

Time-domain Wavefield Reconstruction Inversion for large-scale seismic inversion

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Agenda

- Motivations:
robustness and computation
- Theory:
FWI vs WRI vs WRI*
- Numerical experiments:
acoustic, TTI, small 3D
- Discussion and conclusions

Full Waveform Inversion

$$\min_{\mathbf{m}} \mathcal{J}_{\text{FWI}}(\mathbf{m}) = \frac{1}{2} \|F(\mathbf{m})\mathbf{q} - \mathbf{d}\|^2, \quad F(\mathbf{m}) = RA(\mathbf{m})^{-1}$$

[Tarantola, A., '84; Haber, E., et al, 2000; Epanomeritakis, I., et al, 2008]

- ✓ **3D** computations are affordable
via large HPC systems (or cloud computing) [Witte et al., 2019]
- ✗ (Effectively) **multimodal** problem:
it needs a good starting model!

Wavefield Reconstruction Inversion

$$\min_{\mathbf{m}} \mathcal{J}_{\text{WRI}}(\mathbf{m}, \bar{\mathbf{u}}(\mathbf{m})) = \frac{1}{2} \|R\bar{\mathbf{u}} - \mathbf{d}\|^2 + \frac{\lambda^2}{2} \|A(\mathbf{m})\bar{\mathbf{u}} - \mathbf{q}\|^2$$

[van den Berg, P. M., and Kleinman, R. E., 1997; van Leeuwen, T. and Herrmann, F. J., 2013]

✓ **Better conditioning**

Wavefield Reconstruction Inversion

$$\min_{\mathbf{m}} \mathcal{J}_{\text{WRI}}(\mathbf{m}, \bar{\mathbf{u}}(\mathbf{m})) = \frac{1}{2} \|\mathbf{R}\bar{\mathbf{u}} - \mathbf{d}\|^2 + \frac{\lambda^2}{2} \|A(\mathbf{m})\bar{\mathbf{u}} - \mathbf{q}\|^2$$

[van den Berg, P. M., and K

$$\begin{bmatrix} R \\ \lambda A(\mathbf{m}) \end{bmatrix} \bar{\mathbf{u}} = \begin{bmatrix} \mathbf{d} \\ \lambda \mathbf{q} \end{bmatrix}$$

and Herrmann, F. J., 2013]

augmented wave equation

✓ **Better conditioning**

✗ **Augmented solver: hard to scale to 3D**
(explicit time-marching schemes?)

Prior art: extended-source formulation

$$\min_{\mathbf{m}} \mathcal{J}_{\text{WRI}}(\mathbf{m}, \bar{\mathbf{u}}) = \frac{1}{2} \|\mathbf{R}\bar{\mathbf{u}} - \mathbf{d}\|^2 + \frac{\lambda^2}{2} \|\mathbf{A}(\mathbf{m})\bar{\mathbf{u}} - \mathbf{q}\|^2$$

[van den Berg, P. M., and Kleinman, R. E., 1997; van Leeuwen, T. and Herrmann, F. J., 2013]

=

$$\min_{\mathbf{m}} \mathcal{J}_{\text{WRI}-\mathbf{q}}(\mathbf{m}, \bar{\mathbf{q}}) = \frac{1}{2} \|\mathbf{F}(\mathbf{m})\bar{\mathbf{q}} - \mathbf{d}\|^2 + \frac{\lambda^2}{2} \|\bar{\mathbf{q}} - \mathbf{q}\|^2$$

[Wang et al, 2016, Huang et al, 2018]

Prior art: augmented state approximation

$$\min_{\mathbf{m}} \mathcal{J}_{\text{WRI}-\mathbf{q}}(\mathbf{m}, \bar{\mathbf{q}}) = \frac{1}{2} \|F(\mathbf{m})\bar{\mathbf{q}} - \mathbf{d}\|^2 + \frac{\lambda^2}{2} \|\bar{\mathbf{q}} - \mathbf{q}\|^2$$

[Wang et al, 2016]

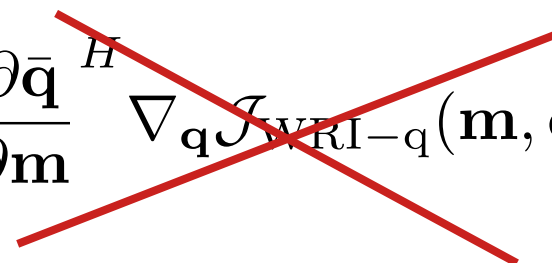
$$\begin{bmatrix} F(\mathbf{m}) \\ \lambda I \end{bmatrix} \bar{\mathbf{q}} = \begin{bmatrix} \mathbf{d} \\ \lambda \mathbf{q} \end{bmatrix} \implies \bar{\mathbf{q}} \approx \mathbf{q} + F(\mathbf{m})^H (\mathbf{d} - F(\mathbf{m})\mathbf{q}) / \lambda^2$$

✓ (Approximated) augmented solver: scale to **3D**

Prior art: augmented state approximation

$$\min_{\mathbf{m}} \mathcal{J}_{\text{WRI}-q}(\mathbf{m}, \bar{\mathbf{q}}) = \frac{1}{2} \|F(\mathbf{m})\bar{\mathbf{q}} - \mathbf{d}\|^2 + \frac{\lambda^2}{2} \|\bar{\mathbf{q}} - \mathbf{q}\|^2$$

[Wang et al, 2016]

$$\nabla_{\mathbf{m}} \mathcal{J}_{\text{WRI}-q} = \nabla_{\mathbf{m}} \mathcal{J}_{\text{WRI}-q}(\mathbf{m}, \bar{\mathbf{q}}) + \frac{\partial \bar{\mathbf{q}}}{\partial \mathbf{m}} \nabla_{\mathbf{q}} \mathcal{J}_{\text{WRI}-q}(\mathbf{m}, \bar{\mathbf{q}})$$


