

- 2012.08-Present
  - Visiting student at Rice University
  - Advisor: Dr. William W. Symes
  - Project: EFWI, IVA, LSRTM, Sparse optimization
- 2010.09-Present
  - PhD candidate in geophysics at China University of Petroleum (Huadong)
  - Advisor: Dr. Zhenchun Li
  - Project: Wavefield inversion, Data regularization, Image domain denoising
- Awards
  - CSC Scholarship
  - National Scholarship
  - National Scholarship for Encouragement
  - CNPC Scholarship
  - National Excellent PhD Thesis Scholarship of CUP

# Linearized Extended Waveform Inversion and Inversion Velocity Analysis

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Annual Meeting 2013

# Overview

- 1 Background
- 2 Inversion Velocity Analysis
- 3 Numerical Tests
- 4 Summary and Future Plan

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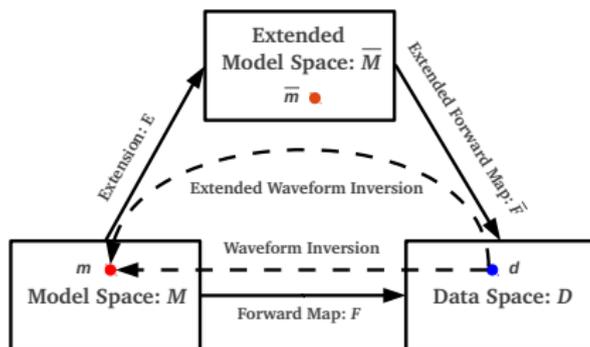
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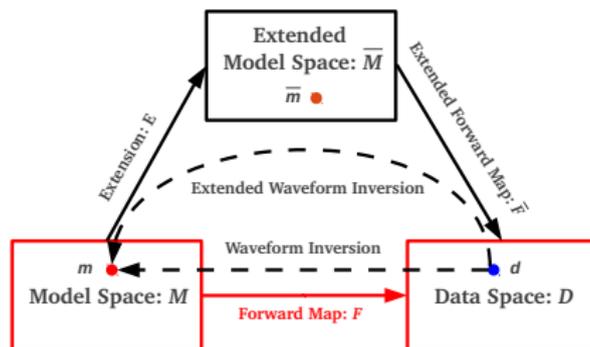
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  - Differential semblance optimization

# Waveform Inversion (WI)

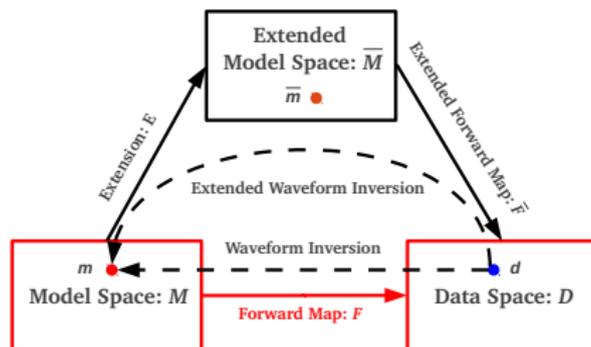


# Waveform Inversion (WI)



- Model space  $M := \{m(\mathbf{x}) = \frac{1}{v^2(\mathbf{x})}\}$ ; Data space  $D := \{d(\mathbf{x}_r, \mathbf{x}_s, \omega)\}$

# Waveform Inversion (WI)

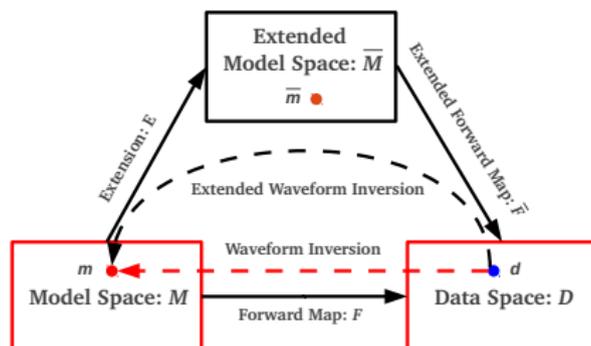


- Model space  $M := \{m(\mathbf{x}) = \frac{1}{v^2(\mathbf{x})}\}$ ; Data space  $D := \{d(\mathbf{x}_r, \mathbf{x}_s, \omega)\}$
- Forward map  $F$  in acoustic constant density medium

$$(\nabla^2 + \omega^2 m(\mathbf{x}))u(\mathbf{x}, \omega) = -f(\omega)\delta(\mathbf{x} - \mathbf{x}_s) \quad (1)$$

$$d(\mathbf{x}_r, \mathbf{x}_s, \omega) = S(\mathbf{x}_r, \mathbf{x}_s)u(\mathbf{x}, \omega) \quad (2)$$

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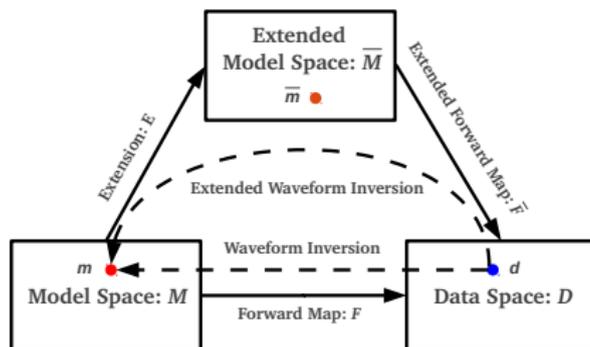
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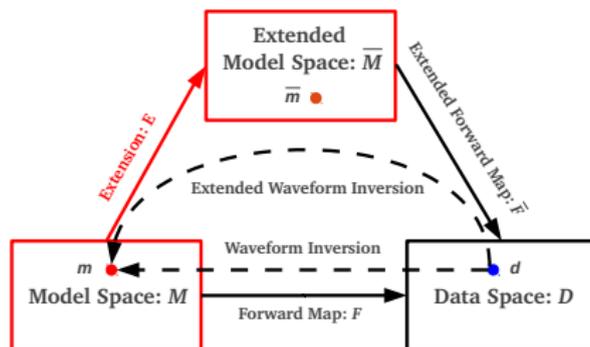
- Waveform inversion

$$\min_m J_{WI}[m, d] = \frac{1}{2} \|F[m] - d\|^2 \quad (3)$$

# Extended Waveform Inversion (EWI)

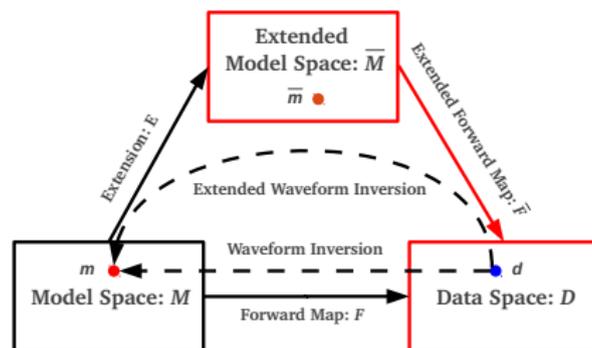


# Extended Waveform Inversion (EWI)



- Extended Model Space  $\bar{M} := \{ \bar{m}(\mathbf{x}, \mathbf{y}) = \frac{1}{v^2(\mathbf{x}, \mathbf{y})} \}$

