

An iterative multilevel method for computing wavefields in frequency-domain seismic inversion

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Outline of the Talk

Objective: iterative method for frequency-domain wavefield computation

- acoustic (Helmholtz) equation
- Helmholtz solver
 - direct vs. iterative methods
 - problems with iterative methods
- First ingredient: Preconditioning for Krylov methods
 - Shifted Laplacian
- Second ingredient: Multilevel Krylov (MK) methods
 - MK on preconditioned Helmholtz
- Numerical examples
- Conclusions

Frequency-domain acoustic wave equation

$$\mathcal{H}u = -\rho(x)\nabla \cdot \left(\frac{1}{\rho(x)}\nabla u(x) \right) - \omega^2(x)u(x) = b$$

↓

$$H[\omega]u = b$$

Illustration

Frequency domain seismic imaging

For one shot, one frequency:

- forward modeling: $Hu = b \rightarrow u = H^{-1}b$
- back-propagation: $H^*v = \delta u \rightarrow v = (H^*)^{-1}\delta u$
- imaging: $\delta m = \text{Re}(u \odot v)$

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How to compute u and v ?

Factorization of H into upper and lower triangular matrix

Also known as **direct methods** or **Gaussian-elimination based methods**

Direct Methods

Cost of computing u (and v):

$$\mathcal{O}(n^{d^d}), \quad d = 2, 3$$

Memory:

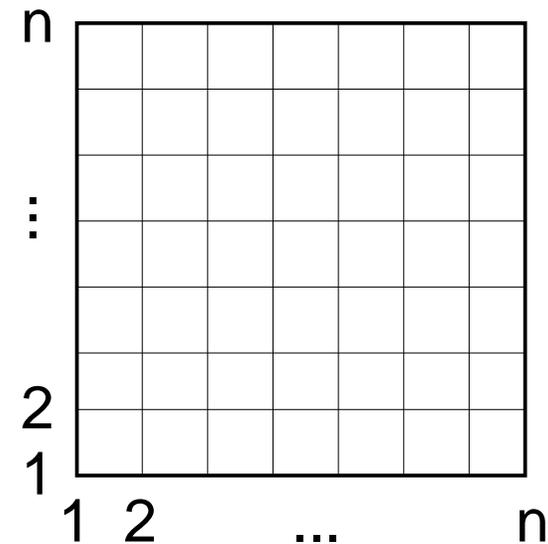
$$\mathcal{O}(n^{d-1}n^d), \quad d = 2, 3$$

Typical problems, $n = 10^3$.

In 2D:

- Computational cost: $\sim 10^{12}$ flops
- Memory: $\sim 10^9$ (1 Giga) units

In 2D, still possible, but not in 3D!



In 3D:

- Computational cost: $\sim 10^{27}$ flops
- Memory: $\sim 10^{15}$ (1 Peta) units

Illustration

Is it possible to replace **direct methods**?

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Yes! use iterative methods

Example:

- basic iterative methods: Jacobi, Gauß-Seidel
- **Krylov methods**: CG, GMRES, Bi-CGSTAB, etc

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Facts:

- Computational cost: $\mathcal{O}(n^d)$ flops **per iteration**, in total, $n_{iter}\mathcal{O}(n^d)$
- Memory: $\mathcal{O}(n^d)$ units

- **not robust methods**

Worst case, $n_{iter} = \mathcal{O}(n^d) \rightarrow$ computational cost: $\mathcal{O}(n^{d+2})$ flops

