylov method as an outer method,
- **apply a single** *shifted Laplace* preconditioner as a first layer.

In this talk, we will concentrate on 1-2, and comment on 3 when presenting numerical tests.



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## Outlook

1 Nested Krylov methods for shifted systems

2 The time-harmonic elastic wave equation

#### 3 Numerical examples

- Example 1: Academic test case
- Example 2: Marmousi-II



Solve the shifted linear systems

$$(\mathcal{A} - \sigma_k I) \mathbf{x}_k = \mathbf{b}, \quad k = 1, ..., N_{\sigma},$$

$$\mathcal{K}_m(\mathcal{A},\mathbf{r}_0) = \mathcal{K}_m(\mathcal{A} - \sigma I,\mathbf{r}_0) \quad \forall \sigma.$$





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![](_page_11_Picture_6.jpeg)

## Inner method: multi-shift FOM

Classical result: In FOM, the residuals are:

$$\mathbf{r}_j = \mathbf{b} - \mathcal{A}\mathbf{x}_j = \dots = -h_{j+1,j}\mathbf{e}_j^T\mathbf{y}_j\mathbf{v}_{j+1}.$$

Thus, for the shifted residuals it holds:

$$\mathbf{r}_{j}^{(\sigma)} = \mathbf{b} - (\mathcal{A} - \sigma \mathbf{I})\mathbf{x}_{j}^{(\sigma)} = ... = -h_{j+1,j}^{(\sigma)}\mathbf{e}_{j}^{\mathsf{T}}\mathbf{y}_{j}^{(\sigma)}\mathbf{v}_{j+1}.$$

Hence, we obtain collinear residuals,

$$\mathbf{r}_{j}^{(\sigma)}=\gamma_{j}$$
 with factor  $\gamma_{j}=y_{j}^{(\sigma)}/y_{j}.$ 

#### Reference

V. Simoncini. *Restarted full orthogonalization method for shifted linear systems*. BIT Numerical Mathematics, 43 (2003).

![](_page_12_Picture_9.jpeg)

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![](_page_13_Picture_9.jpeg)

Use flexible GMRES in the outer loop,

$$(\mathcal{A} - \sigma I)\widehat{V}_{m}^{(\sigma)} = V_{m+1}\underline{H}_{m}^{(\sigma)},$$

where one column yields:

$$(\mathcal{A} - \sigma I) \underbrace{\mathcal{P}(\sigma)_j^{-1} \mathbf{v}_j}_{\text{inner loop}} = V_{m+1} \underline{\mathbf{h}}_j^{(\sigma)}, \quad 1 \leq j \leq m.$$

Recap: The "inner loop" is the truncated solution of  $(\mathcal{A} - \sigma I)$  with right-hand side  $\mathbf{v}_i$  using msFOM.

![](_page_14_Picture_6.jpeg)

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Recap: The "inner loop" is the truncated solution of  $(\mathcal{A} - \sigma I)$  with right-hand side  $\mathbf{v}_i$  using msFOM.

![](_page_15_Picture_6.jpeg)

The inner residuals are:

$$\mathbf{r}_{j}^{(\sigma)} = \mathbf{v}_{j} - (\mathcal{A} - \sigma I)\mathcal{P}(\sigma)_{j}^{-1}\mathbf{v}_{j},$$
  
$$\mathbf{r}_{j} = \mathbf{v}_{j} - \mathcal{A}\mathcal{P}_{j}^{-1}\mathbf{v}_{j}.$$

Imposing collinearity,  $\mathbf{r}_{j}^{(\sigma)} = \gamma_{j}\mathbf{r}_{j}$ , yields:

$$(\mathcal{A} - \sigma I)\mathcal{P}(\sigma)_j^{-1}\mathbf{v}_j = \gamma_j \mathcal{A} \mathcal{P}_j^{-1}\mathbf{v}_j - (\gamma_j - 1)\mathbf{v}_j$$

![](_page_16_Picture_5.jpeg)

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Krylov methods for linear elasticity problems

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![](_page_17_Picture_5.jpeg)

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Altogether,

$$(\mathcal{A} - \sigma I)\mathcal{P}(\sigma)_{j}^{-1}\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\sigma)}$$
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which yields:

$$\underline{\mathbf{H}}_{m}^{(\sigma)} = (\underline{\mathbf{H}}_{m} - \underline{\mathbf{I}}_{m}) \, \mathbf{\Gamma}_{m} + \underline{\mathbf{I}}_{m},$$

with  $\Gamma_m := diag(\gamma_1, ..., \gamma_j, ..., \gamma_m).$ 

![](_page_18_Picture_6.jpeg)