# Extended Waveform Inversion (EWI)

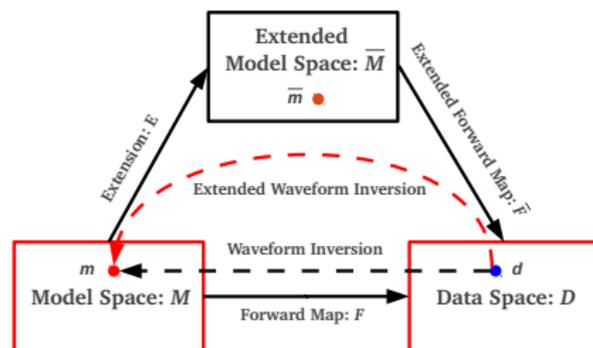


- Extended Model Space  $\bar{M} := \{\bar{m}(\mathbf{x}, \mathbf{y}) = \frac{1}{v^2(\mathbf{x}, \mathbf{y})}\}$
- Extended forward map  $\bar{F}$  in acoustic constant density medium

$$\nabla^2 u(\mathbf{x}, \omega) + \omega^2 \int d\mathbf{y} m(\mathbf{x}, \mathbf{y}) u(\mathbf{y}, \omega) = -f(\omega) \delta(\mathbf{x} - \mathbf{x}_s) \quad (4)$$

$$d(\mathbf{x}_r, \mathbf{x}_s, \omega) = S(\mathbf{x}_r, \mathbf{x}_s) u(\mathbf{x}, \omega) \quad (5)$$

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$$d(\mathbf{x}_r, \mathbf{x}_s, \omega) = S(\mathbf{x}_r, \mathbf{x}_s) u(\mathbf{x}, \omega) \quad (5)$$

- Extended waveform inversion

$$\min_{\bar{m}} J_{EWI}[\bar{m}, d] = \frac{1}{2} \| \bar{F}[\bar{m}] - d \|^2 + \frac{\sigma}{2} \| A[\bar{m}] \|^2 \quad (6)$$

# Linearized Extended Waveform Inversion (LEWI)

Problems of EWI: computational cost is extremely high!

Solution:

(1) Linearized approximation:

$$\bar{m} \simeq m_0 + \delta\bar{m}; \quad \bar{F}[\bar{m}] \simeq F[m_0] + D\bar{F}[m_0] * \delta\bar{m}$$

where  $D\bar{F}[m_0]$  is one order derivative of  $F$  to  $m$  at  $m_0$

(2) LEWI:

$$\min_{m_0, \delta\bar{m}} J_{LEWI}[m_0, \delta\bar{m}] = \frac{1}{2} \| D\bar{F}[m_0]\delta\bar{m} - (d - F[m_0]) \|^2 + \frac{\sigma}{2} \| A\delta\bar{m} \|^2$$

## Connection with LSM and MVA

when  $\sigma = 0$ , it limits to migration velocity analysis (MVA); when  $\sigma \rightarrow \infty$ , it limits to least-squares migration (LSM).

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# Linearized Extended Waveform Inversion Scheme

## LEWI:

$$\min_{m_0, \delta\bar{m}} J_{LEWI}[m_0, \delta\bar{m}] = \frac{1}{2} \| D\bar{F}[m_0]\delta\bar{m} - F_d \|^2 + \frac{\sigma}{2} \| A\delta\bar{m} \|^2$$

Solve the above problem with two level of loops:

- ① **Inner Loop:** Invert short scales (i.e. reflectivity) to get  $\delta\bar{m}_k[m_0]$

$$\min_{\delta\bar{m}} J_{LEWI}[m_0, \delta\bar{m}] = \frac{1}{2} \| D\bar{F}[m_0]\delta\bar{m} - F_d \|^2 + \frac{\sigma}{2} \| A\delta\bar{m} \|^2$$

- ② **Outer Loop:** Invert long scales (i.e. Background velocity)

$$\min_{m_0} J_{LEWI}[m_0, \delta\bar{m}_k[m_0]] = \frac{1}{2} \| D\bar{F}[m_0]\delta\bar{m}_k[m_0] - F_d \|^2 + \frac{\sigma}{2} \| A\delta\bar{m}_k[m_0] \|^2$$

# Gradient of short scales

The gradient of the objective function  $J_{DS}[m_0, \delta\bar{m}]$  with respect to  $\delta\bar{m}$ :

$$\nabla_{\delta\bar{m}} J_{DS}[m_0, \delta\bar{m}] = D\bar{F}^T[m_0](D\bar{F}[m_0]\delta\bar{m} - F_d) + \sigma A^T A \delta\bar{m} \quad (7)$$

Set the gradient to zero gives the normal equation, i.e.

$$(D\bar{F}^T[m_0]D\bar{F}[m_0] + \sigma A^T A)\delta\bar{m} = D\bar{F}^T[m_0]F_d \quad (8)$$

which can be re-written as:

$$N[m_0]\delta\bar{m} = M[m_0]F_d \quad (9)$$

where  $N[m_0]$  is normal operator and  $M[m_0]$  is migration operator.

# Gradient of long scales

The gradient of the objective function  $J_{DS}[m_0, \delta\bar{m}[m_0]]$  with respect to  $m_0$ :

$$\nabla_{m_0} J_{DS}[m_0, \delta\bar{m}_k[m_0]] = B[\delta\bar{m}_k, D\bar{F}[m_0]\delta\bar{m}_k - F_d] + B[P(N[m_0])e_k, F_d]$$

where  $B$  is bilinear operator,  $P(N[m_0])$  is a polynomial in the normal operator  $N[m_0]$ ,  $\delta\bar{m}_k$  is the inverted reflectivity and  $e_k$  is the normal equation error  $D_{\delta\bar{m}}J_{DS}[m_0, \delta\bar{m}]$ . The derivation can be found in [Liu, Symes; 2013].

## Notes:

*This formula is only justified when we use Chebyshev iteration to solve normal equation 9 in the case of depth-oriented model extension, but we can approximate it in different degree, migration velocity analysis is one of the approximations.*

# Inversion Velocity Analysis Scheme

## IVA:

$$\min_{m_0, \delta\bar{m}} J_{LEWI}[m_0, \delta\bar{m}] = \frac{1}{2} \| D\bar{F}[m_0]\delta\bar{m} - F_d \|^2 + \frac{\sigma}{2} \| A\delta\bar{m} \|^2$$

Compared with LEWI scheme, we solve the above problem separately:

- 1 **Inner Loop:** Invert short scales (i.e. reflectivity) to get  $\delta\bar{m}_k[m_0]$

$$\min_{\delta\bar{m}} J_{IVA}[m_0, \delta\bar{m}] = \frac{1}{2} \| D\bar{F}[m_0]\delta\bar{m} - F_d \|^2$$

- 2 **Outer Loop:** Invert long scales (i.e. Background velocity)

$$\min_{m_0} J_{IVA}[m_0, \delta\bar{m}_k[m_0]] = \frac{1}{2} \| A\delta\bar{m}_k[m_0] \|^2$$

# Gradient of short scales and long scales

- The gradient of the objective function  $J_{IVA}[m_0, \delta\bar{m}]$  with respect to  $\delta\bar{m}$ :

$$\nabla_{\delta\bar{m}} J_{IVA}[m_0, \delta\bar{m}] = D\bar{F}^T[m_0](D\bar{F}[m_0]\delta\bar{m} - F_d)$$

- The gradient of the objective function  $J_{IVA}[m_0, \delta\bar{m}[m_0]]$  with respect to  $m_0$ :

$$\nabla_{m_0} J_{IVA}[m_0, \delta\bar{m}_k[m_0]] = B[P(N[m_0])A^T A\delta\bar{m}_k, F_d]$$

## Notes:

*If the iteration number of inner loop is set to be zero and approximate  $P(N[m_0])$  to be identity matrix, the above formula is actually equivalent to the gradient given by WEMVA.*

- Extended linearized Born modeling  $d = D\bar{F}[m_0]\delta\bar{m}$ :

$$d(\mathbf{x}_r, \mathbf{x}_s, \omega) = -\omega^2 f(\omega) \int d\mathbf{x}d\mathbf{h} G(\mathbf{x}_r, \mathbf{x} + \mathbf{h}, \omega) \delta m(\mathbf{x}, \mathbf{h}) G(\mathbf{x} - \mathbf{h}, \mathbf{x}_s, \omega)$$

