# Fast Iterative Solution of the Time-Harmonic Elastic Wave Equation at Multiple Frequencies

Manuel M. Baumann

Email: m.m.baumann@tudelft.nl Delft Institute of Applied Mathematics Delft University of Technology Delft, The Netherlands

Royal Institute of Technology in Stockholm - February 1, 2018



# Nice to be back!

Hilling

3900

## Seismic Full-Waveform Inversion



Inverse problem:

- least-squares fit of measurements and simulations,
- requires *fast* forward solves.



# Seismic Full-Waveform Inversion



Inverse problem:

- least-squares fit of measurements and simulations,
- requires *fast* forward solves.

"Solve the linear systems of equations,

$$(K + i\omega_k C - \omega_k^2 M)\mathbf{x}_k = \mathbf{b},$$

efficiently (= fast and at low memory) for multiple frequencies. "



## Seismic Full-Waveform Inversion



Inverse problem:

- least-squares fit of measurements and simulations,
- requires *fast* forward solves.

(a)





The time-harmonic elastic wave equation

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

$$egin{aligned} \mathcal{K}_{ij} &= \int_\Omega \sigma(oldsymbol{arphi}_i) : 
abla oldsymbol{arphi}_j \; d\Omega, \ \mathcal{M}_{ij} &= \int_\Omega 
ho(\mathbf{x}) oldsymbol{arphi}_i \cdot oldsymbol{arphi}_j \; d\Omega, \ \mathcal{C}_{ij} &= \int_{\partial\Omega_a} 
ho(\mathbf{x}) B oldsymbol{arphi}_i \cdot oldsymbol{arphi}_j \; d\Gamma. \end{aligned}$$

</₽> < E> < E>



Two main approaches for solving,

$$(K + i\omega_k C - \omega_k^2 M) \mathbf{x}_k = \mathbf{b}, \quad k > 1.$$



M. Baumann

(a)

Two main approaches for solving,

$$(K + i\omega_k C - \omega_k^2 M) \mathbf{x}_k = \mathbf{b}, \quad k > 1.$$

#### Shifted systems

$$\begin{pmatrix} \begin{bmatrix} iC & K \\ I & 0 \end{bmatrix} - \omega_k \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix} \end{pmatrix} \begin{bmatrix} \omega_k \mathbf{x}_k \\ \mathbf{x}_k \end{bmatrix} = \begin{bmatrix} \mathbf{b} \\ \mathbf{0} \end{bmatrix}$$

- Most work for  $\mathbf{x}_0$  (at  $\omega = 0$ )
- Requires preconditioning



Two main approaches for solving,

$$(K + i\omega_k C - \omega_k^2 M) \mathbf{x}_k = \mathbf{b}, \quad k > 1.$$

Shifted systems

Matrix equation

$$\begin{pmatrix} \begin{bmatrix} iC & K \\ I & 0 \end{bmatrix} - \omega_k \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix} \end{pmatrix} \begin{bmatrix} \omega_k \mathbf{x}_k \\ \mathbf{x}_k \end{bmatrix} = \begin{bmatrix} \mathbf{b} \\ \mathbf{0} \end{bmatrix}$$

 $K\mathbf{X} + iC\mathbf{X}\Omega - M\mathbf{X}\Omega^2 = \mathbf{B}$ 

- Most work for  $\mathbf{x}_0$  (at  $\omega = 0$ )
- Requires preconditioning

- Solve for  $\mathbf{X} = [\mathbf{x}_1, ..., \mathbf{x}_N]$ all-at-once
- Requires preconditioning



Two main approaches for solving,

$$(K + i\omega_k C - \omega_k^2 M) \mathbf{x}_k = \mathbf{b}, \quad k > 1.$$

Shifted systems

Matrix equation

$$\begin{pmatrix} \begin{bmatrix} iC & K \\ I & 0 \end{bmatrix} - \omega_k \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix} \end{pmatrix} \begin{bmatrix} \omega_k \mathbf{x}_k \\ \mathbf{x}_k \end{bmatrix} = \begin{bmatrix} \mathbf{b} \\ \mathbf{0} \end{bmatrix}$$

