

Nested Krylov methods for shifted linear systems

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Motivation (1/3)

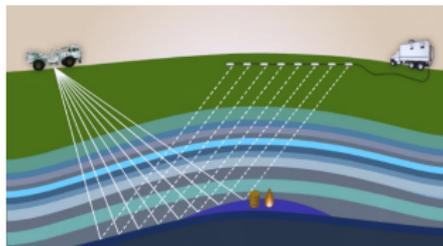
Full-waveform inversion

PDE-constrained optimization:

$$\min_{\rho(\mathbf{x}), c_p(\mathbf{x}), c_s(\mathbf{x})} \|\mathbf{u}_{sim} - \mathbf{u}_{meas}\|,$$

where in our application:

- \mathbf{u}_{sim} is the (numerical) solution of the **elastic wave equation**,
- \mathbf{u}_{meas} is obtained from measurements,
- ρ, c_p, c_s are properties of **earth layers** we are interested in.



The modeling is done in frequency-domain.

Motivation (2/3)

The discrete forward model

An FEM discretization of the time-harmonic, inhomogeneous elastic wave equation at **multiple frequencies** ω_k is given by:

$$(K + i\omega_k C - \omega_k^2 M)\mathbf{u}_k = \mathbf{s}, \quad k = 1, \dots, N,$$

where

- K, C, M are sparse and symmetric,
- \mathbf{s} usually models a point source,
- we need to compute the **displacement vector** \mathbf{u}_k for multiple frequencies (shifts) ω_k .

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Motivation (3/3)

Multi-shift Krylov methods

We can re-formulate the previous problem to:

$$\left[\begin{pmatrix} iC & K \\ I & 0 \end{pmatrix} - \omega_k \begin{pmatrix} M & 0 \\ 0 & I \end{pmatrix} \right] \begin{pmatrix} \omega_k \mathbf{u}_k \\ \mathbf{u}_k \end{pmatrix} = \begin{pmatrix} \mathbf{s} \\ 0 \end{pmatrix},$$

which is of the form:

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

Shift-invariance of Krylov subspaces:

$$\mathcal{K}_m(\mathcal{A}, \mathbf{r}_0) \equiv \text{span}\{\mathbf{r}_0, \mathcal{A}\mathbf{r}_0, \dots, \mathcal{A}^{m-1}\mathbf{r}_0\} = \mathcal{K}_m(\mathcal{A} - \omega I, \mathbf{r}_0)$$

is challenging to *preserve* when preconditioning!

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Preconditioners for shifted linear systems

2004 (Complex) shifted Laplace preconditioner:

$$\mathcal{P} = (\mathcal{A} - \tau\mathcal{M}), \quad \tau \approx \{\omega_1, \dots, \omega_N\}$$

2007 Many shifted Laplace preconditioners:

$$\mathcal{P}_j = (\mathcal{A} - \tau_j\mathcal{M})$$

2013 Polynomial preconditioners [Plenary talk at PRECON13]

2014 Question: Can we use a Krylov method as preconditioner?
↪ Nested Krylov methods

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Outline

- 1 The shifted Laplace preconditioner and its relation to Möbius transformations
- 2 Inner-outer Krylov methods for shifted linear systems
- 3 Numerical experiments
- 4 Conclusion

The shifted Laplacian

The *generalized* shifted Laplace preconditioner,

$$\mathcal{P} = (\mathcal{A} - \tau\mathcal{M}), \quad \tau \in \mathbb{C}, \quad (*)$$

has two benefits:

- 1 it transforms our problem to shifted linear systems and, hence, enables the benefits of **shift-invariant Krylov spaces**,
- 2 it maps the original **spectrum** to circles of known center and radius.

Moreover, (*) is **easy to apply** because $\tau \in \mathbb{C}$ leads to a damped problem \rightsquigarrow multigrid works!