- Extended reverse time migration  $\delta\bar{m} = D\bar{F}^T[m_0]d$ :

$$\delta m(\mathbf{x}, \mathbf{h}) = - \int d\mathbf{x}_s d\mathbf{x}_r d\omega \omega^2 f^*(\omega) G^*(\mathbf{x}_s, \mathbf{x} - \mathbf{h}, \omega) G^*(\mathbf{x} + \mathbf{h}, \mathbf{x}_r, \omega) d(\mathbf{x}_r, \mathbf{x}_s, \omega)$$

- Bilinear operator  $\Delta m_0 = B[\delta\bar{m}, \Delta d]$ :

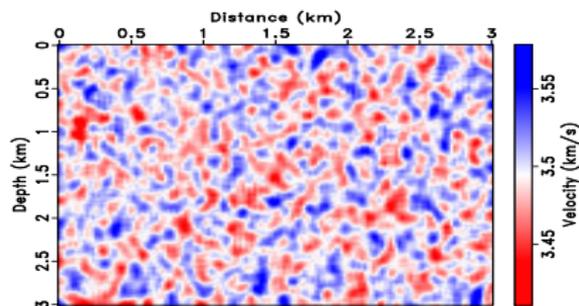
$$\Delta m_0(\mathbf{y})$$

$$= \int d\mathbf{x}_s d\mathbf{x}_r d\mathbf{x}d\mathbf{h}d\omega \left\{ G_0(\mathbf{y}, \mathbf{x}_s, \omega) \omega^4 f(\omega) \right\}^* \left\{ G_0^*(\mathbf{y}, \mathbf{x} - \mathbf{h}, \omega) \delta\bar{m}(\mathbf{x}, \mathbf{h}) G_0^*(\mathbf{x} + \mathbf{h}, \mathbf{x}_r, \omega) \Delta d(\mathbf{x}_r, \mathbf{x}_s, \omega) \right\}$$
$$+ \int d\mathbf{x}_s d\mathbf{x}_r d\mathbf{x}d\mathbf{h}d\omega \left\{ G_0(\mathbf{y}, \mathbf{x} + \mathbf{h}, \omega) \delta\bar{m}(\mathbf{x}, \mathbf{h}) G_0(\mathbf{x} - \mathbf{h}, \mathbf{x}_s, \omega) \omega^4 f(\omega) \right\}^* \left\{ G_0^*(\mathbf{y}, \mathbf{x}_r, \omega) \Delta d(\mathbf{x}_r, \mathbf{x}_s, \omega) \right\}$$

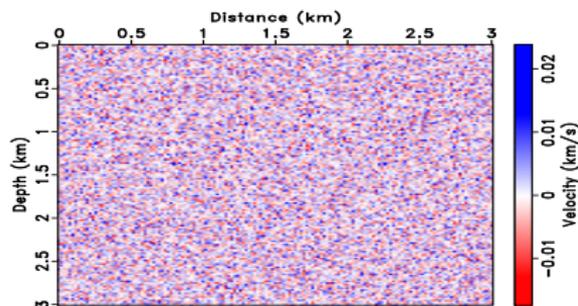
# Numerical Tests

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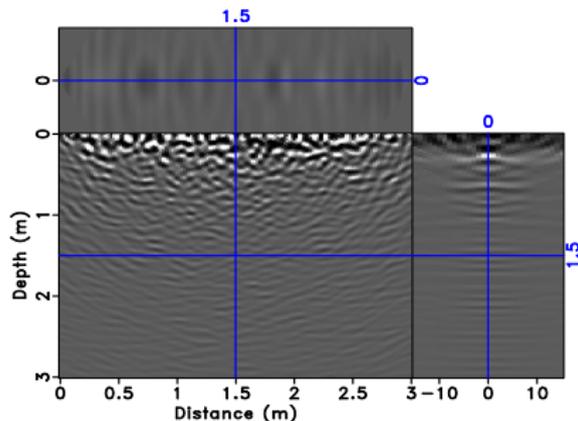
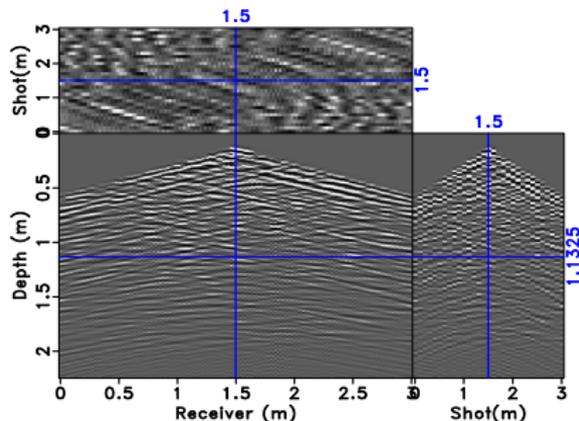
# Scan tests of LEWI Objective Function



Background velocity  
Synthetic data



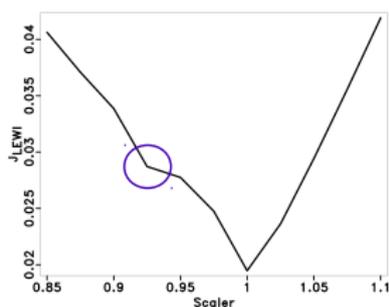
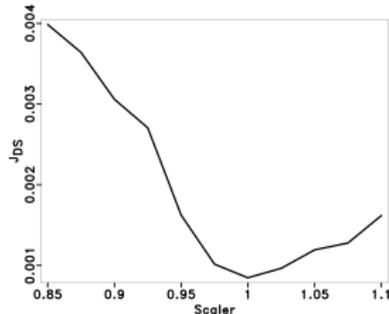
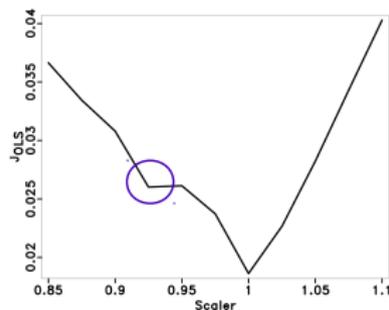
Reflectivity  
ERTM with correct velocity



# Scan tests of LEWI Objective Function

$$\begin{aligned} J_{LEWI}[m_0, \delta\bar{m}_k[m_0]] &= J_{OLS}[m_0, \delta\bar{m}_k[m_0]] + \sigma J_{DS}[m_0, \delta\bar{m}_k[m_0]] \\ &= \frac{1}{2} \| D\bar{F}[m_0]\delta\bar{m}_k[m_0] - F_d \|^2 + \frac{\sigma}{2} \| A\delta\bar{m}_k[m_0] \|^2 \end{aligned}$$

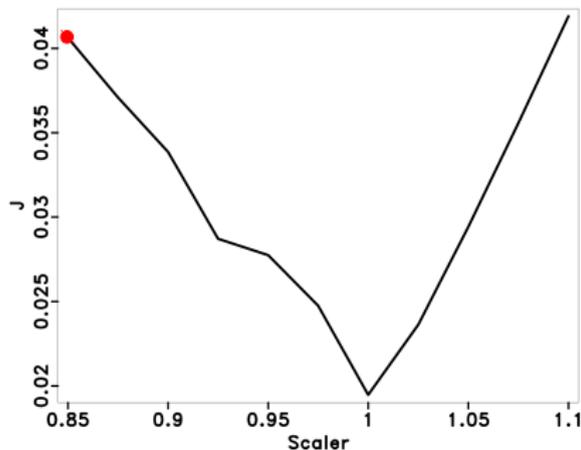
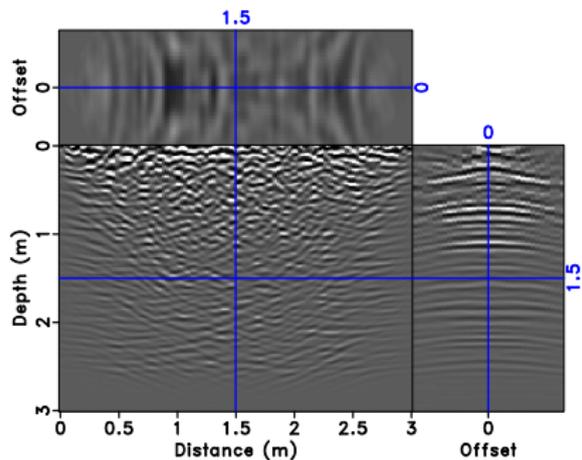
- Scan  $J_{OLS}[m_0, \delta\bar{m}_k[m_0]]$  along  $m_0 = \mu m_0^*$ ,  $\mu \in [0.85, 1.1]$
- Scan  $J_{DS}[m_0, \delta\bar{m}_k[m_0]]$  along  $m_0 = \mu m_0^*$ ,  $\mu \in [0.85, 1.1]$
- Scan  $J_{LEWI}[m_0, \delta\bar{m}_k[m_0]]$  along  $m_0 = \mu m_0^*$ ,  $\mu \in [0.85, 1.1]$