✘ Gradient computation inconsistency

WRI*: denoising reformulation

$$\min_{\mathbf{m}, \mathbf{u}} \frac{1}{2} \|A(\mathbf{m})\mathbf{u} - \mathbf{q}\|^2 \quad \text{s.t.} \quad \|R\mathbf{u} - \mathbf{d}\| \leq \varepsilon$$

[Wang, R., and Herrmann, F. J., 2017]

WRI*: Lagrangian

$$\min_{\mathbf{m}, \mathbf{u}} \frac{1}{2} \|A(\mathbf{m})\mathbf{u} - \mathbf{q}\|^2 \quad \text{s.t.} \quad \|R\mathbf{u} - \mathbf{d}\| \leq \varepsilon$$

[Wang, R., and Herrmann, F. J., 2017]

$$\max_{\mathbf{y}} \min_{\mathbf{m}, \mathbf{u}} \mathcal{L}(\mathbf{m}, \mathbf{u}, \mathbf{y}) = \frac{1}{2} \|A(\mathbf{m})\mathbf{u} - \mathbf{q}\|^2 + \mathbf{y} \cdot (R\mathbf{u} - \mathbf{d}) - \varepsilon \|\mathbf{y}\|$$

WRI*: Lagrangian

$$\min_{\mathbf{m}, \mathbf{u}} \frac{1}{2} \|A(\mathbf{m})\mathbf{u} - \mathbf{q}\|^2 \quad \text{s.t.} \quad \|R\mathbf{u} - \mathbf{d}\| \leq \varepsilon$$

[Wang, R., and Herrmann, F. J., 2017]

$$\max_{\mathbf{y}} \min_{\mathbf{m}} \bar{\mathcal{L}}(\mathbf{m}, \mathbf{y}) = \frac{1}{2} \|A(\mathbf{m})\bar{\mathbf{u}} - \mathbf{q}\|^2 + \mathbf{y} \cdot (R\bar{\mathbf{u}} - \mathbf{d}) - \varepsilon \|\mathbf{y}\|$$

$$A(\mathbf{m})\bar{\mathbf{u}} = \mathbf{q} + F(\mathbf{m})^H \mathbf{y}$$

WRI*: augmented state approximation

$$\min_{\mathbf{m}, \mathbf{u}} \frac{1}{2} \|A(\mathbf{m})\mathbf{u} - \mathbf{q}\|^2 \quad \text{s.t.} \quad \|R\mathbf{u} - \mathbf{d}\| \leq \varepsilon$$

[Wang, R., and Herrmann, F. J., 2017]

$$\min_{\mathbf{m}} \tilde{\mathcal{L}}(\mathbf{m}) = \frac{1}{2} \|A(\mathbf{m})\bar{\mathbf{u}} - \mathbf{q}\|^2 + \tilde{\mathbf{y}} \cdot (R\bar{\mathbf{u}} - \mathbf{d}) - \varepsilon \|\tilde{\mathbf{y}}\|$$