 $K\mathbf{X} + iC\mathbf{X}\Omega - M\mathbf{X}\Omega^2 = \mathbf{B}$ 

- Most work for  $\mathbf{x}_0$  (at  $\omega = 0$ )
- Requires preconditioning

- Solve for  $\mathbf{X} = [\mathbf{x}_1, ..., \mathbf{x}_N]$ all-at-once
- Requires preconditioning



#### **Table of Content**



2 Optimal preconditioner for ms-GMRES







< □ ▶

## What's a shifted linear system?

# Definition Shifted linear systems are of the form $(A - \omega I)\mathbf{x}^{(\omega)} = \mathbf{b},$

where  $\omega \in \mathbb{C}$  is the *shift*.

For the simultaneous solution, **Krylov methods** are well-suited because of the *shift-invariance* property:

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

#### "Proof" (shift-invariance)

For 
$$m = 2$$
:  $\mathcal{K}_2(A, \mathbf{b}) = \operatorname{span}\{\mathbf{b}, A\mathbf{b}\}\$   
 $\mathcal{K}_2(A - \omega I, \mathbf{b}) = \operatorname{span}\{\mathbf{b}, A\mathbf{b} - \omega \mathbf{b}\} = \operatorname{span}\{\mathbf{b}, A\mathbf{b}\}$ 



· • @ • • = • • = •

# What's a shifted linear system?

# Definition Shifted linear systems are of the form $(A - \omega I)\mathbf{x}^{(\omega)} = \mathbf{b},$

where  $\omega \in \mathbb{C}$  is the *shift*.

For the simultaneous solution, **Krylov methods** are well-suited because of the *shift-invariance* property:

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

# "Proof" (shift-invariance) For m = 2: $\mathcal{K}_2(A, \mathbf{b}) = \operatorname{span}\{\mathbf{b}, A\mathbf{b}\}$ $\mathcal{K}_2(A - \omega I, \mathbf{b}) = \operatorname{span}\{\mathbf{b}, A\mathbf{b} - \omega \mathbf{b}\} = \operatorname{span}\{\mathbf{b}, A\mathbf{b}\}$



## What's a shifted linear system?

# Definition Shifted linear systems are of the form $(A - \omega I)\mathbf{x}^{(\omega)} = \mathbf{b},$

where  $\omega \in \mathbb{C}$  is the *shift*.

For the simultaneous solution, **Krylov methods** are well-suited because of the *shift-invariance* property:

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

"Proof" (shift-invariance)  
For 
$$m = 2$$
:  $\mathcal{K}_2(A, \mathbf{b}) = \operatorname{span}\{\mathbf{b}, A\mathbf{b}\}$   
 $\mathcal{K}_2(A - \omega I, \mathbf{b}) = \operatorname{span}\{\mathbf{b}, A\mathbf{b} - \omega \mathbf{b}\} = \operatorname{span}\{\mathbf{b}, A\mathbf{b}\}$ 



- ∢ /= ► - ∢ = -

#### Multi-shift GMRES

[Frommer/Glässner, 1998]

· • @ • • Ξ • • Ξ •

After *m* steps of Arnoldi, we have,

$$AV_m = V_{m+1}\underline{H}_m,$$

and the approximate solution yields:

$$\mathbf{x}_m pprox V_m \mathbf{y}_m, \quad ext{where } \mathbf{y}_m = \operatorname*{argmin}_{\mathbf{y} \in \mathbb{C}^m} \| \underline{\mathbf{H}}_m \mathbf{y} - \| \mathbf{b} \| \mathbf{e}_1 \| \, .$$

For shifted systems, we get

$$(A - \omega I)V_m = V_{m+1}(\underline{\mathbf{H}}_m - \omega \underline{\mathbf{I}}_m),$$

and, therefore,

$$\mathbf{x}_m^{(\omega)} pprox V_m \mathbf{y}_m^{(\omega)}, \quad ext{where } \mathbf{y}_m^{(\omega)} = \operatorname*{argmin}_{\mathbf{y} \in \mathbb{C}^m} \left\| \underline{\mathbf{H}}_m^{(\omega)} \mathbf{y} - \| \mathbf{b} \| \mathbf{e_1} \right\|.$$