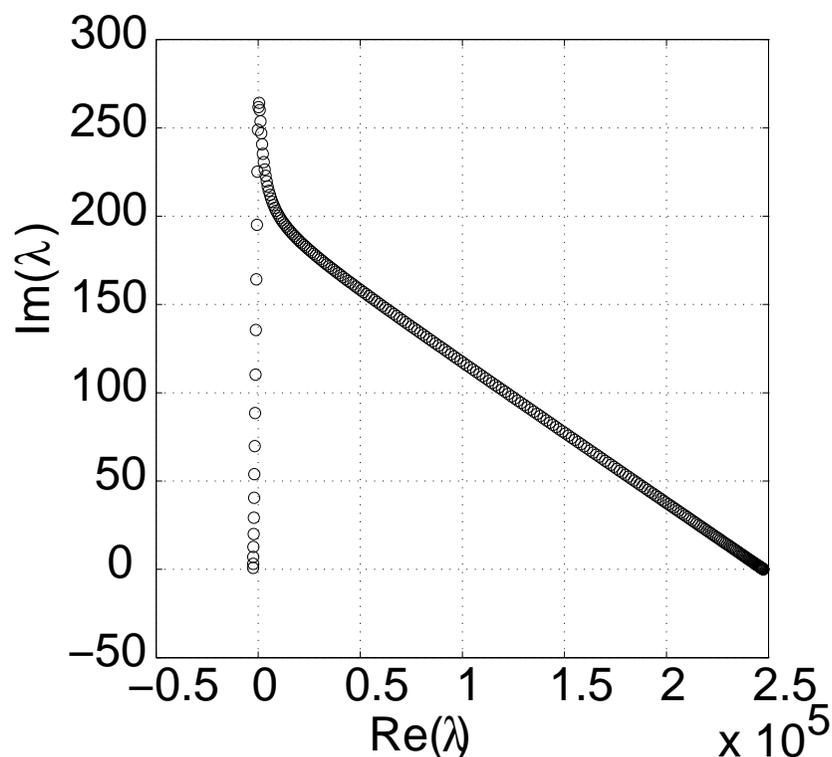
Ideally, $n_{iter} = \mathcal{O}(1) \rightarrow$ computational cost: $\mathcal{O}(n^d)$

Iterative Methods for Helmholtz problems

From theory:

Relation between convergence of iterative methods and eigenvalues of H

1D Helmholtz equation, $k := 2\pi fL/c = 50$.



- Small eigenvalues close to zero
- Large eigenvalues unbounded

Ill-conditioned

- Real parts of eigenvalues:
change signs \rightarrow Indefinite

Slow convergence or divergence

Tackling **ill-conditioning**:

- use preconditioning matrix M
eigenvalues of HM^{-1} are more clustered

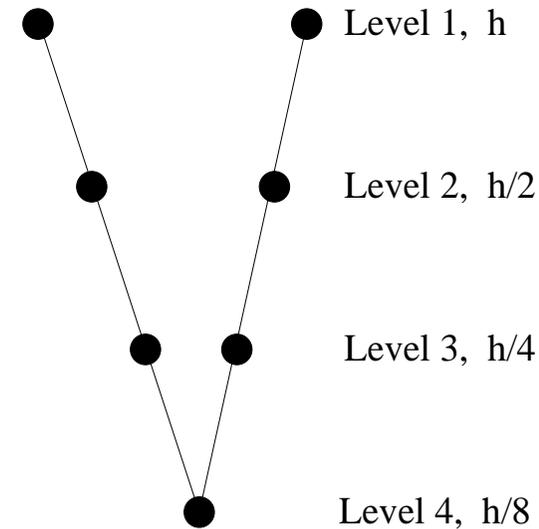
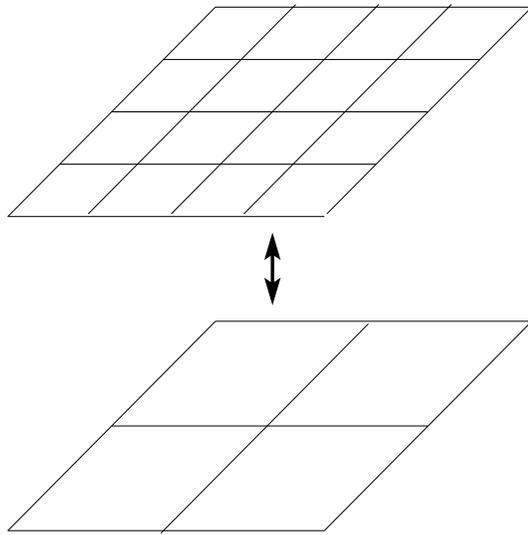
If $M^{-1} \approx H^{-1}$, eigenvalues of HM^{-1} are close to 1 \rightarrow fast convergence

- use multilevel technique
error components associated with small eigenvalues are corrected on the coarser grid

Indefiniteness is the most difficult part to handle!