$$\tilde{\mathbf{y}} \propto F(\mathbf{m})\mathbf{q} - \mathbf{d}$$

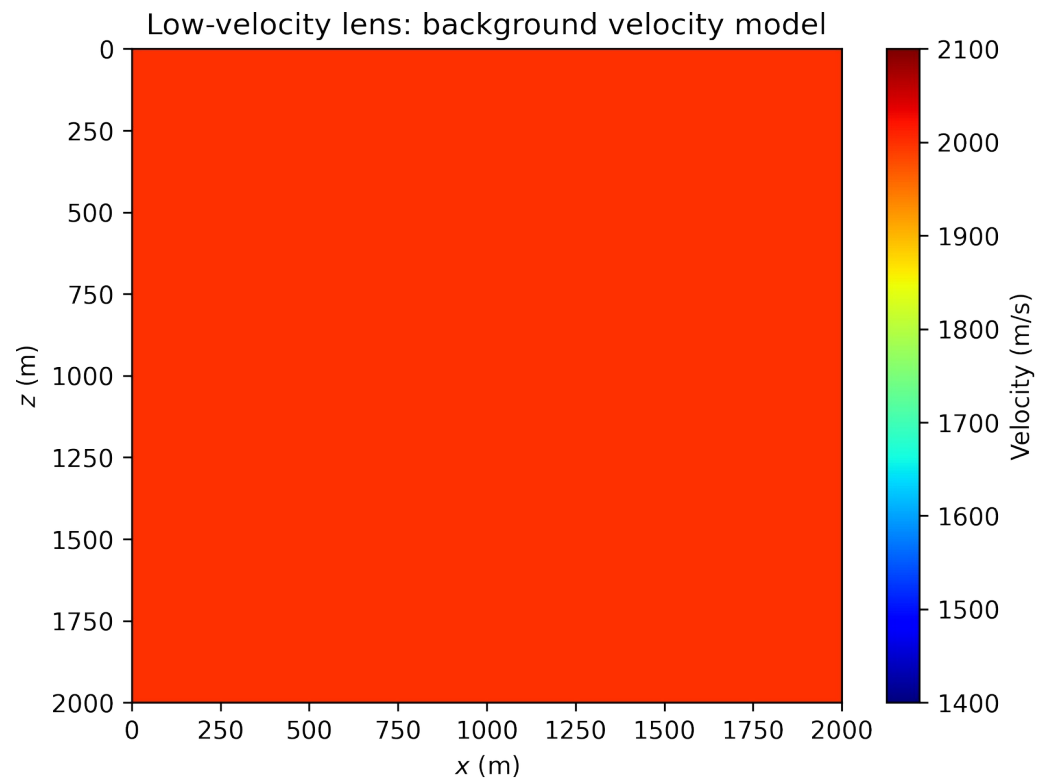
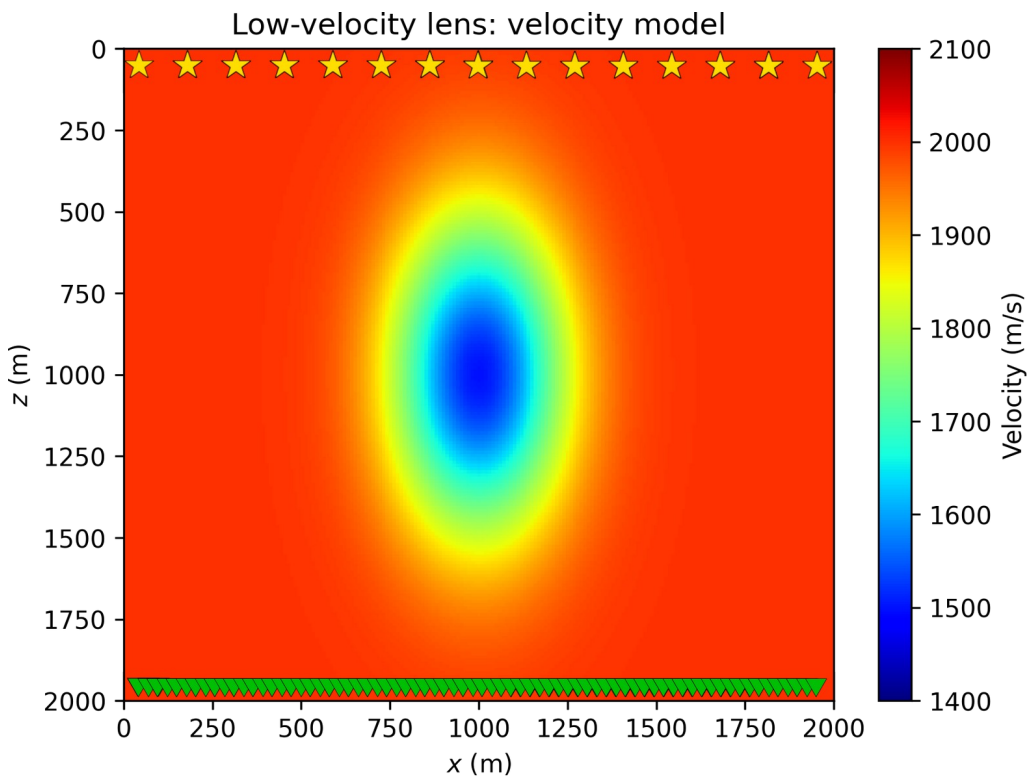
can be differentiated through...