#### Multi-shift GMRES

[Frommer/Glässner, 1998]

</₽> < E> < E>

After *m* steps of Arnoldi, we have,

$$AV_m = V_{m+1}\underline{H}_m,$$

and the approximate solution yields:

$$\mathbf{x}_m pprox V_m \mathbf{y}_m, \quad ext{where } \mathbf{y}_m = \operatorname*{argmin}_{\mathbf{y} \in \mathbb{C}^m} \| \underline{\mathbf{H}}_m \mathbf{y} - \| \mathbf{b} \| \mathbf{e}_1 \| \, .$$

For shifted systems, we get

$$(A - \omega I)V_m = V_{m+1}(\underline{\mathsf{H}}_m - \omega \underline{\mathsf{I}}_m),$$

and, therefore,

$$\mathbf{x}_m^{(\omega)} pprox V_m \mathbf{y}_m^{(\omega)}, \quad ext{where } \mathbf{y}_m^{(\omega)} = \operatorname*{argmin}_{\mathbf{y} \in \mathbb{C}^m} \left\| \underline{\mathbb{H}}_m^{(\omega)} \mathbf{y} - \| \mathbf{b} \| \mathbf{e}_1 \right\|.$$

**T**UDelft

#### Multi-shift GMRES

[Frommer/Glässner, 1998]

/⊒ ▶ ∢ ⊒

After *m* steps of Arnoldi, we have,

$$AV_m = V_{m+1}\underline{H}_m,$$

and the approximate solution yields:

$$\mathbf{x}_m pprox \mathbf{V}_m \mathbf{y}_m, \quad ext{where } \mathbf{y}_m = \operatorname*{argmin}_{\mathbf{y} \in \mathbb{C}^m} \| \underline{\mathbf{H}}_m \mathbf{y} - \| \mathbf{b} \| \mathbf{e}_1 \|.$$

For shifted systems, we get

$$(A - \omega I)V_m = V_{m+1}(\underline{\mathsf{H}}_m - \omega \underline{\mathsf{I}}_m),$$

and, therefore,

$$\mathbf{x}_m^{(\omega)} pprox \mathbf{V}_m \mathbf{y}_m^{(\omega)}, \quad \text{where } \mathbf{y}_m^{(\omega)} = \operatorname*{argmin}_{\mathbf{y} \in \mathbb{C}^m} \left\| \underline{\mathbf{H}}_m^{(\omega)} \mathbf{y} - \| \mathbf{b} \| \mathbf{e}_1 \right\|.$$

**T**UDelft

#### The shift-and-invert preconditioner

$$\mathcal{A} := \begin{bmatrix} iC & K \\ I & 0 \end{bmatrix}, \quad \mathcal{B} := \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix}$$

The multi-frequency problem reads,

$$(\mathcal{A} - \omega_k \mathcal{B})\mathbf{x}_k = \mathbf{b}, \quad k = 1, ..., N_{\omega},$$

with a single preconditioner  $\mathcal{P}(\tau) := (\mathcal{A} - \tau \mathcal{B})$ :

$$(\mathcal{A} - \omega_k \mathcal{B})\mathcal{P}_k^{-1}\mathbf{y}_k = \mathbf{b} \quad \Leftrightarrow \quad (\mathcal{A}(\mathcal{A} - \tau \mathcal{B})^{-1} - \eta_k I)\mathbf{y}_k = \mathbf{b}$$



#### The shift-and-invert preconditioner

$$\mathcal{A} := \begin{bmatrix} iC & K \\ I & 0 \end{bmatrix}, \quad \mathcal{B} := \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix}$$

< /⊒ ► < ≣

The multi-frequency problem reads,

$$(\mathcal{A} - \omega_k \mathcal{B})\mathbf{x}_k = \mathbf{b}, \quad k = 1, ..., N_{\omega},$$

with a single preconditioner  $\mathcal{P}(\tau) := (\mathcal{A} - \tau \mathcal{B})$ :

$$(\mathcal{A} - \omega_k \mathcal{B}) \mathcal{P}_k^{-1} \mathbf{y}_k = \mathbf{b} \quad \Leftrightarrow \quad (\mathcal{A} (\mathcal{A} - \tau \mathcal{B})^{-1} - \eta_k I) \mathbf{y}_k = \mathbf{b}$$