# Scan tests of LEWI Objective Function

$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 0.850m_0^*$$

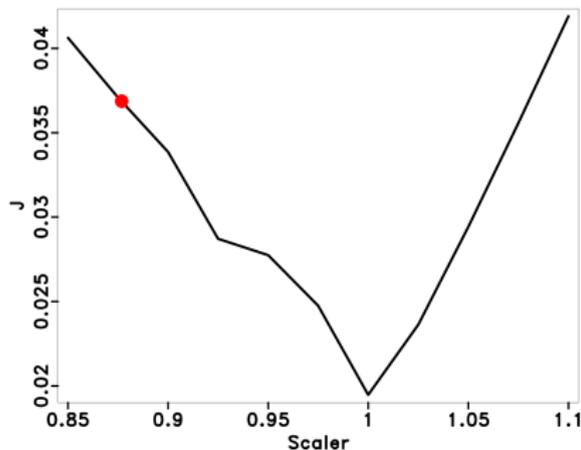
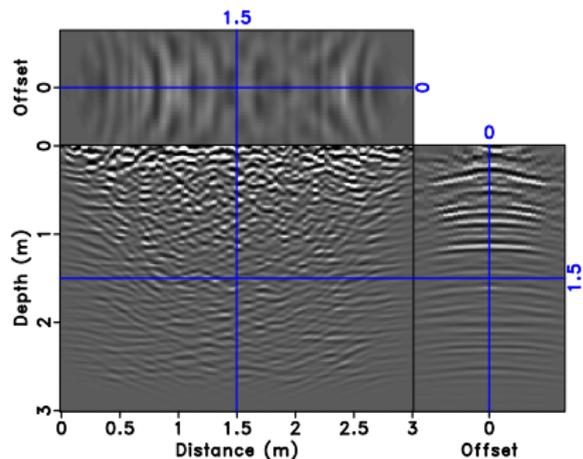
LSERTM with Chebyshev iteration



# Scan tests of LEWI Objective Function

$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 0.875m_0^*$$

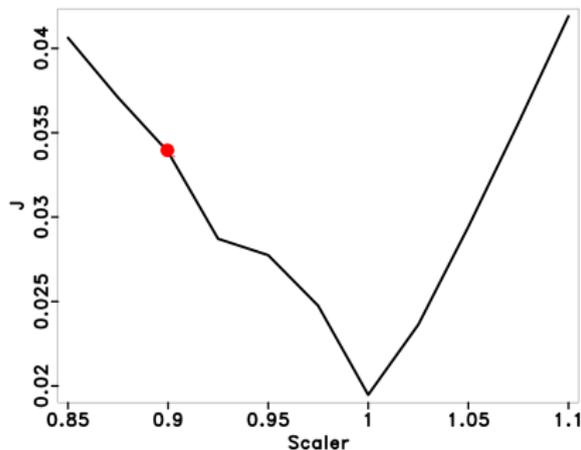
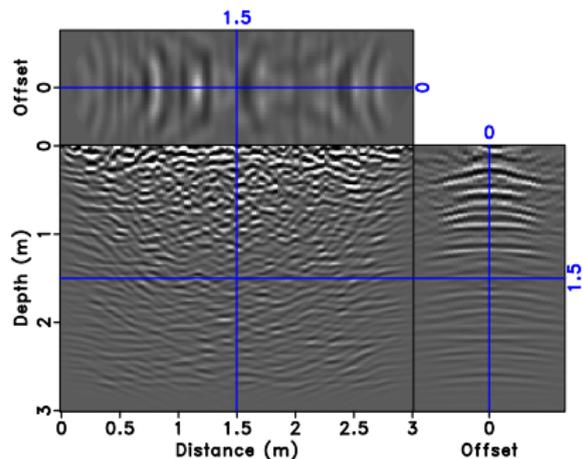
LSERTM with Chebyshev iteration



# Scan tests of LEWI Objective Function

$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 0.900m_0^*$$

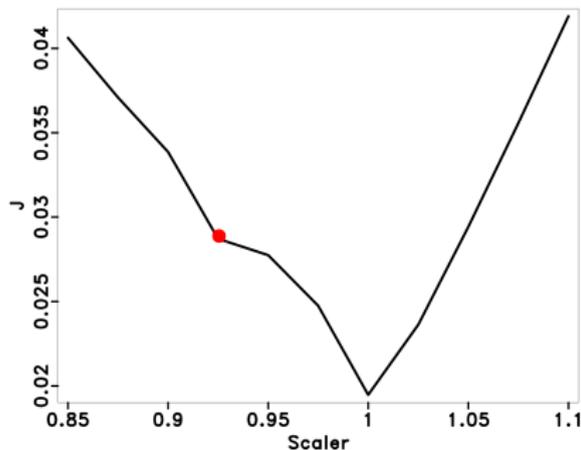
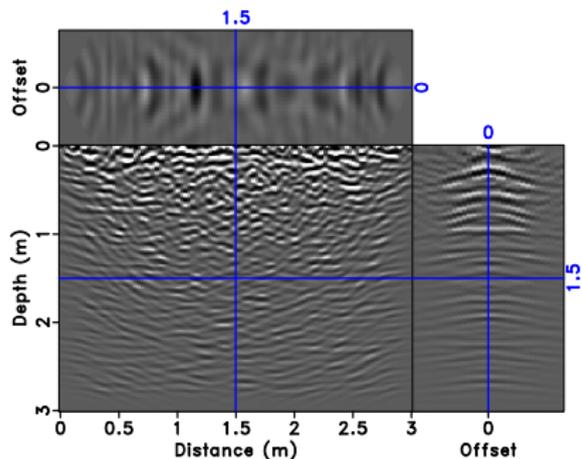
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$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 0.925m_0^*$$

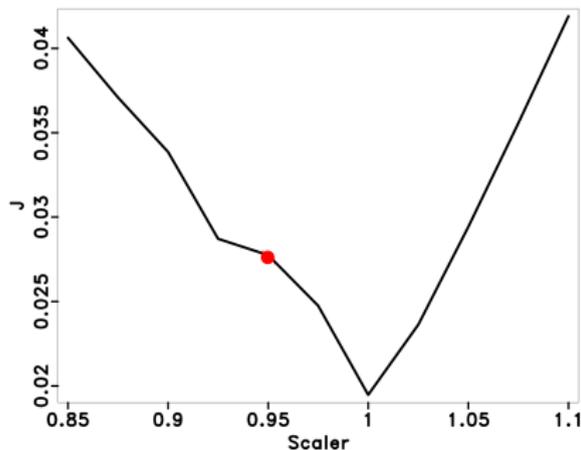
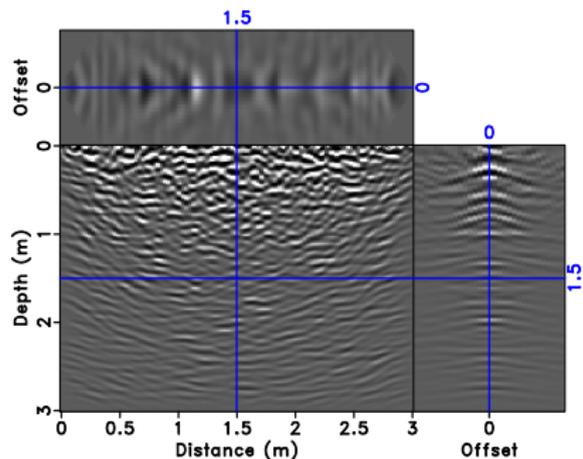
LSERTM with Chebyshev iteration