WRI*: reduced formulation

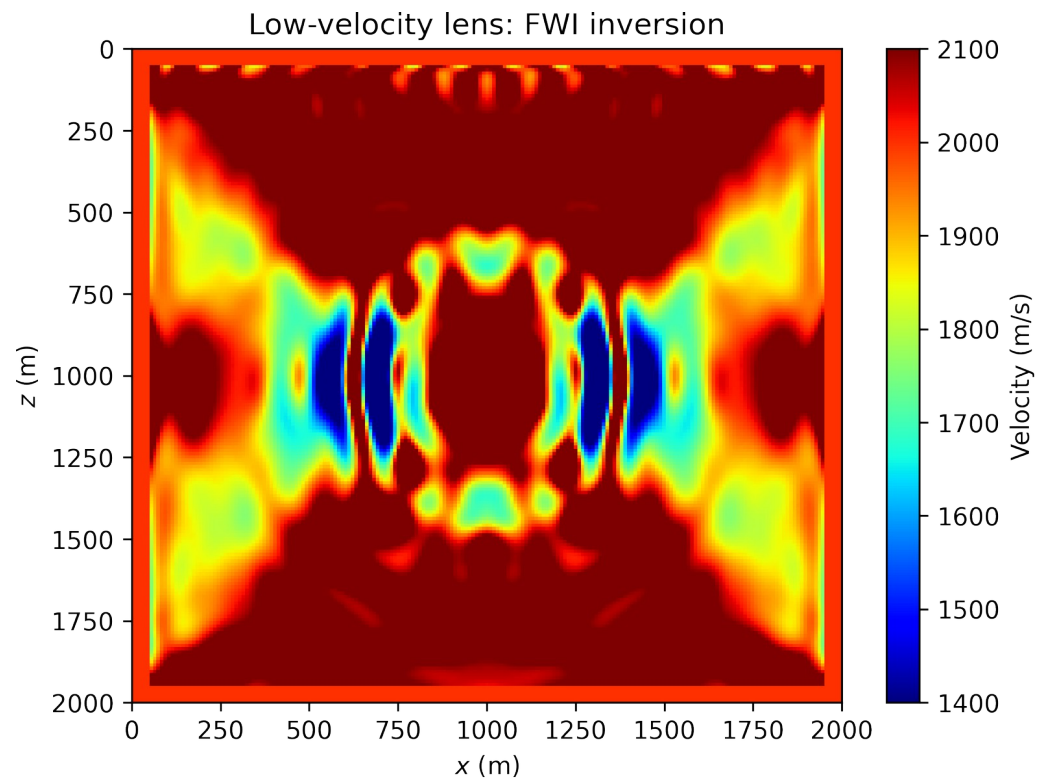
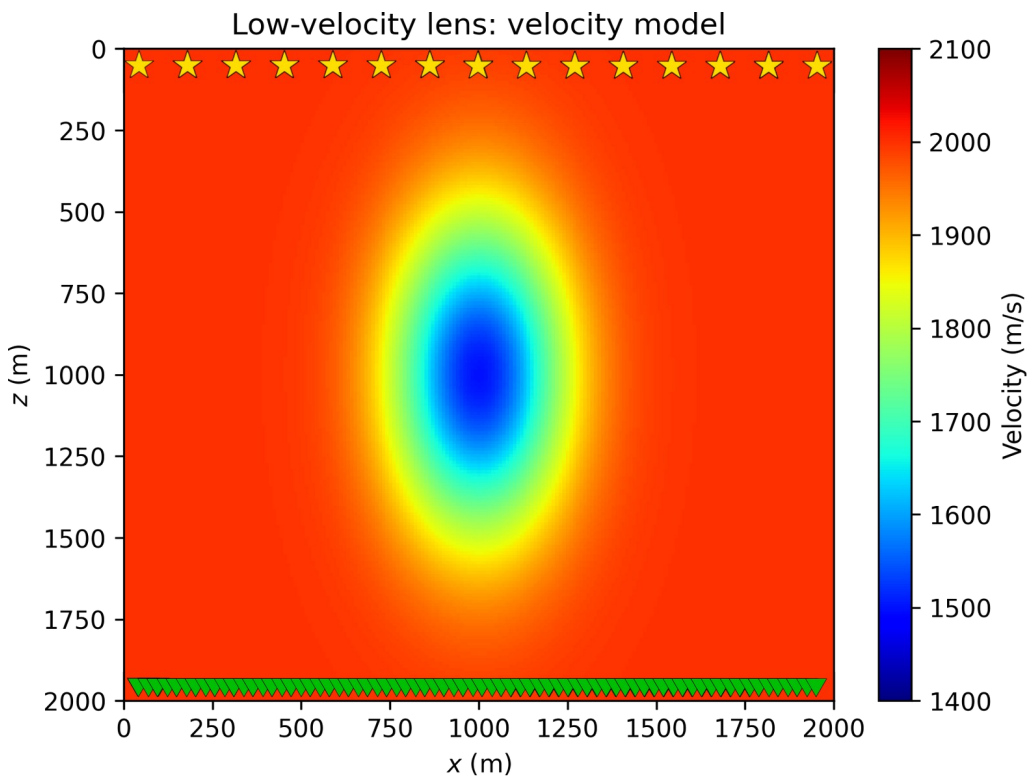
$$\min_{\mathbf{m}} \tilde{\mathcal{L}}(\mathbf{m}) = \frac{1}{2} \|A(\mathbf{m})\bar{\mathbf{u}} - \mathbf{q}\|^2 + \tilde{\mathbf{y}} \cdot (R\bar{\mathbf{u}} - \mathbf{d}) - \varepsilon \|\tilde{\mathbf{y}}\|$$

- ✓ Requires standard wave equation solver: scale to 3D
- ✓ No gradient computation inconsistency
- ✓ Retains WRI robustness
- ✓ Roughly equivalent to 2x the computational cost of FWI
- ✗ Roughly equivalent to 2x the computational cost of FWI
- ✗ Model resolution generally inferior to WRI