#### The shift-and-invert preconditioner

$$\mathcal{A} := \begin{bmatrix} iC & K \\ I & 0 \end{bmatrix}, \quad \mathcal{B} := \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix}$$

· • @ • • Ξ • • Ξ •

The multi-frequency problem reads,

$$(\mathcal{A} - \omega_k \mathcal{B})\mathbf{x}_k = \mathbf{b}, \quad k = 1, ..., N_\omega,$$

with a single preconditioner  $\mathcal{P}(\tau) := (\mathcal{A} - \tau \mathcal{B})$ :

$$(\mathcal{A} - \omega_k \mathcal{B}) \mathcal{P}_k^{-1} \mathbf{y}_k = \mathbf{b} \quad \Leftrightarrow \quad (\mathcal{C} - \eta_k I) \mathbf{y}_k = \mathbf{b}$$
  
•  $\mathcal{C} := \mathcal{A} (\mathcal{A} - \tau \mathcal{B})^{-1}$   
•  $\eta_k := \omega_k / (\omega_k - \tau)$ 

M. Baumann

**T**UDelft

#### **Table of Content**



#### 2 Optimal preconditioner for ms-GMRES

Nested multi-shift Krylov methods

#### 4 Conclusions



## Optimization of seed frequency

$$\left(\mathcal{A}(\mathcal{A}-\boldsymbol{ au}\mathcal{B})^{-1}-rac{\omega_k}{\omega_k-\boldsymbol{ au}}I
ight)\mathbf{y}_k=\mathbf{b}$$

**Theorem:** GMRES convergence bound [Saad, Iter. Methods] Let the eigenvalues of a matrix be enclosed by a circle with radius *R* and center *c*. Then the GMRES-residual norm after *i* iterations  $\|\mathbf{r}^{(i)}\|$  satisfies,

$$\frac{\|\mathbf{r}^{(i)}\|}{\|\mathbf{r}^{(0)}\|} \le c_2(X) \left(\frac{R(\tau)}{|c(\tau)|}\right)^i,$$

where X is the matrix of eigenvectors, and  $c_2(X)$  its condition number in the 2-norm.



## Optimization of seed frequency

$$\left(\mathcal{A}(\mathcal{A}- au\mathcal{B})^{-1}-rac{\omega_k}{\omega_k- au}I
ight)\mathbf{y}_k=\mathbf{b}$$

**Theorem:** msGMRES convergence bound [Saad, Iter. Methods] Let the eigenvalues of a matrix be enclosed by a circle with radius  $R_k$  and center  $c_k$ . Then the GMRES-residual norm after *i* iterations  $\|\mathbf{r}_k^{(i)}\|$  satisfies,

$$\frac{\|\mathbf{r}_{k}^{(i)}\|}{\|\mathbf{r}^{(0)}\|} \leq c_{2}(X) \left(\frac{R_{k}(\tau)}{|c_{k}(\tau)|}\right)^{i}, \quad k = 1, ..., N_{\omega},$$

where X is the matrix of eigenvectors, and  $c_2(X)$  its condition number in the 2-norm.