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The shifted Laplacian

For $\mathcal{P} = (\mathcal{A} - \tau\mathcal{M})$, the following relation holds:

$$(\mathcal{A} - \omega_k\mathcal{M})\mathcal{P}_k^{-1} = \mathcal{A}\mathcal{P}^{-1} - \eta_k(\omega)I, \quad (**)$$

with

- $\mathcal{P}_k^{-1} = \frac{\tau}{\tau - \omega_k}\mathcal{P}^{-1}$
- $\eta_k = \omega_k / (\omega_k - \tau)$
- τ is a free parameter (seed shift)

For the spectrum of the RHS in (**), we see:

$$\sigma(\mathcal{A}\mathcal{M}^{-1}) \ni \lambda \mapsto \frac{\lambda}{\lambda - \tau} - \eta_k$$

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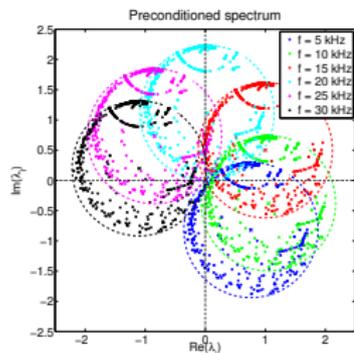
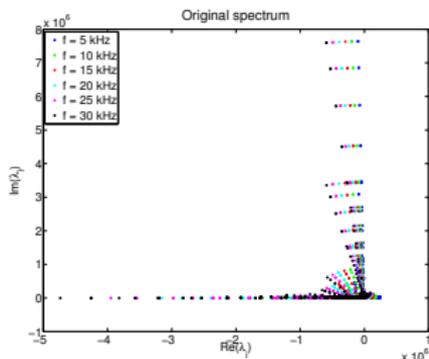
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The shifted Laplacian - Spectral analysis

Compare:

$$\sigma(\mathcal{A} - \omega_k \mathcal{M}) \quad \text{vs.} \quad \sigma((\mathcal{A} - \omega_k \mathcal{M})\mathcal{P}_k^{-1})$$



Open question: What's the optimal τ for equidistantly spaced frequencies $\omega_1, \dots, \omega_N$???

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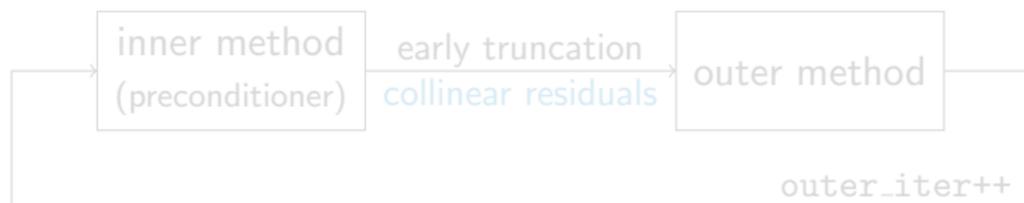
Nested (inner-outer) multi-shift Krylov methods

Solve the preconditioned shifted problem, $\mathcal{B} := \mathcal{A}\mathcal{P}^{-1}$,

$$(\mathcal{B} - \eta_k I) \mathbf{x}_k = \mathbf{b}, \quad k = 1, \dots, N,$$

with a nested Krylov method based on

$$\mathcal{K}_k(\mathcal{B}, \mathbf{r}_0) = \mathcal{K}_k(\mathcal{B} - \eta I, \mathbf{r}_0) \quad \forall \eta.$$