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![](_page_19_Picture_6.jpeg)

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![](_page_20_Picture_6.jpeg)

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![](_page_21_Picture_6.jpeg)

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Krylov methods for linear elasticity problems

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## Summary: Nested FOM-FGMRES

In nested FOM-FGMRES, we solve the following (small) optimization problems,

$$\begin{aligned} \mathbf{x}_{m}^{(\sigma)} &= \operatorname*{argmin}_{\mathbf{x}\in\widehat{\mathcal{V}}_{m}^{(\sigma)}} \|\mathbf{b} - (\mathcal{A} - \sigma I)\mathbf{x}\| \\ &= \operatorname*{argmin}_{\mathbf{y}\in\mathbb{C}^{m}} \|\mathbf{b} - (\mathcal{A} - \sigma I)\widehat{\mathcal{V}}_{m}\mathbf{y}\| \\ &= \operatorname*{argmin}_{\mathbf{y}\in\mathbb{C}^{m}} \|\mathbf{b} - \mathcal{V}_{m+1}\underline{\mathbf{H}}_{m}^{(\sigma)}\mathbf{y}\| \\ &= \operatorname*{argmin}_{\mathbf{y}\in\mathbb{C}^{m}} \|\beta\mathbf{e}_{1} - \left((\underline{\mathbf{H}}_{m} - \underline{\mathbf{I}}_{m})\,\mathbf{\Gamma}_{m}^{(\sigma)} + \underline{\mathbf{I}}_{m}\right)\mathbf{y}\|, \end{aligned}$$

where the entries of  $\Gamma_m^{(\sigma)}$  are collinearity factors of inner FOM.

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## Relation to Rational Krylov methods

The FOM-FGMRES Hessenberg matrix,

$$\underline{\mathbf{H}}_{m}^{(\sigma)} = (\underline{\mathbf{H}}_{m} - \underline{\mathbf{I}}_{m})\,\boldsymbol{\Gamma}_{m}^{(\sigma)} + \underline{\mathbf{I}}_{m}, \quad \boldsymbol{\Gamma}_{m}^{(\sigma)} := \textit{diag}(\gamma_{1}^{(\sigma)}, ..., \gamma_{m}^{(\sigma)}),$$

is closely related to the formula,

 $\underline{\mathrm{H}}(\sigma; T_m) = (\underline{\mathrm{H}}_m(T_m - \sigma I_m) + \underline{\mathrm{I}}_m), \quad T_m := \operatorname{diag}(\tau_1, ..., \tau_m),$ 

taken from [Saibaba et al., 2013].

![](_page_24_Picture_6.jpeg)

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## Relation to Rational Krylov methods

The FOM-FGMRES Hessenberg matrix,

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![](_page_25_Picture_6.jpeg)

### 1 Nested Krylov methods for shifted systems

### 2 The time-harmonic elastic wave equation

#### Numerical examples

- Example 1: Academic test case
- Example 2: Marmousi-II

![](_page_26_Picture_5.jpeg)

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Krylov methods for linear elasticity problems

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Continuous setting Elastic wave equation

$$-\omega_k^2 \rho(\mathbf{x}) \mathbf{u}_k - \nabla \cdot \sigma(\mathbf{u}_k) = \mathbf{s}$$

with boundary conditions

$$\begin{split} i\omega_k\rho(\mathbf{x})B\mathbf{u}_k + \sigma(\mathbf{u}_k)\mathbf{\hat{n}} &= \mathbf{0},\\ \sigma(\mathbf{u}_k)\mathbf{\hat{n}} &= \mathbf{0}, \end{split}$$

on  $\partial \Omega_a \cup \partial \Omega_r$ .

Discrete setting Solve

$$(K+i\omega_k C-\omega_k^2 M)\mathbf{u}_k=\mathbf{s}$$

with FEM matrices

$$\begin{split} \mathcal{K}_{ij} &= \int_{\Omega} \sigma(\boldsymbol{\varphi}_i) : \nabla \boldsymbol{\varphi}_j \, d\Omega, \\ \mathcal{M}_{ij} &= \int_{\Omega} \rho(\mathbf{x}) \boldsymbol{\varphi}_i \cdot \boldsymbol{\varphi}_j \, d\Omega, \\ \mathcal{C}_{ij} &= \int_{\partial \Omega_a} \rho(\mathbf{x}) B \boldsymbol{\varphi}_i \cdot \boldsymbol{\varphi}_j \, d\Gamma \end{split}$$