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$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 0.950m_0^*$$

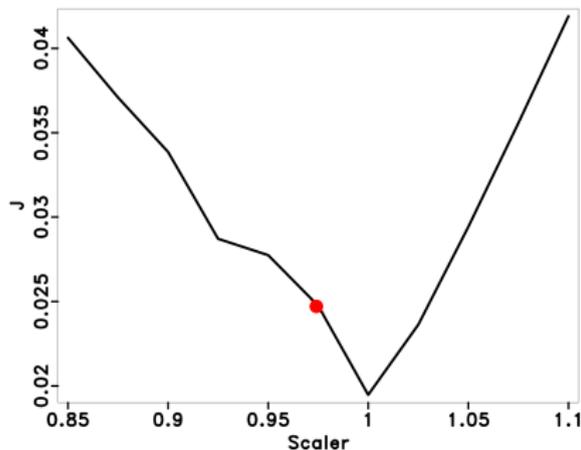
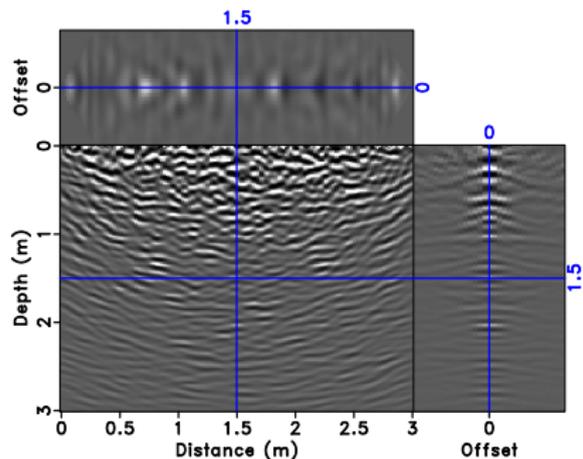
LSERTM with Chebyshev iteration



# Scan tests of LEWI Objective Function

$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 0.975m_0^*$$

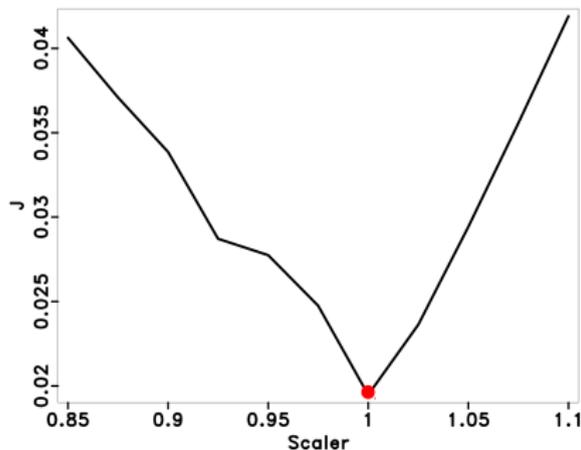
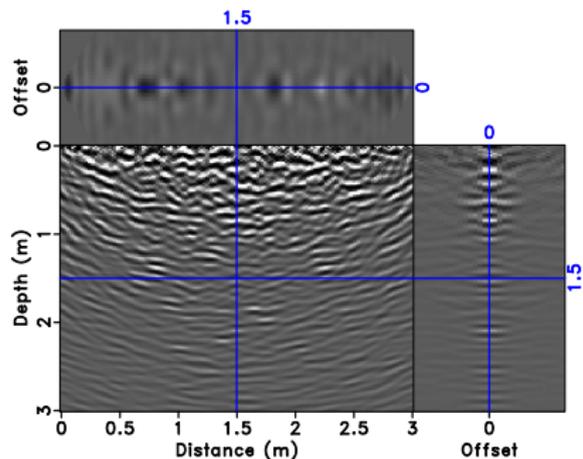
LSERTM with Chebyshev iteration



# Scan tests of LEWI Objective Function

$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 1.000m_0^*$$

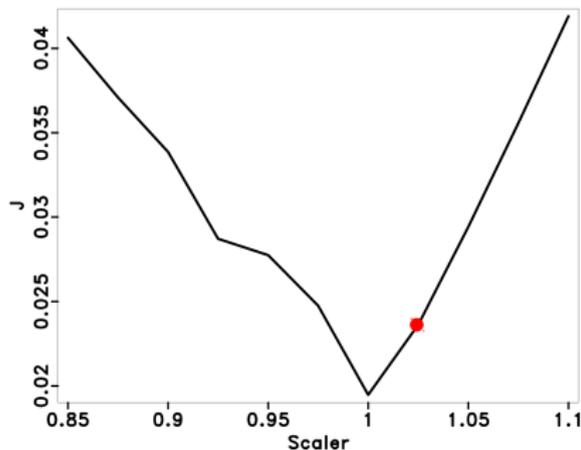
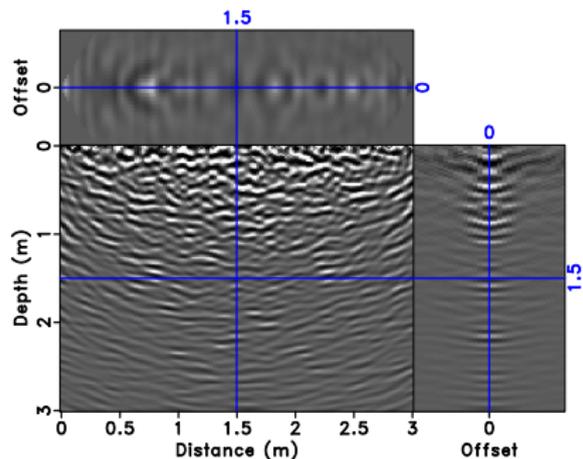
LSERTM with Chebyshev iteration



# Scan tests of LEWI Objective Function

$$J_{LEWI}[m_0, \delta\bar{m}_k[m_0]] \text{ at } m_0 = 1.025m_0^*$$

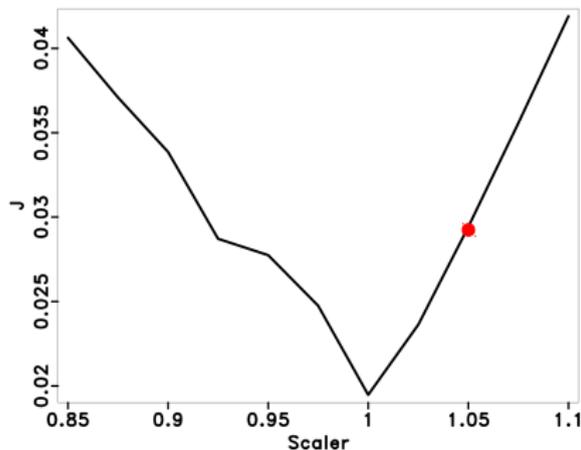
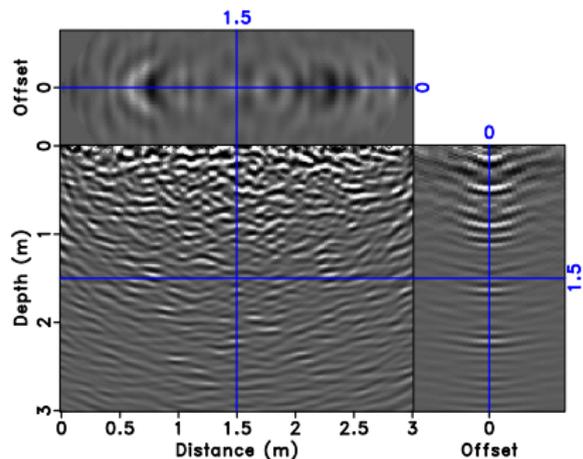
LSERTM with Chebyshev iteration