**T**UDelft

#### The preconditioned spectra - no damping



**T**UDelft

#### The preconditioned spectra - no damping



**T**UDelft

# The preconditioned spectra – with damping $\epsilon > 0$ $\hat{\omega}_k := (1 - \epsilon i)\omega_k$





· • @ • • = • • = •

# The preconditioned spectra – with damping $\epsilon > 0$ $\hat{\omega}_k := (1 - \epsilon i)\omega_k$





1 / 29

· • @ • • = • • = •

Lemma: Optimal seed shift for ms-GMRES [B/vG. 2016] (i) For  $\lambda \in \Lambda[\mathcal{AB}^{-1}]$  it holds  $\Im(\lambda) > 0$ . (ii) The preconditioned spectra are enclosed by circles of radii  $R_k$  and center points  $c_k$ . (iii) The points  $\{c_k\}_{k=1}^{N_{\omega}} \subset \mathbb{C}$  described in statement (ii) lie on a circle with center c and radius R. (iv) Consider the preconditioner  $\mathcal{P}(\tau^*) = \mathcal{A} - \tau^* \mathcal{B}$ . An optimal seed frequency  $\tau^*$  for preconditioned multi-shift GMRES is given by,  $\tau^*(\epsilon) = \min_{\tau \in \mathbb{C}} \max_{k=1,\dots,N_{\ell}} \left( \frac{R_k(\tau)}{|\alpha|} \right) = \dots =$  $=\frac{2\omega_1\omega_{N_{\omega}}}{\omega_1+\omega_{N_{\omega}}}-i\frac{\sqrt{[\epsilon^2(\omega_1+\omega_{N_{\omega}})^2+(\omega_{N_{\omega}}-\omega_1)^2]\omega_1\omega_{N_{\omega}}}}{\omega_1+\omega_{N_{\omega}}}$ 



**Proof:** Maybe later.



M. Baumann

< □ > < □ > < □ > < □ > < □ >

#### **Proof:** There is an App for that.





(a)

#### Numerical experiments I

#### Set-up: An *elastic* wedge problem.



#### Reference

M. Baumann and M.B. van Gijzen (2017). *An Efficient Two-Level Preconditioner for Multi-Frequency Wave Propagation Problems*. DIAM Technical Report **17-03**, TU Delft.

## **T**UDelft

- ∢ 🗇 🕨 -

#### Convergence behavior and spectral bounds

For any  $\tau$ ...





M. Baumann

<u> イロト ( 同 ) ( 三 ) ( 三 )</u>

#### Convergence behavior and spectral bounds

For the optimal  $\tau^*$ ...





Preconditioning the Elastic Wave Equation

(a)

#### An interval splitting strategy

Suppose  $n_p \ge 2$  parallel processors are available.



#### **T**UDelft

### **Table of Content**



2 Optimal preconditioner for ms-GMRES



#### 4 Conclusions



M. Baumann

Solve the shifted linear systems

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

with a nested Krylov method, and preserve,

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





Solve the shifted linear systems

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

with a nested Krylov method, and preserve,

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





Solve the shifted linear systems

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

with a <u>nested</u> Krylov method, and preserve,

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





Solve the shifted linear systems

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

with a <u>nested</u> Krylov method, and preserve,

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





#### Multi-shift FOM as inner method

Classical result: In FOM, the residuals are

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

Thus, for the shifted residuals it holds:

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

which gives  $\gamma = y_j^{(\omega)}/y_j$ .

#### Reference

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



#### Flexible multi-shift GMRES as outer method

Use flexible GMRES in the outer loop,

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

where one column yields

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

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



#### Flexible multi-shift GMRES as outer method

Use flexible GMRES in the outer loop,

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

where one column yields

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

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

Thus, the inner residuals are:

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

Imposing 
$$\mathbf{r}_{i}^{(\omega)} = \gamma \mathbf{r}_{j}$$
 yields:

$$(A - \omega I)\mathcal{P}(\omega)_j^{-1}\mathbf{v}_j = \gamma A \mathcal{P}_j^{-1}\mathbf{v}_j - (\gamma - 1)\mathbf{v}_j \qquad (*)$$

Note that the right-hand side in (\*) is a preconditioned shifted system!



Thus, the inner residuals are:

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

Imposing 
$$\mathbf{r}_{j}^{(\omega)} = \gamma \mathbf{r}_{j}$$
 yields:  
 $(A - \omega I)\mathcal{P}(\omega)_{j}^{-1}\mathbf{v}_{j} = \gamma A \mathcal{P}_{j}^{-1}\mathbf{v}_{j} - (\gamma - 1)\mathbf{v}_{j}$  (\*)

Note that the right-hand side in (\*) is a preconditioned shifted system!