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Krylov methods for linear elasticity problems

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$$\begin{split} &i\omega_k\rho(\mathbf{x})B\mathbf{u}_k+\sigma(\mathbf{u}_k)\mathbf{\hat{n}}=\mathbf{0},\\ &\sigma(\mathbf{u}_k)\mathbf{\hat{n}}=\mathbf{0}, \end{split}$$

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Krylov methods for linear elasticity problems

The discretized problem

$$(K+i\omega_k C-\omega_k^2 M)\mathbf{u}_k=\mathbf{s}$$

can be re-formulated to

$$\begin{bmatrix} \begin{pmatrix} iC & K \\ I & 0 \end{pmatrix} - \frac{\omega_k}{\omega_k} \begin{pmatrix} M & 0 \\ 0 & I \end{pmatrix} \end{bmatrix} \begin{pmatrix} \omega_k \mathbf{u}_k \\ \mathbf{u}_k \end{pmatrix} = \begin{pmatrix} \mathbf{s} \\ 0 \end{pmatrix},$$

which is of block form:

$$(\mathcal{A} - \omega_k \mathcal{M})\mathbf{x}_k = \mathbf{b}.$$

![](_page_29_Picture_7.jpeg)

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Right-preconditioning yields,

 $(\mathcal{A} - \omega_k \mathcal{M}) \mathcal{P}_{\omega}^{-1} \mathbf{y}_k = \mathbf{b} \quad \Leftrightarrow \quad (\mathcal{A} \mathcal{P}^{-1} - \sigma_k I) \mathbf{y}_k = \mathbf{b},$ with  $\mathcal{P}^{-1} \equiv (\mathcal{A} - \tau \mathcal{M})^{-1}, \sigma_k = \omega_k / (\omega_k - \tau).$ 

$$\mathcal{P}^{-1} = \begin{bmatrix} I & \tau I \\ 0 & I \end{bmatrix} \begin{bmatrix} I & 0 \\ 0 & (K + i\tau C - \tau^2 M)^{-1} \end{bmatrix} \begin{bmatrix} 0 & I \\ I & -iC + \tau M \end{bmatrix}.$$

![](_page_30_Picture_4.jpeg)

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Right-preconditioning yields,

$$(\mathcal{A} - \omega_k \mathcal{M}) \mathcal{P}_{\omega}^{-1} \mathbf{y}_k = \mathbf{b} \quad \Leftrightarrow \quad (\mathcal{A} \mathcal{P}^{-1} - \sigma_k I) \mathbf{y}_k = \mathbf{b},$$
  
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![](_page_31_Picture_4.jpeg)

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![](_page_32_Picture_4.jpeg)

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Krylov methods for linear elasticity problems

![](_page_33_Figure_1.jpeg)

#### Reference

M. Baumann, R. Astudillo, Y. Qiu, E. Ang, M.B. Van Gijzen, and R.-É. Plessix. *An MSSS-Preconditioned Matrix Equation Approach for the Time-Harmonic Elastic Wave Equation at Multiple Frequencies.* Technical Report 16-04, Delft University of Technology (2016).

![](_page_33_Picture_4.jpeg)

![](_page_34_Picture_0.jpeg)

![](_page_34_Picture_2.jpeg)

#### 3 Numerical examples

- Example 1: Academic test case
- Example 2: Marmousi-II

![](_page_34_Picture_6.jpeg)

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# Numerical example # 1 as presented in [B./vG., 2014]

Test case from literature:

- $\Omega = [0,1] \times [0,1]$
- *h* = 0.01 implying
   *n* = 10.201 grid points
- system size:
   4n = 40.804
- $N_{\sigma} = 6$  frequencies
- point source at center

#### Reference

T. Airaksinen, A. Pennanen, and J. Toivanen. *A damping preconditioner for time-harmonic wave equations in fluid and elastic material.* J. Comput. Phys., **228**:1466-1479 (2009).