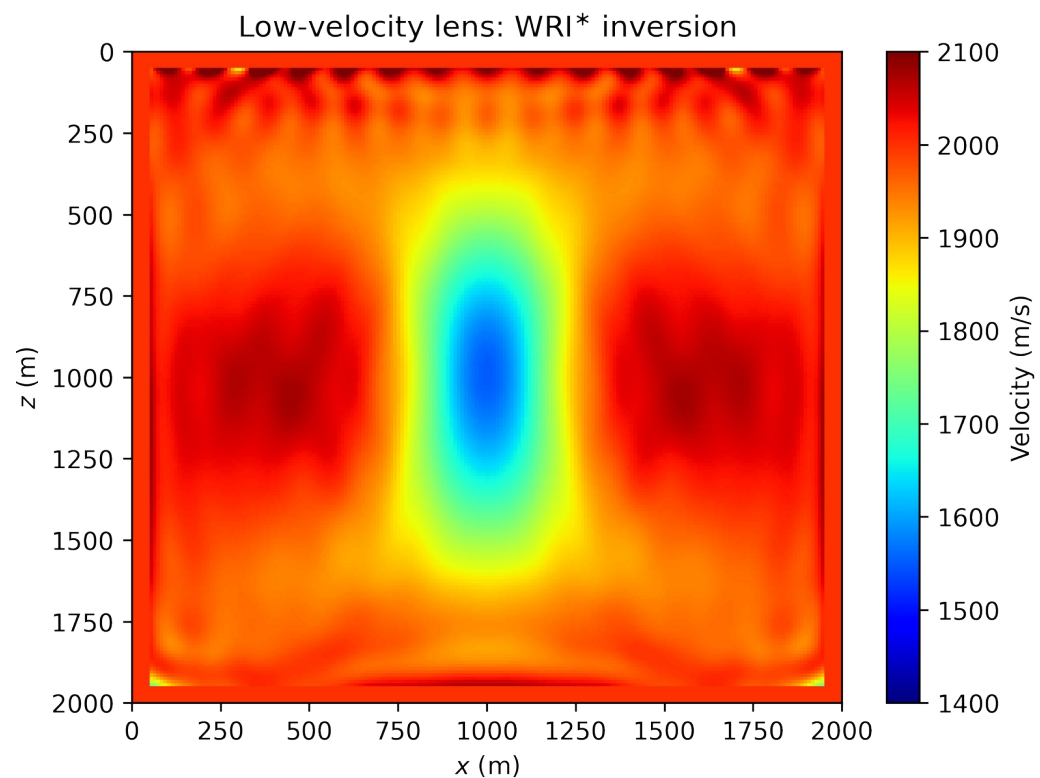
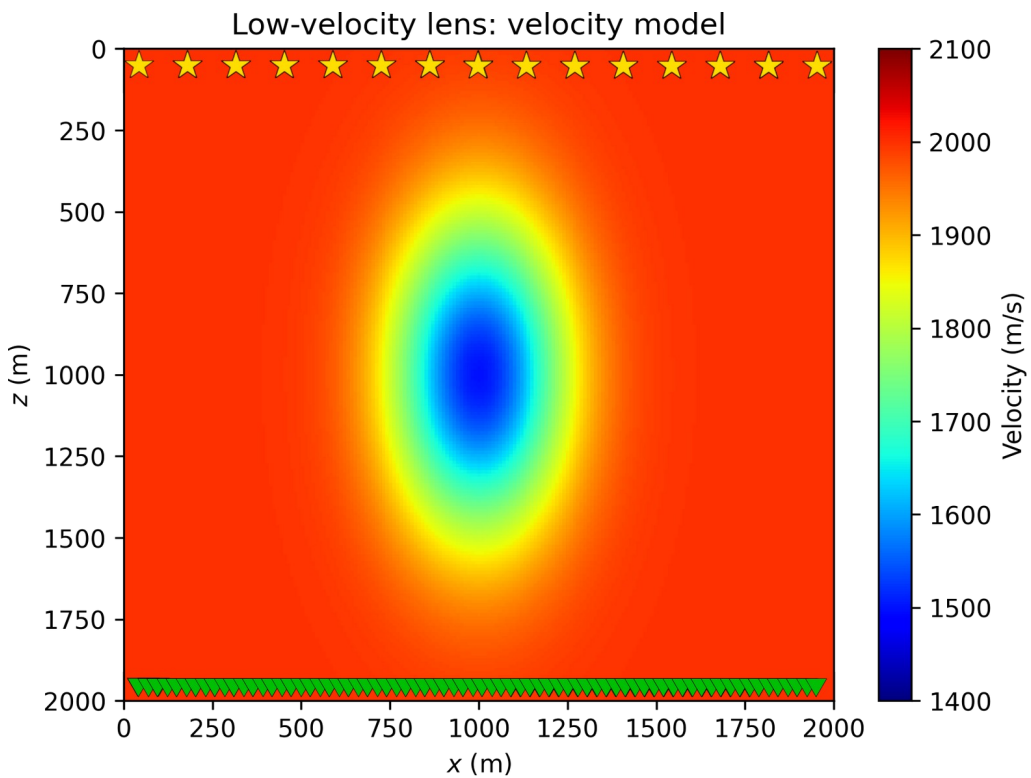
Numerical examples: transmission