Altogether,

$$(\boldsymbol{A} - \boldsymbol{\omega} \boldsymbol{I}) \mathcal{P}(\boldsymbol{\omega})_{j}^{-1} \mathbf{v}_{j} = V_{m+1} \underline{\mathbf{h}}_{j}^{(\boldsymbol{\omega})}$$
$$\gamma A \mathcal{P}_{j}^{-1} \mathbf{v}_{j} - (\gamma - 1) \mathbf{v}_{j} = V_{m+1} \underline{\mathbf{h}}_{j}^{(\boldsymbol{\omega})}$$
$$\gamma V_{m+1} \underline{\mathbf{h}}_{j} - V_{m+1} (\gamma - 1) \underline{\mathbf{e}}_{j} = V_{m+1} \underline{\mathbf{h}}_{j}^{(\boldsymbol{\omega})}$$
$$V_{m+1} (\gamma \underline{\mathbf{h}}_{j} - (\gamma - 1) \underline{\mathbf{e}}_{j}) = V_{m+1} \underline{\mathbf{h}}_{j}^{(\boldsymbol{\omega})}$$

which yields:

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

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



Altogether,

$$(A - \omega I)\mathcal{P}(\omega)_{j}^{-1}\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$\gamma A\mathcal{P}_{j}^{-1}\mathbf{v}_{j} - (\gamma - 1)\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$\gamma V_{m+1}\underline{\mathbf{h}}_{j} - V_{m+1}(\gamma - 1)\underline{\mathbf{e}}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$V_{m+1}(\gamma \underline{\mathbf{h}}_{j} - (\gamma - 1)\underline{\mathbf{e}}_{j}) = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$

which yields:

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

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



Altogether,

$$(A - \omega I)\mathcal{P}(\omega)_{j}^{-1}\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$\gamma A\mathcal{P}_{j}^{-1}\mathbf{v}_{j} - (\gamma - 1)\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$\gamma V_{m+1}\underline{\mathbf{h}}_{j} - V_{m+1}(\gamma - 1)\underline{\mathbf{e}}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$V_{m+1}(\gamma \underline{\mathbf{h}}_{j} - (\gamma - 1)\underline{\mathbf{e}}_{j}) = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$

which yields:

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

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



Altogether,

$$(A - \omega I)\mathcal{P}(\omega)_{j}^{-1}\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$\gamma A\mathcal{P}_{j}^{-1}\mathbf{v}_{j} - (\gamma - 1)\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$\gamma V_{m+1}\underline{\mathbf{h}}_{j} - V_{m+1}(\gamma - 1)\underline{\mathbf{e}}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$V_{m+1}(\gamma \underline{\mathbf{h}}_{j} - (\gamma - 1)\underline{\mathbf{e}}_{j}) = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$

which yields:

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

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



22 / 29

Altogether,

$$(A - \omega I)\mathcal{P}(\omega)_{j}^{-1}\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$\gamma A\mathcal{P}_{j}^{-1}\mathbf{v}_{j} - (\gamma - 1)\mathbf{v}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$\gamma V_{m+1}\underline{\mathbf{h}}_{j} - V_{m+1}(\gamma - 1)\underline{\mathbf{e}}_{j} = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$
$$V_{m+1}(\gamma \underline{\mathbf{h}}_{j} - (\gamma - 1)\underline{\mathbf{e}}_{j}) = V_{m+1}\underline{\mathbf{h}}_{j}^{(\omega)}$$

which yields:

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

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

M. Baumann

#### Summary: Nested FOM-FGMRES

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

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

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

**T**UDelft

I → I = I

# Numerical experiments II

Test case from literature:

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

# f, = 5,000 f\_ = 10,000 6 f<sub>o</sub> = 15,000 f, = 20,000 f<sub>5</sub> = 25,000 f<sub>o</sub> = 30,000 0.5

< /⊒ > < ≣

#### Reference

M. Baumann and M.B. van Gijzen (2015). *Nested Krylov methods for shifted linear systems*. SIAM J. Sci. Comput., **37**(5), S90-S112.