![](_page_35_Picture_9.jpeg)

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## Numerical example # 1 as presented in [B./vG., 2014]

## Preconditioned **multi-shift GMRES**:

- simultaneous solve
- linear convergence rates
- $\tau = (0.7 0.7i)\sigma_{max}$
- CPU time: 17.71s

![](_page_36_Figure_6.jpeg)

![](_page_36_Picture_7.jpeg)

## Numerical example # 1 as presented in [B./vG., 2014]

#### Preconditioned **nested FOM-FGMRES**:

- 20 inner iterations
- truncate when inner residual norm  $\sim 0.1$
- very few outer iterations
- CPU time: 9.12s

![](_page_37_Figure_6.jpeg)

(a)

![](_page_37_Picture_7.jpeg)

## Numerical example # 1 as presented in [B./vG., 2014]

Various combinations of inner/outer methods *possible*:

	multi-shift Krylov methods			
	msGMRES	msGMRESr	QMRIDR(4)	msIDR(4)
# inner iterations	-	20	-	-
# outer iterations	103	7	136	134
CPU time	17.71s	6.13s	22.35s	22.58s
	nested multi-shift Krylov methods			
	FOM-FGMRES	IDR(4)-FGMRES	FOM-FQMRIDR(4)	IDR(4)-FQMRIDR(4)
# inner iterations	20	25	30	30
# outer iterations	7	9	5	15
CPU time	9.12s	32.99s	8.14s	58.36s

Different combination (CMRH-FCMRH) is exploited in:

X.-M. Gu, T.-Z. Huang, B. Carpentieri, A. Imakura, K. Zhang, L. Du. *Variants of the CMRH method for solving multi-shifted non-Hermitian linear systems.* Technical Report, University of Groningen (2016).

![](_page_38_Picture_5.jpeg)

## Numerical example # 2

The Marmousi-II test case:

- 2D benchmark problem
- $f_k = \{1, 2, 4\}Hz$
- $\tau \in \mathbb{C}$  "optimal"
- 593,600 # dofs
- 20 ppw
- single source

![](_page_39_Figure_8.jpeg)

#### Reference

G. Rizzuti and W. Mulder. *Multigrid-based 'shifted-Laplacian'* preconditioning for the time-harmonic elastic wave equation. J. Comput. Phys., **317**:47-65 (2016).

M. Baumann @ AN16

## Numerical example # 2

 $- f_1 = 1 \text{ Hz}$  $- f_2 = 2 \text{ Hz}$  $- f_3 = 4 \text{ Hz}$  $10^{0}$  $10^{-1}$  $10^{-2}$ Relative residual norm  $10^{-3}$  $10^{-4}$  $10^{-5}$  $10^{-6}$  $10^{-7}$  $10^{-8}$ 100 700 0 200 300 400 500600 800 900 # iterations

Preconditioned multi-shift GMRES

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## Numerical example # 2

	# inner	# outer	CPU time
msGMRES	-	871	2:37:01
msGMRESr	100	18	4:43:41
FOM-FGMRES	100	14	3:27:35

#### References

- A. Frommer and U. Glässner. Restarted GMRES for Shifted Linear Systems. SIAM J. Sci. Comput., 19:15-26, (1998).
- M. Baumann and M. B. van Gijzen. Nested Krylov methods for shifted linear systems. SIAM J. Sci. Comput., 37(5), S90-S112 (2015).

![](_page_41_Picture_5.jpeg)

Krylov methods for linear elasticity problems

## Conclusions and future work

- $\checkmark$  inner-outer Krylov method for shifted systems
- ✓ single preconditioner based on MSSS structure
- ? optimal parameter choice for au~
  ightarrow Rational Krylov

![](_page_42_Figure_4.jpeg)

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## Thank you for your attention!

#### Further reading:

- M. Baumann and M. B. van Gijzen. *Nested Krylov methods for shifted linear systems.* SIAM J. Sci. Comput., **37**(5), S90-S112 (2015).
- M. Baumann, R. Astudillo, Y. Qiu, E. Ang, M.B. Van Gijzen, and R.-E. Plessix. An MSSS-Preconditioned Matrix Equation Approach for the Time-Harmonic Elastic Wave Equation at Multiple Frequencies. Technical report 16-04, TU Delft, 2016.

Research funded by Shell.

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![](_page_43_Picture_5.jpeg)

Krylov methods for linear elasticity problems

## MSSS matrix computations "in a nutshell"

### Definition: SSS matrix [Chandrasekaran et al. 2005]

Let A be an  $n \times n$  block matrix with sequentially semi-seperable structure. Then A can be written in the following block partitioned form

$$A_{i,j} = \begin{cases} U_i W_{i+1} \cdots W_{j-1} V_j^T, & \text{if } i < j; \\ D_i, & \text{if } i = j; \\ P_i R_{i-1} \cdots R_{j+1} Q_j^T, & \text{if } i > j. \end{cases}$$

- $\bullet$  linear computational complexity for  $(\cdot)^{-1}$
- limit off-diagonal rank with MOR
- MSSS: constructors are SSS matrices

![](_page_44_Figure_7.jpeg)

![](_page_44_Picture_8.jpeg)