Intermezzo: Multilevel Method

Solve $Hu = b$ on a hierarchy of grids of different mesh size.



Multigrid (MG):

- reduce nonsmooth errors by basic iterative methods: Jacobi/Gauß-Seidel
- reduce smooth errors on the coarse grid (smaller system)

Efficient methods ($O(n^d)$ methods) for non-indefinite systems

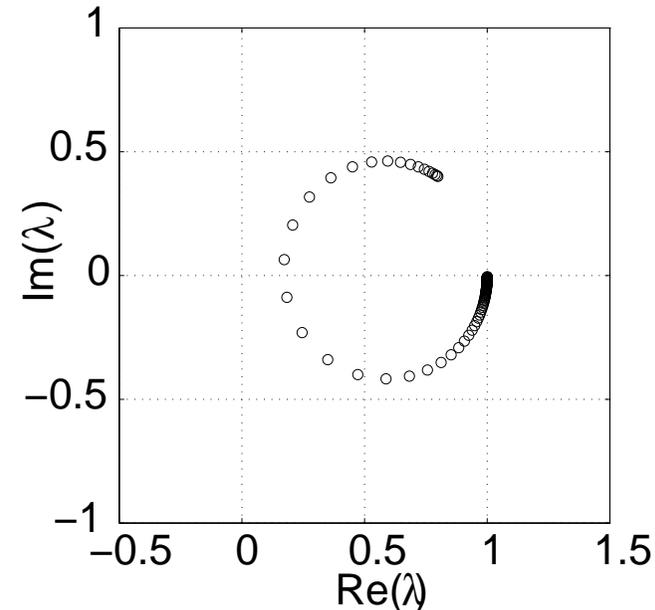
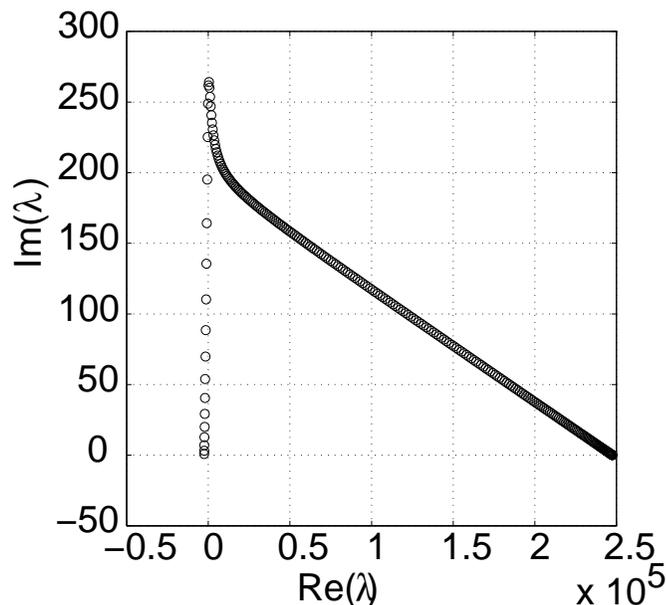
Preconditioner: tackling “indefiniteness”

First step: preconditioner which makes **indefinite system** **definite**

In this case, use shifted-Laplace (damped Helmholtz) operator:

$$M \hat{=} -\rho(x)\nabla \cdot \left(\frac{1}{\rho(x)} \nabla \right) - \left(1 - \frac{1}{2}\hat{j} \right) \omega^2(x), \quad \hat{j} = \sqrt{-1}.$$

Example: 1D Helmholtz, $k := \underline{2\pi fL/c} = 50$.