#### Numerical experiments II as presented in [B./vG., 2015]

# Preconditioned **multi-shift GMRES**:

- simultaneous solve
- linear convergence rates

• 
$$\tau = \tau^* = (0.3 - 0.7i)\omega_{max}$$

• CPU time: 17.71s





#### Numerical experiments II as presented in [B./vG., 2015]

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

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



(a)



#### Numerical experiments II as presented in [B./vG., 2015]

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).



#### Conclusions

What I have shown today:

- ✓ Optimal  $\tau^*$  for multi-shift GMRES
- Nested multi-shift FOM-FGMRES

What I have not shown today:

- MSSS preconditioning techniques
- ✓ matrix equation (global GMRES)
- ✓ large-scale 3D examples
- X deflation, model-order reduction, ...





#### References

- M. Baumann and M.B. van Gijzen (2015). *Nested Krylov methods for shifted linear systems.* SIAM J. Sci. Comput., **37**(5), S90-S112.
- M. Baumann, R. Astudillo, Y. Qiu, E. Ang, M.B. van Gijzen, and R.-E. Plessix (2017). An MSSS-Preconditioned Matrix Equation Approach for the Time-Harmonic Elastic Wave Equation at Multiple Frequencies. Springer Computat. Geosci., DOI: 10.1007/s10596-017-9667-7.
- M. Baumann and M.B. van Gijzen (2017). Efficient iterative methods for multi-frequency wave propagation problems: A comparison study. Procedia Comput. Sci., Vol. 108, pp. 645-654.
- M. Baumann and M.B. van Gijzen (2017). An Efficient Two-Level Preconditioner for Multi-Frequency Wave Propagation Problems. DIAM Technical Report **17-03**, TU Delft. [under review]



#### **Lemma:** Optimal seed shift for msGMRES [B/vG. 2016] (i) For $\lambda \in \Lambda[\mathcal{AB}^{-1}]$ it holds $\Im(\lambda) > 0$ . (ii) The preconditioned spectra are enclosed by circles of radii $R_k$ and center points $c_k$ . (iii) The points $\{c_k\}_{k=1}^{N_{\omega}} \subset \mathbb{C}$ described in statement (ii) lie on a circle with center c and radius R. (iv) Consider the preconditioner $\mathcal{P}(\tau^*) = \mathcal{A} - \tau^* \mathcal{B}$ . An optimal seed frequency $\tau^*$ for preconditioned multi-shift GMRES is given by, $\mathbf{D}$

$$\tau^{*}(\epsilon) = \min_{\tau \in \mathbb{C}} \max_{k=1,...,N_{\omega}} \left( \frac{K_{k}(\tau)}{|c_{k}|} \right) = ... =$$
$$= \frac{2\omega_{1}\omega_{N_{\omega}}}{\omega_{1} + \omega_{N_{\omega}}} - i \frac{\sqrt{[\epsilon^{2}(\omega_{1} + \omega_{N_{\omega}})^{2} + (\omega_{N_{\omega}} - \omega_{1})^{2}]\omega_{1}\omega_{N_{\omega}}}}{\omega_{1} + \omega_{N_{\omega}}}$$

**T**UDelft

#### The preconditioned spectra – Proof (1/4)

**Proof.** (i) We have to show  $\Im(\omega) \ge 0$  for,

$$\begin{bmatrix} iC & K \\ I & 0 \end{bmatrix} x = \omega \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix} x$$

or, alternatively ( $\lambda = i\omega$ ), consider the QEP,

$$(K + \lambda C + \lambda^2 M)v = 0.$$

§3.8		come in pairs $(\lambda, \lambda)$	$\lambda$ then x is a left eigenvector of $\overline{\lambda}$
P5	M Hermitian positive	$\operatorname{Re}(\lambda) \leq 0$	
§3.8	definite, C, K Hermitian		
	positive semidefinite		
P6	M, C symmetric positive	$\lambda$ s are real and negative,	n linearly independent
§3.9	definite, K symmetric	gap between $n$ largest and	eigenvectors associated with

$$\Re(\lambda) \leq 0 \; \Rightarrow \; \Im(\omega) \geq 0$$



· • @ • • = • • = •

#### The preconditioned spectra – Proof (1/4)

**Proof.** (i) We have to show  $\Im(\omega) \ge 0$  for,

$$\begin{bmatrix} iC & K \\ I & 0 \end{bmatrix} x = \omega \begin{bmatrix} M & 0 \\ 0 & I \end{bmatrix} x$$