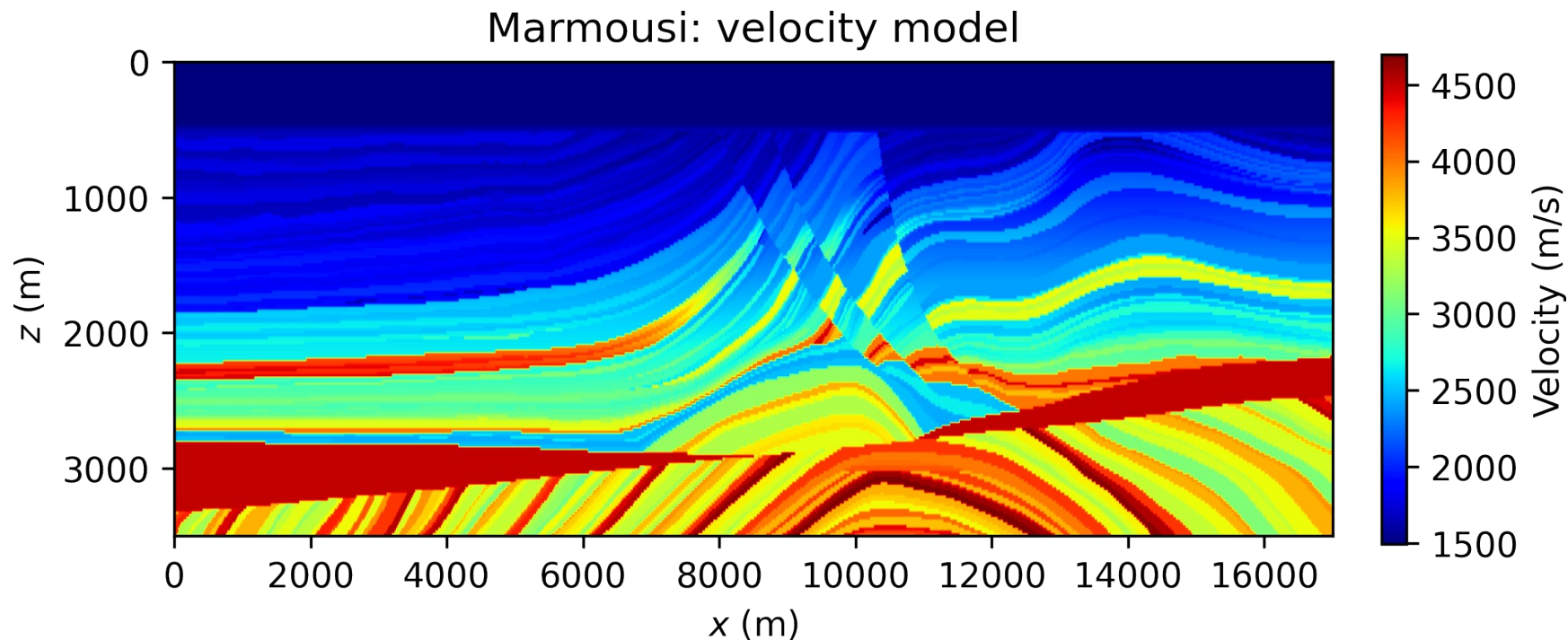
Numerical examples: transmission



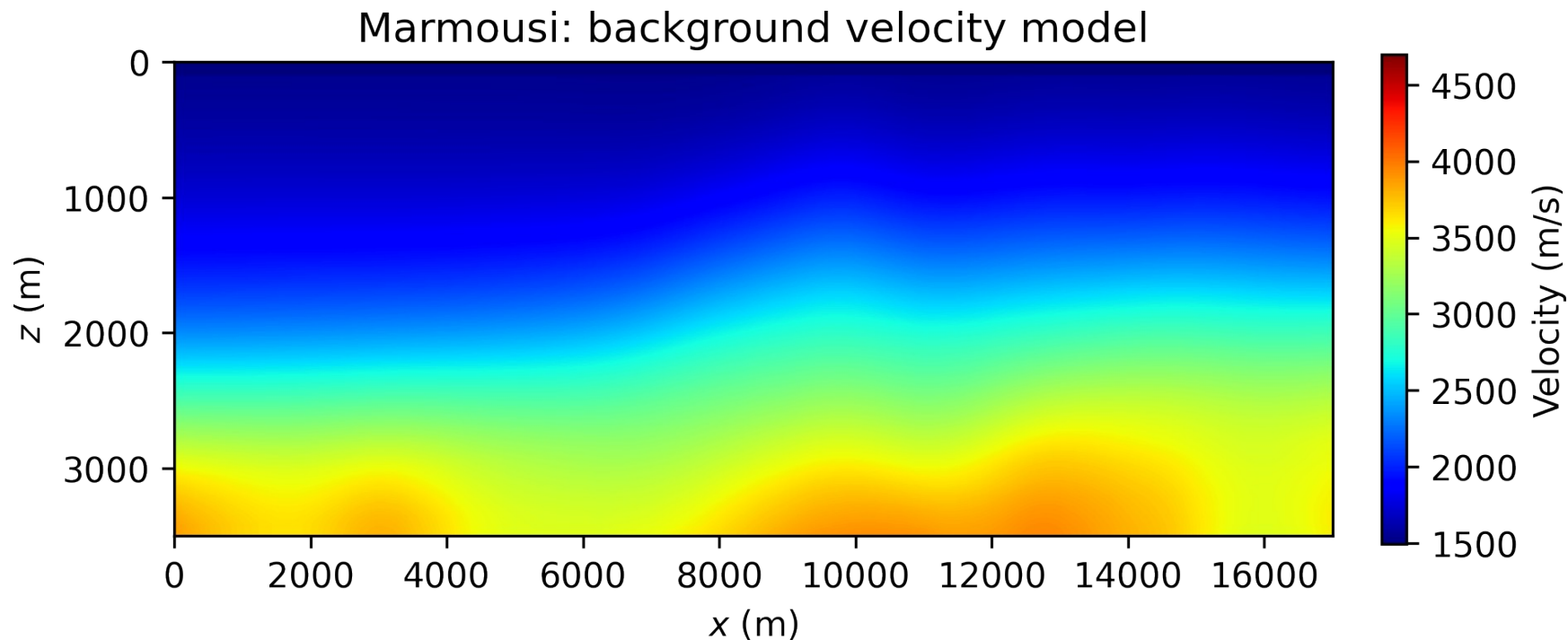
Numerical examples: transmission



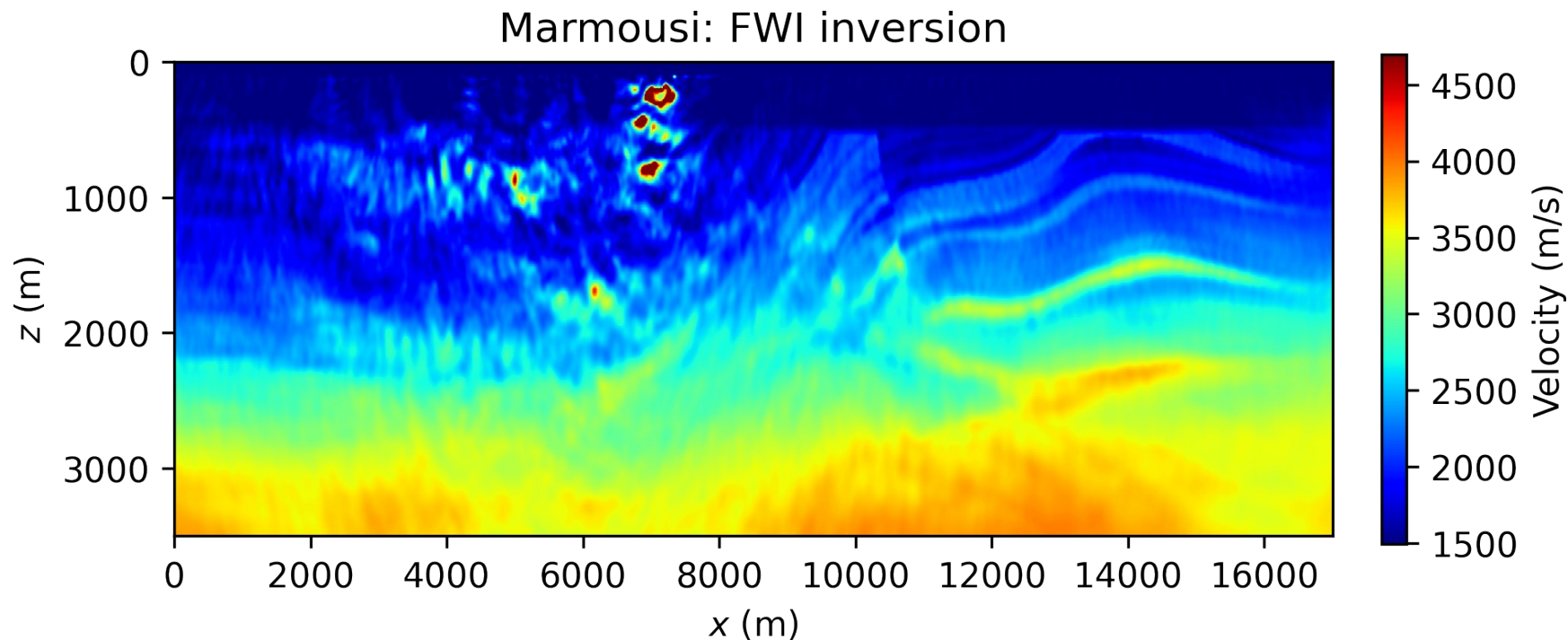
Numerical examples: Marmousi (freq. domain)