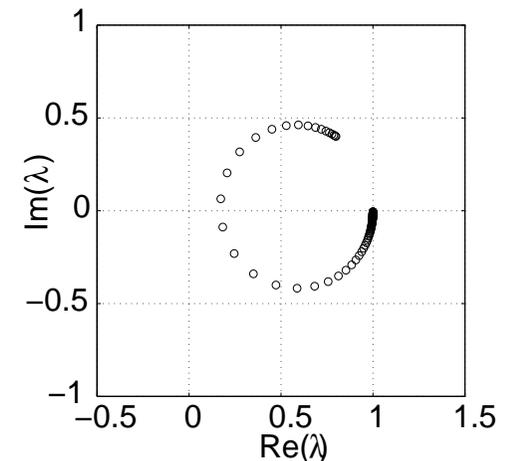
[E., Oosterlee, Vuik, 2006]

Preconditioner: tackling “indefiniteness”

Important aspects:

- the system HM^{-1} becomes **definite**
 - **Krylov methods should converge easier**, meaning $n_{iter} < n^d$
- M is favorable for multigrid (due to damping term)
 - M^{-1} is computed by one multigrid iteration $\rightarrow \mathcal{O}(n^d)$ for preconditioning
- Computational cost: $n_{iter}\mathcal{O}(n^d)$

- Largest eigenvalues are bounded above by 1
- Still **ill-conditioned** (see small eigenvalues)



Example with Marmousi velocity model

Frequency (Hz)	1	10	20	30
w/o M	17445	6623	14687	–
with M	39	54	98	144

Applications:

- 2D: [Riyanti, E., Plessix, Mulder, Vuik, Oosterlee, 2006]
[Duff, Gratton, Pinel, Vasseur]
- 3D: [Riyanti, Kononov, Erlangga, Plessix, Mulder, Vuik, Oosterlee, 2006]

Multilevel Krylov: tackling “ill-condition”

Problem remained with HM^{-1} : ill conditioning

Second step: make this system well-conditioned

Solution:

use operator which shifts small eigenvalues to the largest eigenvalue but keeps the upper bound the same

This is possible with the multilevel operator

$$Q = \overbrace{I - Z\hat{H}^{-1}Z^T HM^{-1}}^{\text{shift small eigenvalues to 0}} + \overbrace{Z\hat{H}^{-1}Z^T}^{\text{shift zero eigenvalues to 1}}, \quad \hat{H} = Z^T HM^{-1}Z,$$

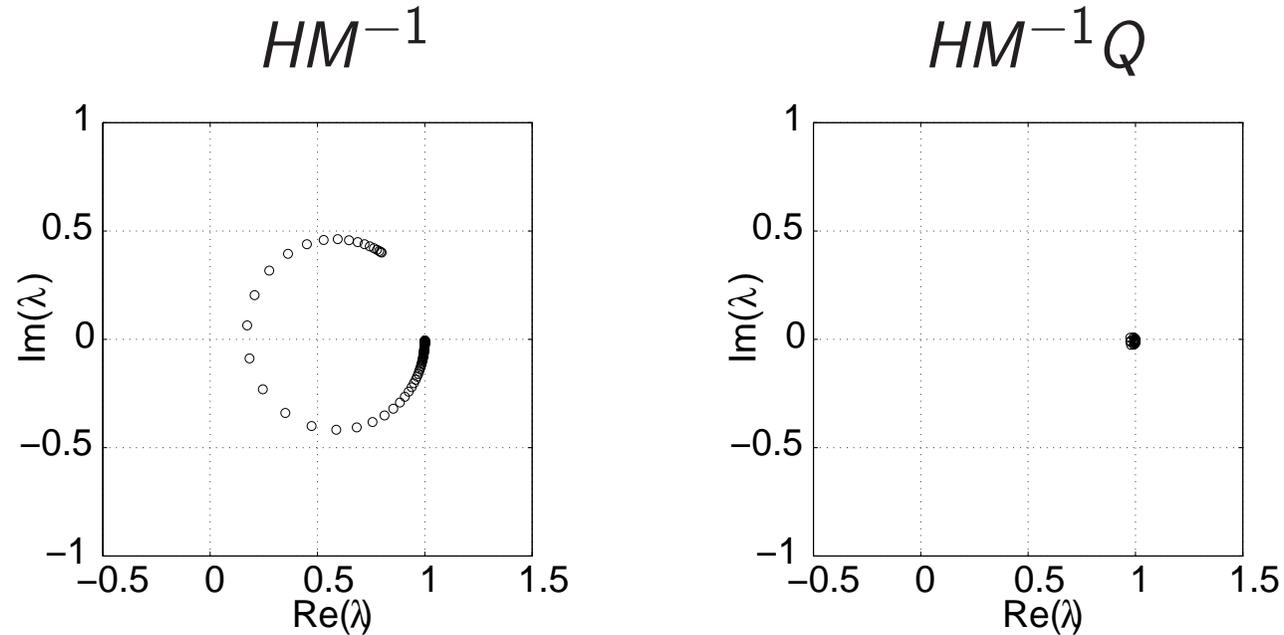
with Z linear mapping from coarse grid to fine grid.