or, alternatively ( $\lambda = i\omega$ ), consider the QEP,

$$(K + \lambda C + \lambda^2 M)v = 0.$$

§3.8		come in pairs $(\lambda, \lambda)$	$\lambda$ then x is a left eigenvector of $\overline{\lambda}$
P5	M Hermitian positive	$\operatorname{Re}(\lambda) \leq 0$	
§3.8	definite, C, K Hermitian		
-	positive semidefinite		
P6	M, C symmetric positive	$\lambda$ s are real and negative,	n linearly independent
§3.9	definite, K symmetric	gap between $n$ largest and	eigenvectors associated with
	1 it a		

$$\Re(\lambda) \leq 0 \; \Rightarrow \; \Im(\omega) \geq 0$$



· • @ • • = • • = •

#### The preconditioned spectra – Proof (2/4)

(ii) The preconditioned spectra are enclosed by circles.

Factor out  $\mathcal{AB}^{-1}$ ,

$$\mathcal{C} - \eta_k I = \mathcal{A}(\mathcal{A} - \tau \mathcal{B})^{-1} - \eta_k I = \mathcal{A}\mathcal{B}^{-1}(\mathcal{A}\mathcal{B}^{-1} - \tau I)^{-1} - \eta_k I,$$

and note that

$$\mathbf{\Lambda}[\mathcal{AB}^{-1}] \ni \lambda \mapsto \frac{\lambda}{\lambda - \tau} - \frac{\omega_k}{\omega_k - \tau},$$

is a Möbius transformation<sup>(\*)</sup>.

#### Reference

M.B. van Gijzen, Y.A. Erlangga, C. Vuik. *Spectral Analysis of the Discrete Helmholtz Operator Preconditioned with a Shifted Laplacian.* SIAM J. Sci. Comput., **29**(5), 1942–1958 (2007)



The preconditioned spectra – Proof (3/4)

- (iii) Spectra are bounded by circles  $(c_k, R)$ . These center point  $\{c_k\}_{k=1}^{N_{\omega}}$  lie on a 'big circle'  $(\underline{c}, \underline{R})$ .
- 1. Construct center:

$$\underline{\mathsf{c}} = \left(0, \frac{\epsilon |\tau|^2}{2\Im(\tau)(\Im(\tau) + \epsilon \Re(\tau))}\right) \in \mathbb{C}$$

2. A point  $c_k$  has constant distance to  $\underline{c}$ :

 $\underline{\mathbf{R}}^2 = \|c_k - \underline{\mathbf{c}}\|_2^2 = \frac{|\tau|^2(\epsilon^2 + 1)}{4(\Im(\tau) + \epsilon \Re(\tau))^2} \quad \text{(independent of } \omega_k\text{)}$ 



# The preconditioned spectra – with damping $\epsilon > 0$ $\hat{\omega}_k := (1 - \epsilon i)\omega_k$





< □ ▶

· • @ • • = • • = •

## The preconditioned spectra – Proof (4/4)

(iv) Find optimal 
$$\tau^*$$
.  

$$\tau^* = \min_{\tau \in \mathbb{C}} \max_{k=1,...,N_{\omega}} \left(\frac{R}{|c_k|}\right)$$

$$|c_k| = f(\underline{c}, \underline{R}, \varphi_k)$$

$$\text{oplar coordinates}$$

$$\frac{\partial \tau}{\partial \varphi} = 0 \text{ (optimize along } \varphi)$$

0.0

0.2

real part (relative)

< □ > < □ > < □ > < □ > < □ >

1.00 0.96 0.92

0.88 0.84 0.80 0.76 0.72 0.68 0.64

 $\hat{\omega}_{\mathrm{max}}$ 

0.8

1.0

## The preconditioned spectra – Proof (4/4)



M. Baumann

9 / 29

0.96

0.88

0.84

0.76

0.72

0.68

1.0

< /⊒ > < ≣

### The preconditioned spectra – Proof (4/4)