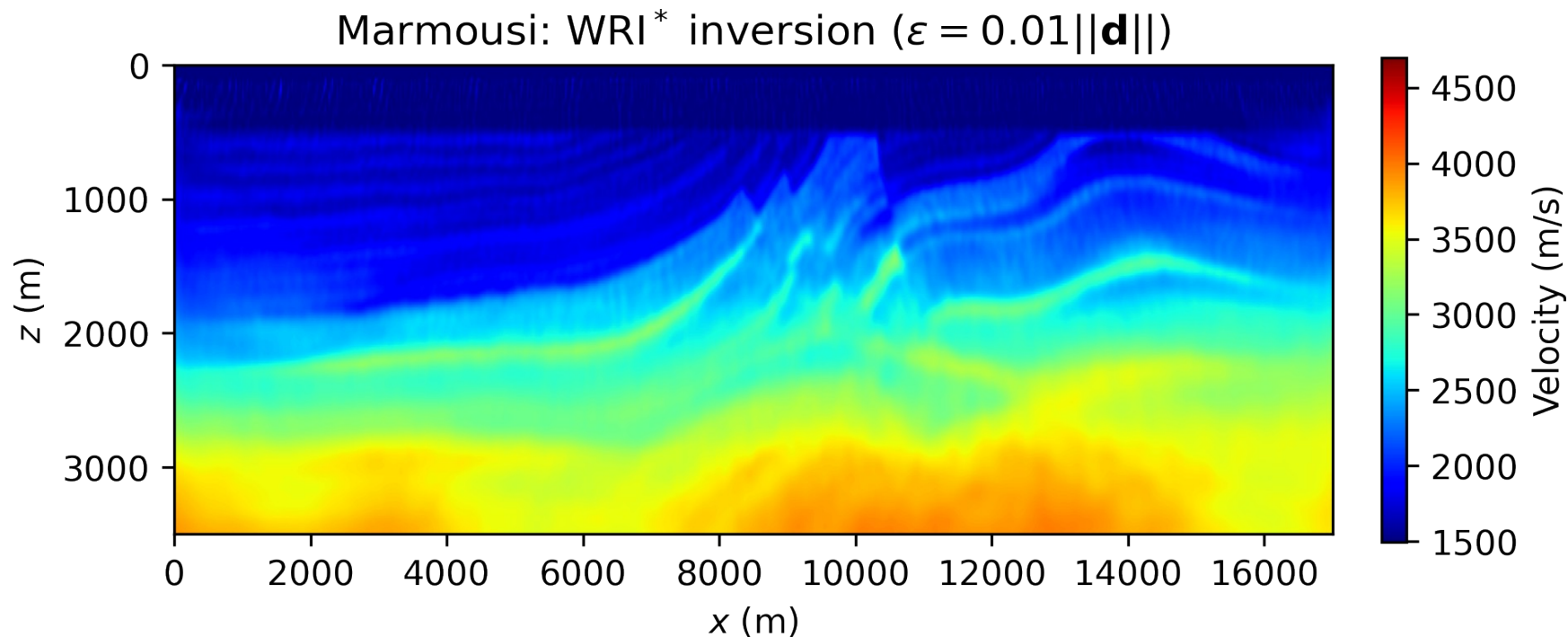
Numerical examples: Marmousi (freq. domain)