\hat{H} implicitly contains information of small eigenvalues to be shifted.

[E., Nabben, 2007]

Multilevel Krylov (MK) for the Helmholtz equation

Example: 1D Helmholtz, $k := 2\pi fL/c = 50$.



Notice shift of small eigenvalues towards one.

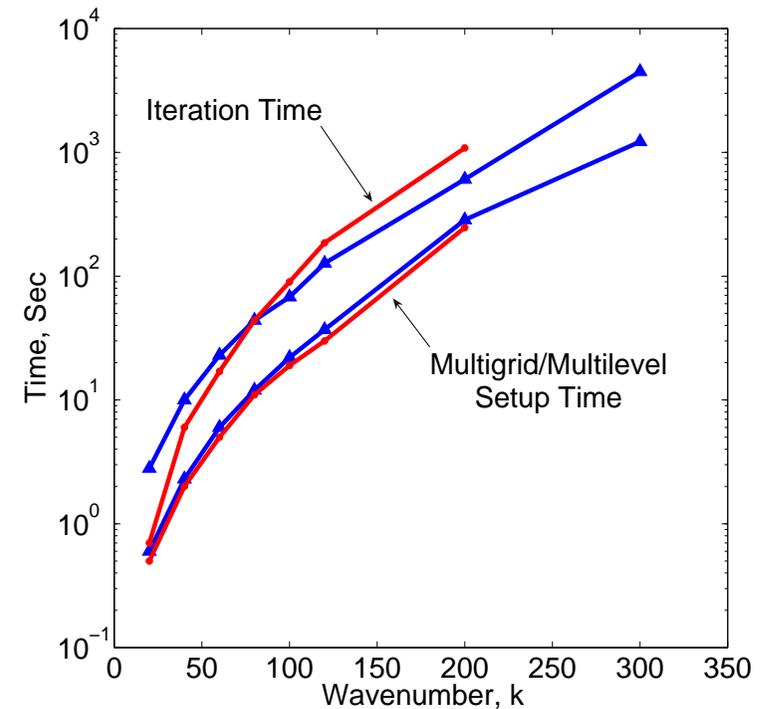
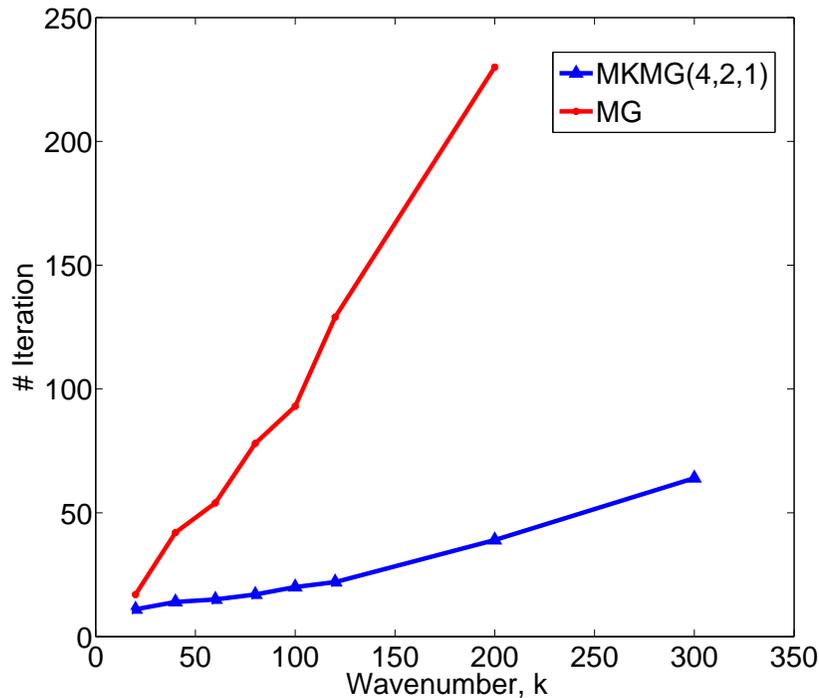
This is very favorable for fast convergence of Krylov methods