# Scan tests of LEWI Objective Function

$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 1.050m_0^*$$

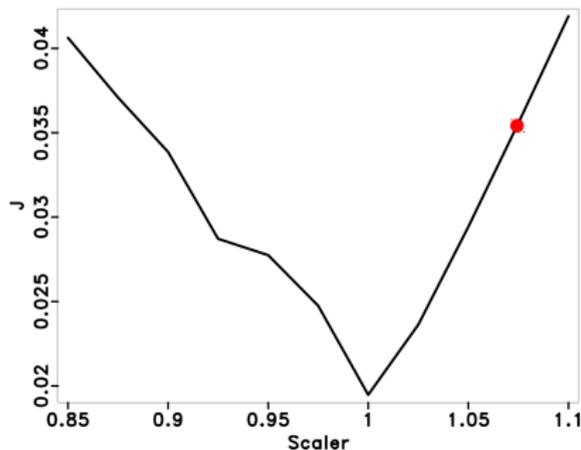
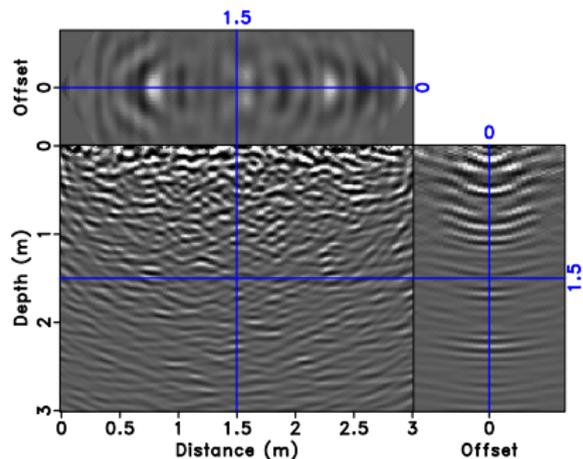
LSERTM with Chebyshev iteration



# Scan tests of LEWI Objective Function

$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 1.075m_0^*$$

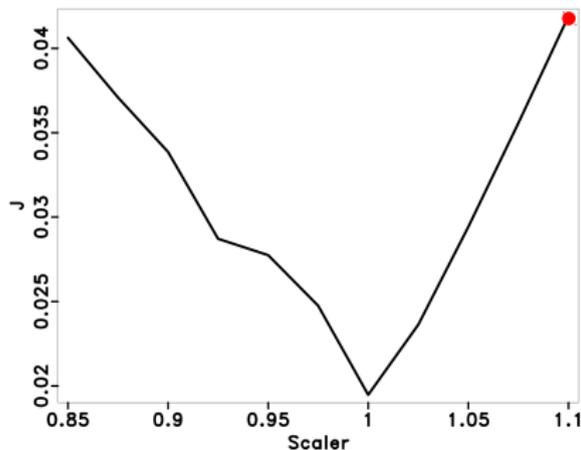
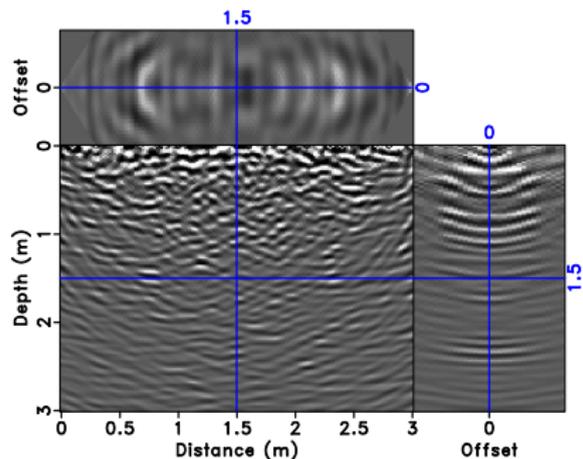
LSERTM with Chebyshev iteration



# Scan tests of LEWI Objective Function

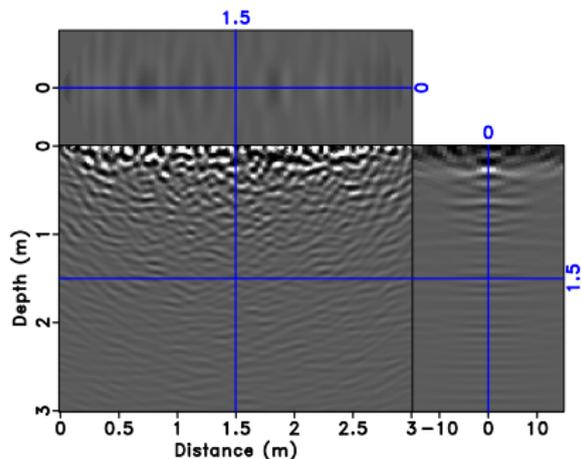
$$J_{LEWI}[m_0, \delta \bar{m}_k[m_0]] \text{ at } m_0 = 1.100m_0^*$$

LSERTM with Chebyshev iteration

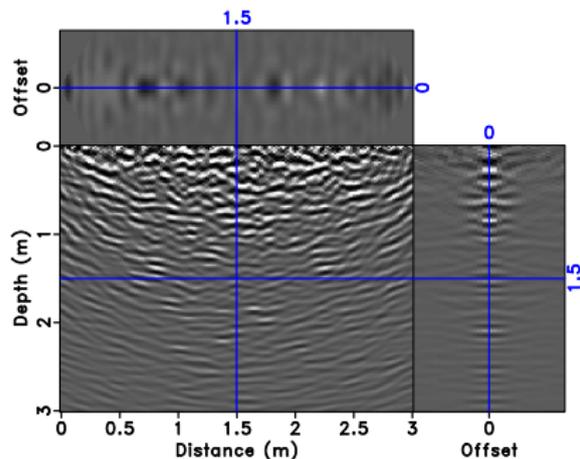


# ERTM vs LSERTM

ERTM with correct velocity



LSERTM with Chebyshev iteration



# LEWI tests: Gaussian Model

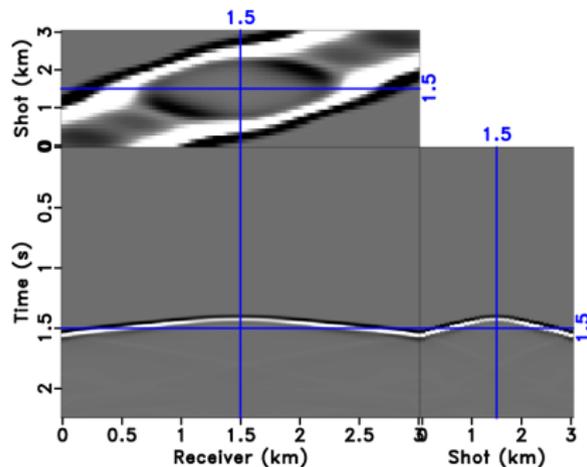
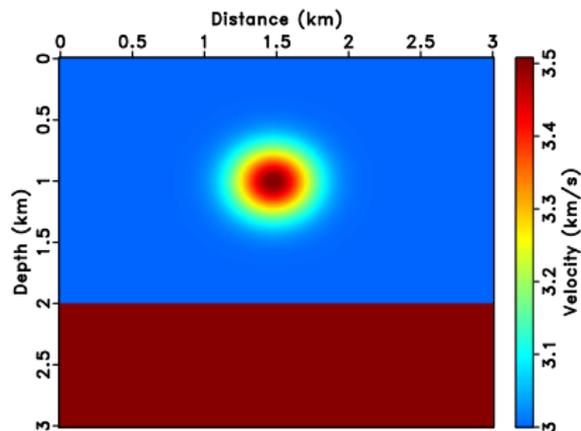
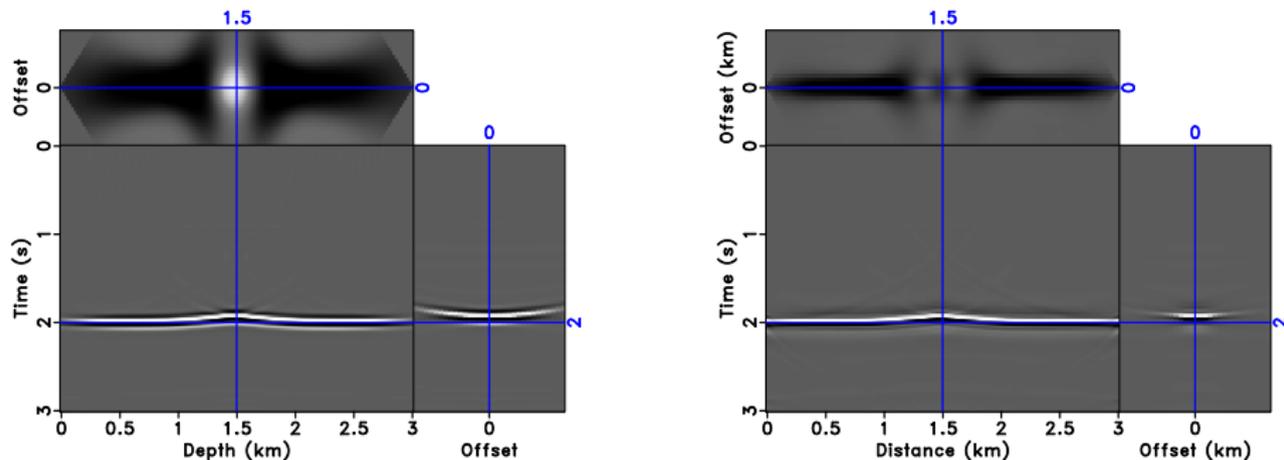