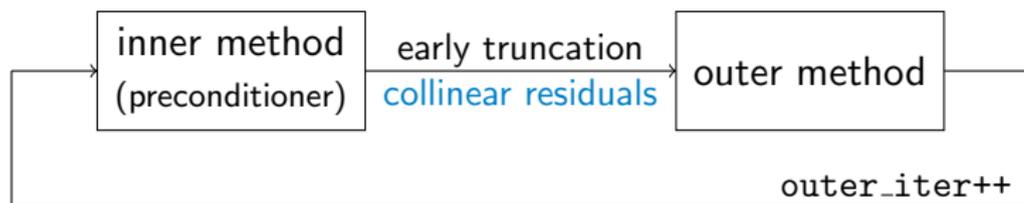
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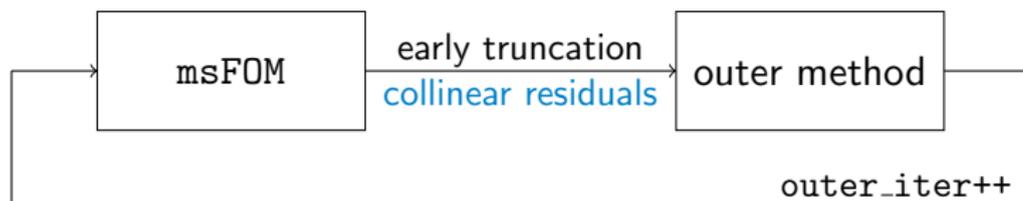
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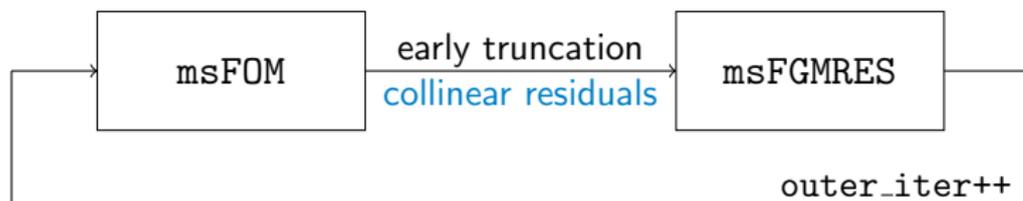
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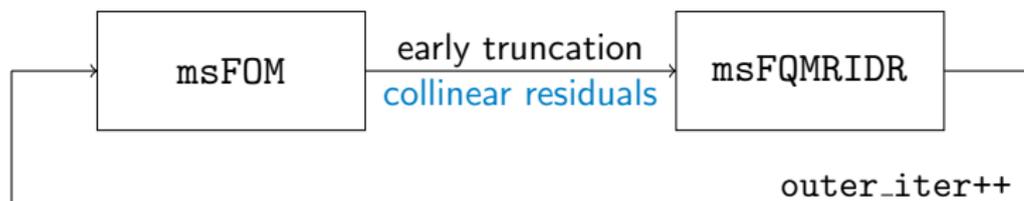
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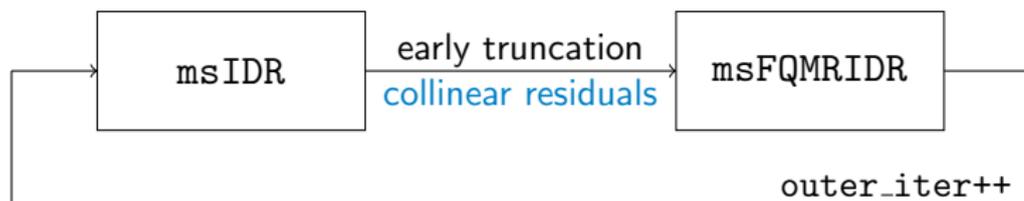
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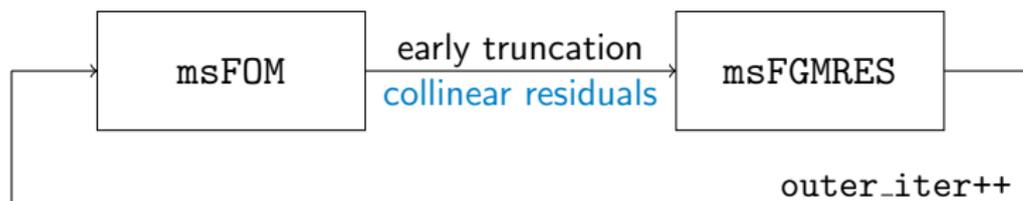
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Inner method: multi-shift FOM

Classical result: In FOM, the residuals are:

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

Thus, for the shifted residuals it holds:

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Hence, we obtain **collinear residuals**,

$$\mathbf{r}_j^{(\eta)} = \gamma \mathbf{r}_j,$$

with factor $\gamma = y_j^{(\eta)} / y_j$.