Numerical experiments: constant wavenumber

Improvement in #iterations and CPU time

High wavenumbers $k := 2\pi fL/c$

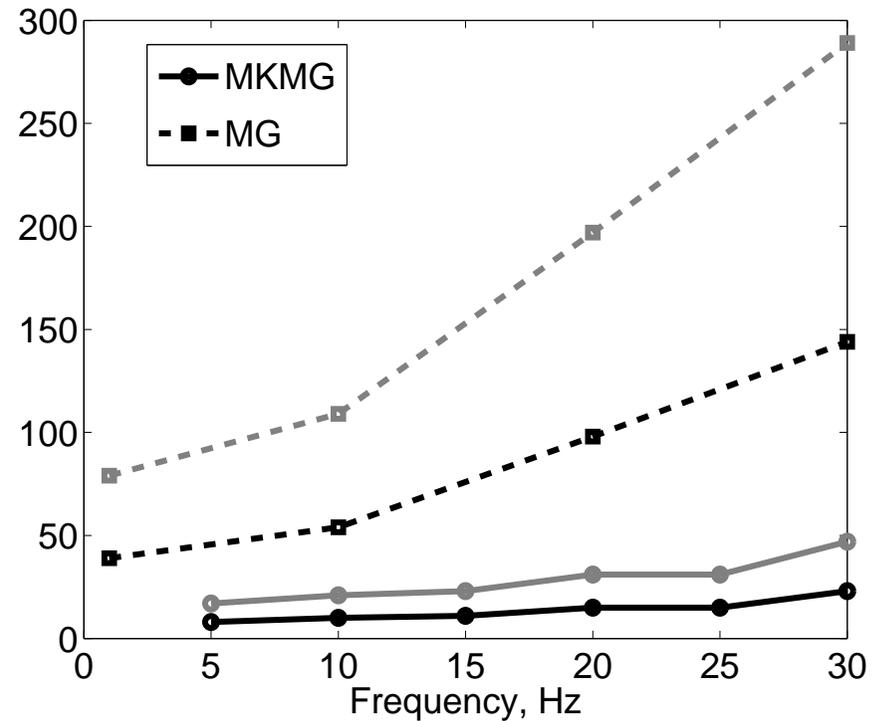
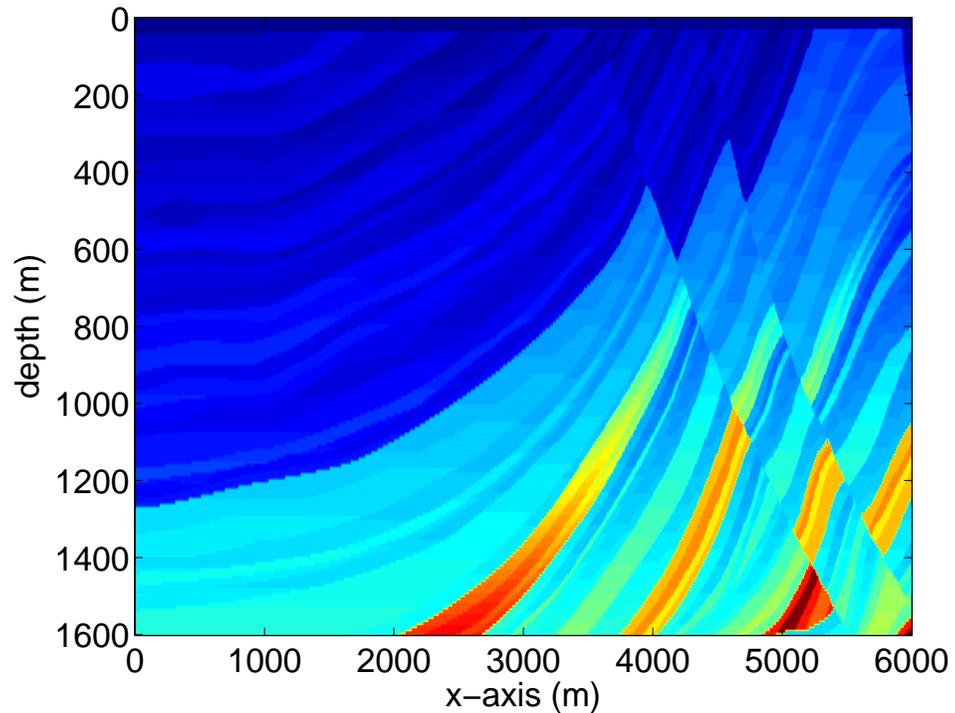
Grid resolution: 15 points per wavelength