Figure: (a) Gaussian velocity model; (b) Synthetic data.

# LEWI tests: Gaussian Model



**Figure:** (a) Reflectivity with extended reverse-time migration; (b) Inverted reflectivity with Chebyshev iteration method.

# LEWI tests: Gaussian Model

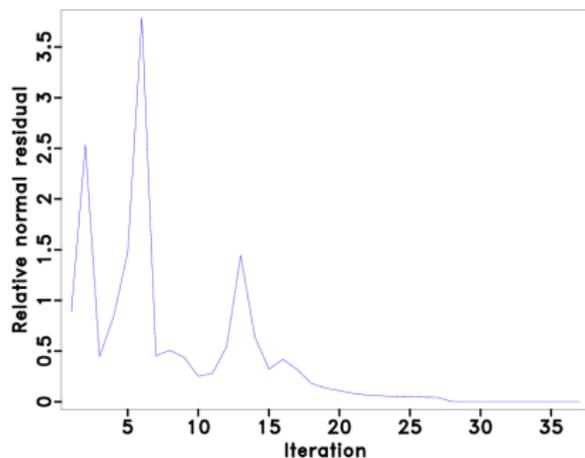
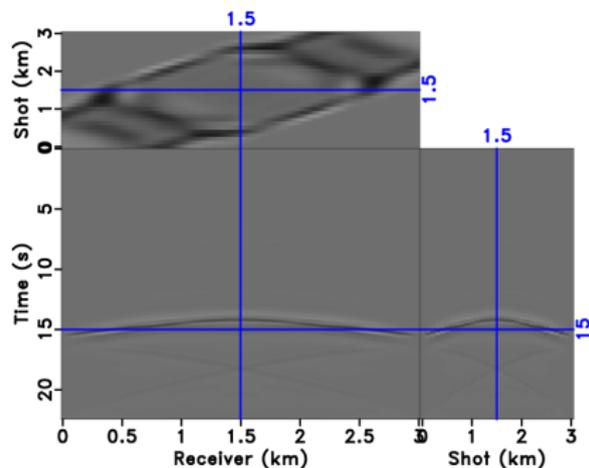


Figure: (a) Data misfit residual of Chebyshev iteration method; (b) Relative normal residual curve.

# LEWI tests: Gaussian Model

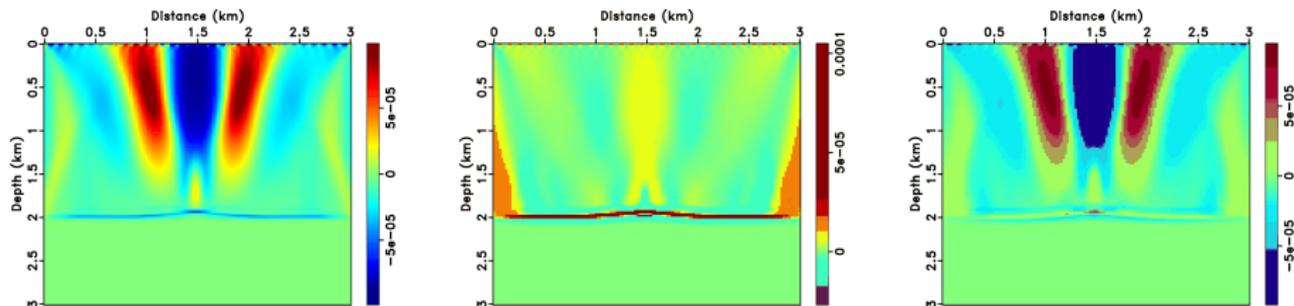
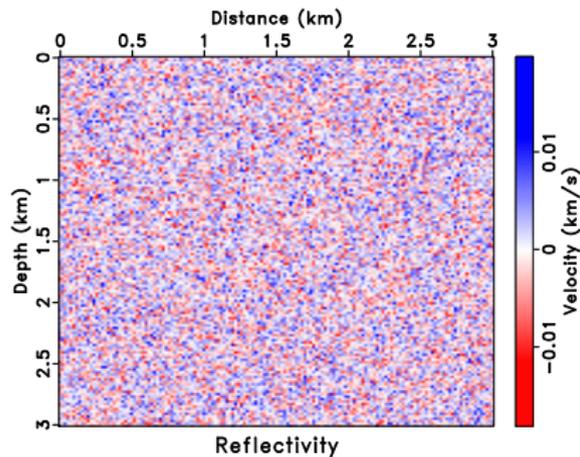
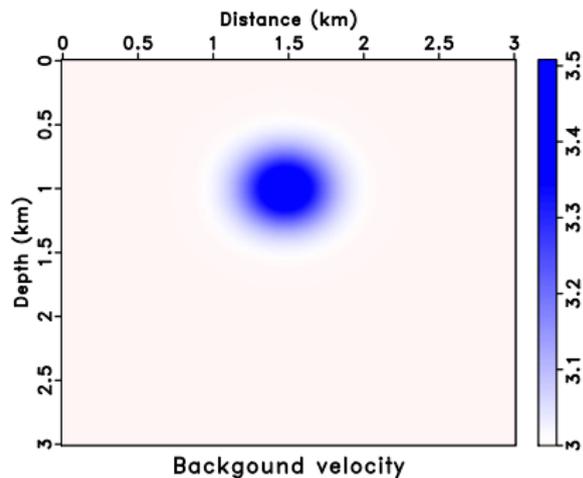


Figure: LEWI gradient of long scales (a) the first term; (b) the second term; (c) total.