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Outer method: flexible multi-shift GMRES (1/3)

Use flexible GMRES in the outer loop,

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

where one column yields:

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

Recap: The “inner loop” is the truncated solution of $(\mathcal{B} - \eta I)$ with right-hand side \mathbf{v}_j using [msFOM](#).

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Outer method: flexible multi-shift GMRES (2/3)

The inner residuals are:

$$\begin{aligned}\mathbf{r}_j^{(\eta)} &= \mathbf{v}_j - (\mathcal{B} - \eta I)\mathcal{P}(\eta)_j^{-1}\mathbf{v}_j, \\ \mathbf{r}_j &= \mathbf{v}_j - \mathcal{B}\mathcal{P}_j^{-1}\mathbf{v}_j.\end{aligned}$$

Imposing $\mathbf{r}_j^{(\eta)} = \gamma \mathbf{r}_j$ yields:

$$(\mathcal{B} - \eta I)\mathcal{P}(\eta)_j^{-1}\mathbf{v}_j = \gamma \mathcal{B}\mathcal{P}_j^{-1}\mathbf{v}_j - (\gamma - 1)\mathbf{v}_j$$

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Altogether,

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which yields:

$$\underline{\mathbf{H}}_m^{(\eta)} = (\underline{\mathbf{H}}_m - \underline{\mathbf{I}}_m) \underline{\mathbf{\Gamma}}_m + \underline{\mathbf{I}}_m,$$

with $\underline{\mathbf{\Gamma}}_m := \text{diag}(\gamma_1, \dots, \gamma_m)$.

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Summary: Nested FOM-FGMRES

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

$$\begin{aligned}\mathbf{x}_m^{(\eta)} &= \operatorname{argmin}_{\mathbf{x} \in \widehat{\mathcal{V}}_m} \|\mathbf{b} - (\mathcal{B} - \eta I)\mathbf{x}\| \\ &= \operatorname{argmin}_{\mathbf{y} \in \mathbb{C}^m} \left\| \mathbf{b} - (\mathcal{B} - \eta I)\widehat{\mathcal{V}}_m \mathbf{y} \right\| \\ &= \operatorname{argmin}_{\mathbf{y} \in \mathbb{C}^m} \left\| \mathbf{b} - V_{m+1} \underline{H}_m^{(\eta)} \mathbf{y} \right\| \\ &= \operatorname{argmin}_{\mathbf{y} \in \mathbb{C}^m} \left\| \beta \mathbf{e}_1 - \left((\underline{H}_m - \underline{I}_m) \Gamma_m^{(\eta)} + \underline{I}_m \right) \mathbf{y} \right\|,\end{aligned}$$

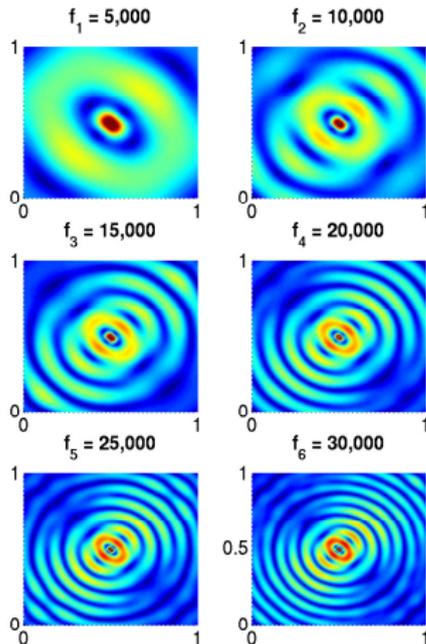
where the entries of $\Gamma_m^{(\eta)}$ are **collinearity factors of inner FOM**.