$k = 300$: equivalent to $f = 60$ Hz for typical 1000 m depth. 750^2 grid points.

Numerical experiments: Marmousi

(—): No. of iterations. (—): No. of matrix-vector multiplies.



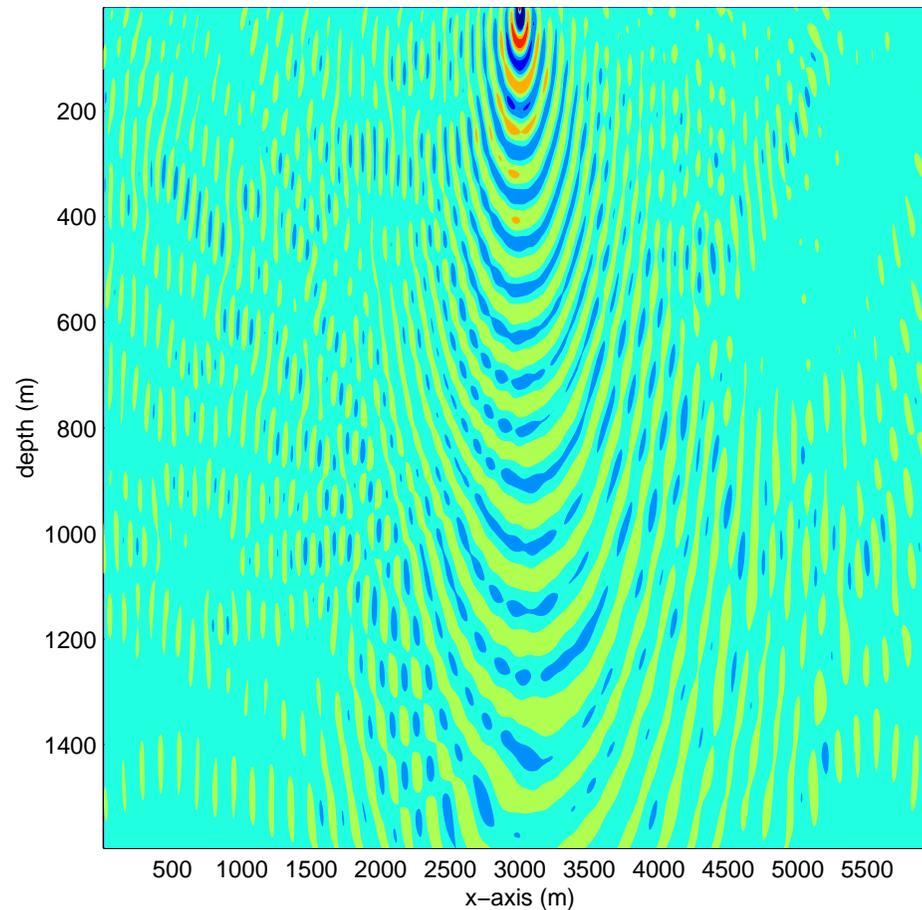
MKMG : GMRES on $HM^{-1}Q\tilde{u} = b$.