# IVA vs MVA: Gaussian-Random Model Tests



# IVA vs MVA: Gaussian-Random Model Tests

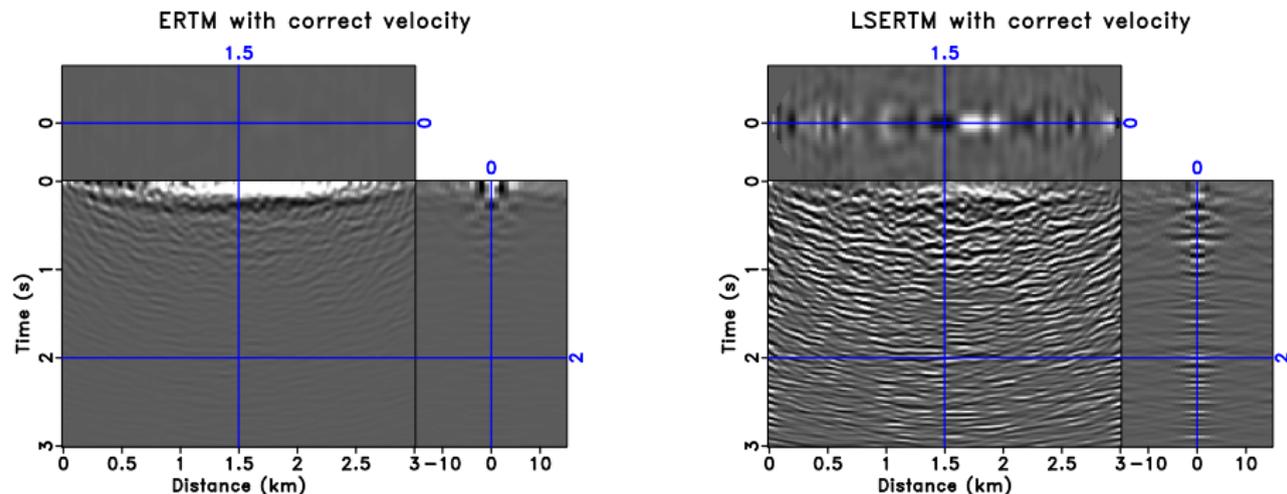


Figure: (a) Extended reverse-time migration; (b) Least-squares extended reverse-time migration when background velocity is correct.

# IVA vs MVA: Gaussian-Random Model Tests

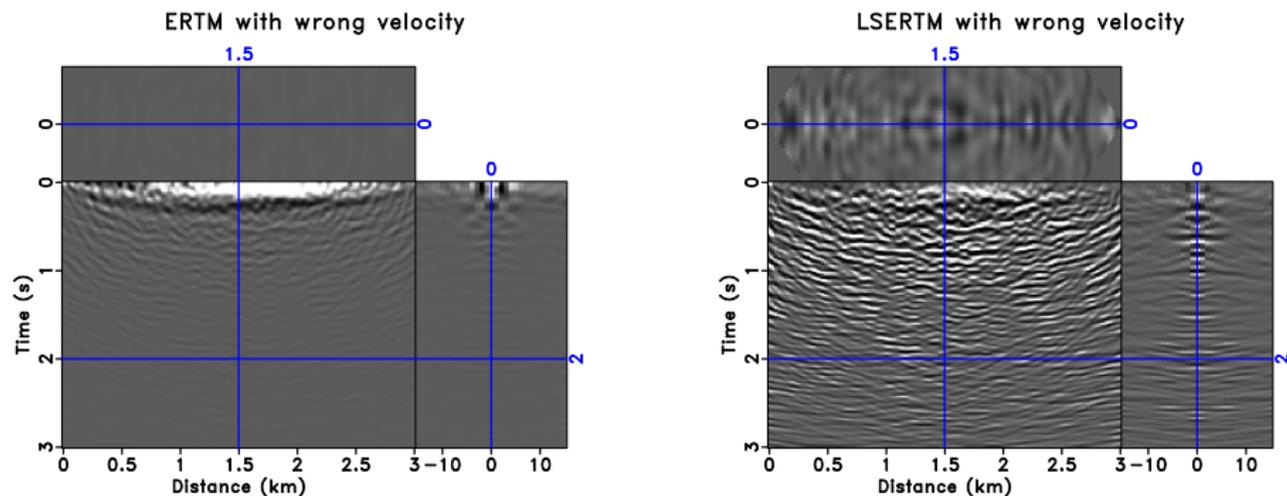


Figure: (a) Extended reverse-time migration; (b) Least-squares extended reverse-time migration when background velocity is 3 km/s.

# IVA vs MVA: Gaussian-Random Model Tests

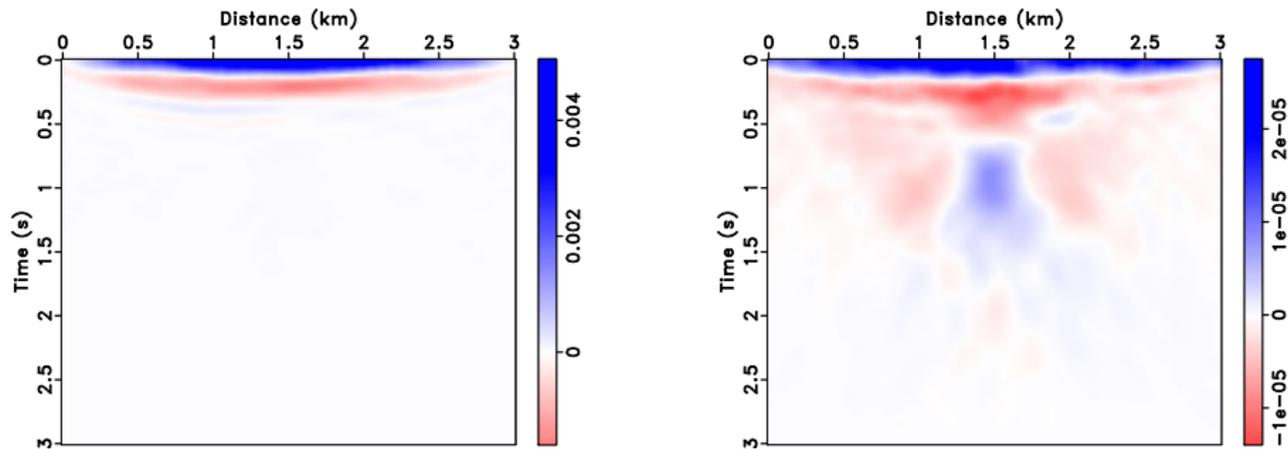


Figure: Gradient computed by (a) extended reverse-time migration; (b) least-squares extended reverse-time migration.

- 1 Background
- 2 Inversion Velocity Analysis
- 3 Numerical Tests
- 4 Summary and Future Plan

# Summary and Future Plan

## Conclusion

- The OLS objective function has local minimum problem, which can be solved with the idea of differential semblance optimization.
- The inverted reflectivity has higher resolution and more balanced amplitude, which is also crucial in background velocity inversion.
- As the inverted reflectivity image, instead of prestack migration approximation, is used to adjust velocity model, IVA is more accurate than MVA.

## Future plan

- Improve the efficiency of IVA  $\Rightarrow$  Preconditioning, Compressive sensing.
- Extend to non-linear case  $\Rightarrow$  Plane-wave domain, depth-oriented extension.

# References

 William W. Symes (2008)  
Migration velocity analysis and waveform inversion  
*Geophysical Prospecting* 56(6), 765–790.

 William W. Symes (2009)  
The seismic reflection inverse problem  
*Inverse problems* 25(12), 123008.

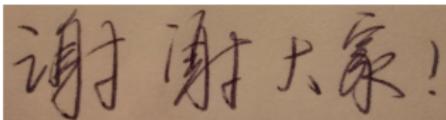
 Yujin Liu, William W. Symes, Yin Huang, Zhenchun Li (2013)  
Inversion Velocity Analysis via Differential Semblance Optimization in the  
Depth-oriented Extension  
*83rd SEG Annual International Meeting, Expanded Abstract*, submitted

 William W. Symes and Kern, Michel (1994)  
Inversion of reflection seismograms by differential semblance analysis: algorithm  
structure and synthetic examples1  
*Geophysical Prospecting* 42(6), 565–614.

# Acknowledgements

- Thank CSC for supporting my visit to TRIP!
- Thank TRIP for hosting me!
- Thank the sponsors of TRIP for their support!

Thank you!

A rectangular image with a brown, textured background. It contains the Chinese characters '谢谢大家!' (Xièxiè dàjiā!) written in a dark brown, cursive calligraphic style. The characters are arranged horizontally and read from left to right.