Numerical experiments (1/4)

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:
 $4n = 40.804$
- $N = 6$ frequencies
- point source at center



Reference

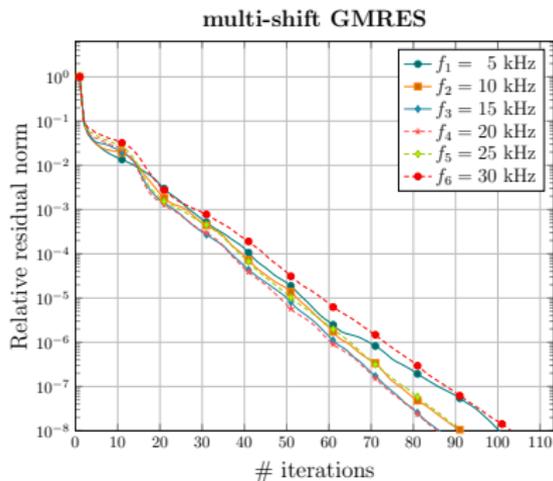
- T. Airaksinen, A. Pennanen, J. Toivanen, *A damping preconditioner for time-harmonic wave equations in fluid and elastic material*. Journal of Computational Physics, 2009.

Numerical experiments (2/4)

as presented in [B./vG., 2014]

Preconditioned **multi-shift GMRES**:

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

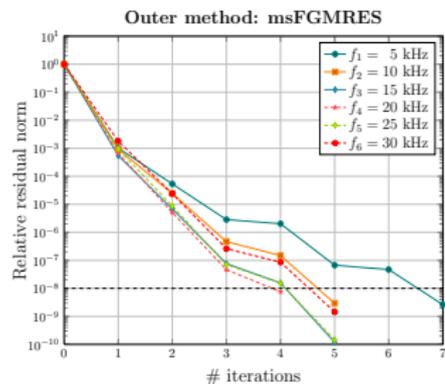
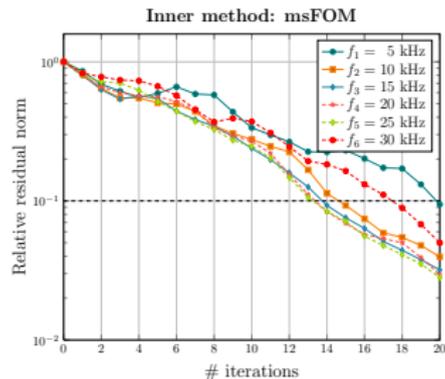


Numerical experiments (3/4)

as presented in [B./vG., 2014]

Preconditioned **nested** FOM-FGMRES:

- 20 inner iterations
- truncate when inner residual norm ~ 0.1
- very few outer iterations
- CPU time: **9.12s**



Numerical experiments (4/4)

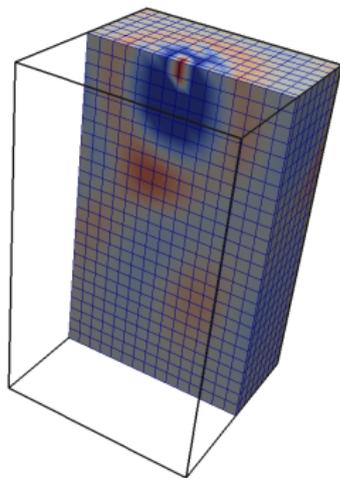
as presented in [B./vG., 2014]

Various combinations of nested algorithms:

	multi-shift Krylov methods			
	msGMRES	rest_msGMRES	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

Conclusions and future work

- ✓ Inner-outer Krylov methods for $A\mathbf{x} = \mathbf{b}$ are widely used
~> We present an **extension to shifted linear systems**
- ✓ The **shifted Laplace preconditioner** is applied as a *first layer*
- ? Future work: **3D problems**
 - ▶ discretization using TU/e package `nutils` (high-order FEM)
 - ▶ approximate shifted Laplacian with AGMG
 - ▶ nested solver in Fortran90



Thank you for your attention!

Further reading:



M. Baumann and M. B. van Gijzen. *Nested Krylov methods for shifted linear systems*. SIAM Journal on Scientific Computing (SISC), Special Issue Copper Mountain Conference 2014 [in press].

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