MG : Bi-CGSTAB on $HM^{-1}\hat{u} = b$ [E., Oosterlee, Vuik, 2006]

Convergence can be made less independent of frequencies

Numerical experiments: Marmousi

Computed wavefield, $f = 20$ Hz



Influence of gridsize to the convergence

Frequency (Hz)	5	10	15	20	25	30
Grid adapted to f	8	10	11	15	15	23
Grid fixed at $f = 30$ Hz	8	8	11	12	15	23

Convergence can be made independent of grid size

Frequency-domain:

- For one frequency, one shot: $n_{iter} = \mathcal{O}(1) \rightarrow \mathcal{O}(n^d)$ flops.
- For multiple frequencies and shots: $n_{freq}n_{shot}\mathcal{O}(n^d)$ flops.
- Assuming $\mathcal{O}(n_{freq}) = \mathcal{O}(n)$ and $\mathcal{O}(n_{shot}) = \mathcal{O}(n^{d-1}) \rightarrow \mathcal{O}(n^{2d})$

Time-domain:

- Simulation with n_t time levels: $n_t n_{shot} \mathcal{O}(n^d) = \mathcal{O}(n^{2d})$ ($n_t = \mathcal{O}(n)$).

Frequency- and time-domain wavefield computations are at the same order of computational complexity!

Frequency-domain wavefield simulation:

- conducive to frequency subsampling [Lin, Lebed, E., Herrmann, this conference]
 - use of $n_{freq} \ll n$ (or $n_{freq} = \mathcal{O}(1)$)
 - full wavefield recovered by (ℓ_1 minimization) sparsity-promoting program

Frequency-domain imaging:

- no “time history” – less memory requirement
in time-domain, check-pointing [Symes, 2008]
- conducive to frequency subsampling
 - [Plessix, Mulder, 2004], ...
 - (ℓ_1 minimization) sparsity-promoting program

Conducive to simultaneous shots simulation [Herrmann, E., Lin, 2008]

Refer to talk by Lin.

Conclusion

- key aspect of successful iterative methods for Helmholtz equation: tackling **indefiniteness** and **ill-conditioning**
- indefiniteness: use shifted-Laplacian (damped Helmholtz)
- ill-conditioning: use multilevel Krylov method

- combined, convergence less independent of frequencies and grid size

- paving the way to
 - compressive full wavefield computation (use of subsets of frequencies)
 - more rigorous compressive seismic imaging

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On Multilevel Krylov Methods:

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On Compressive Wavefield Simulations:

- Felix J. Herrmann, Y.A. Erlangga, Tim T.Y. Lin, Compressive simultaneous full-waveform simulation, submitted
<http://slim.eos.ubc.ca/Publications/Public/TechReports/herrmann08csf-r.pdf>