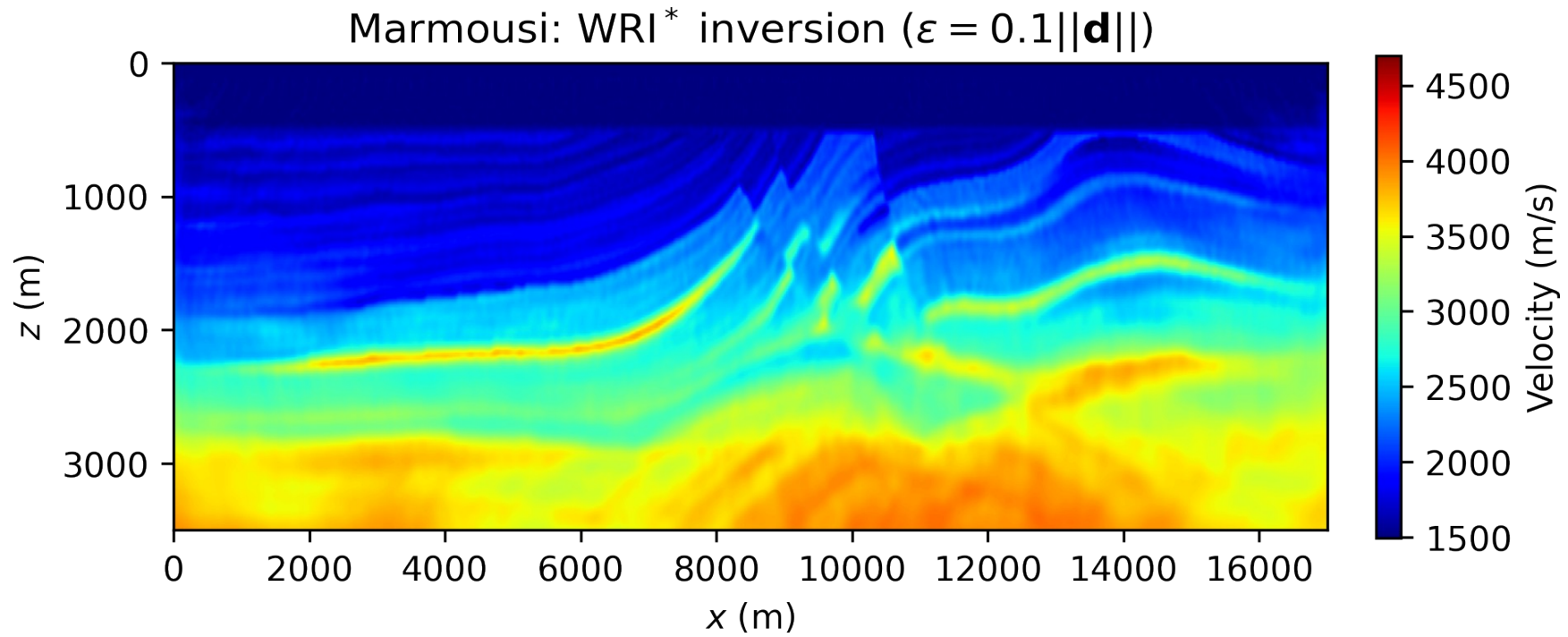
Numerical examples: Marmousi (freq. domain)



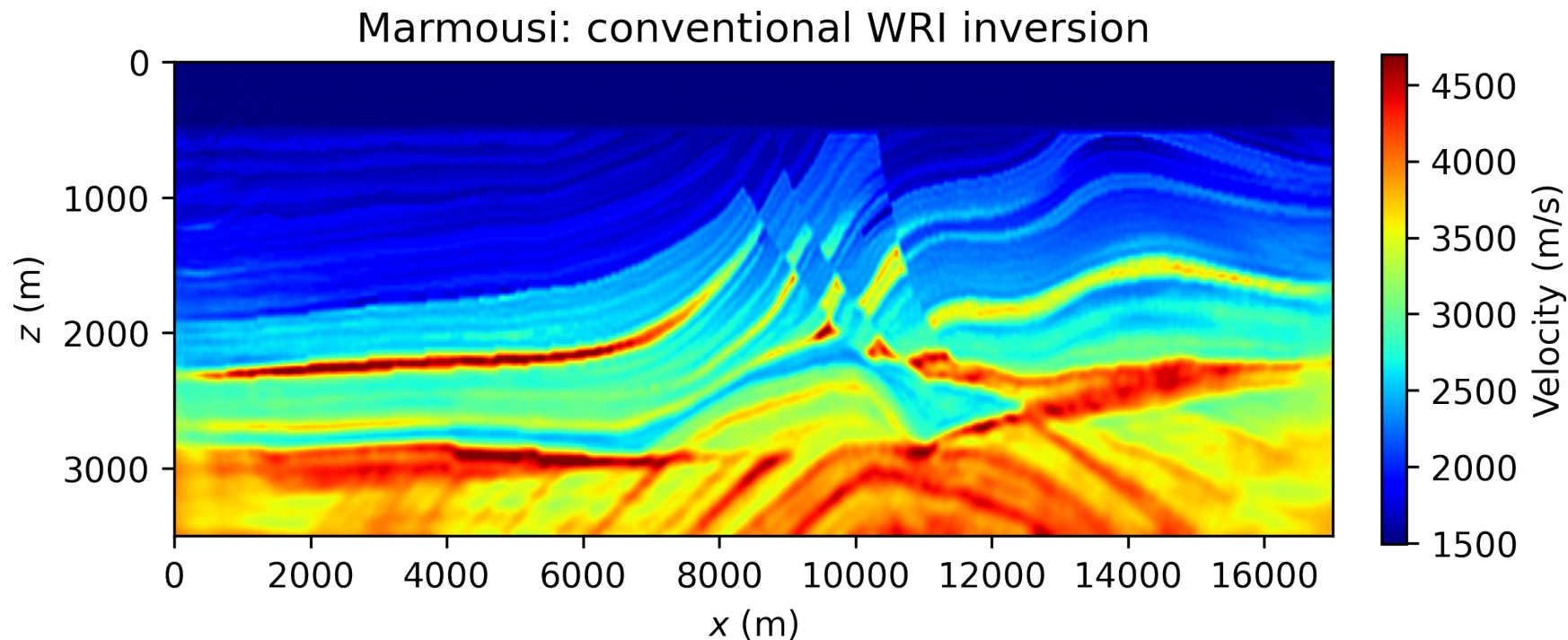
Numerical examples: Marmousi (freq. domain)



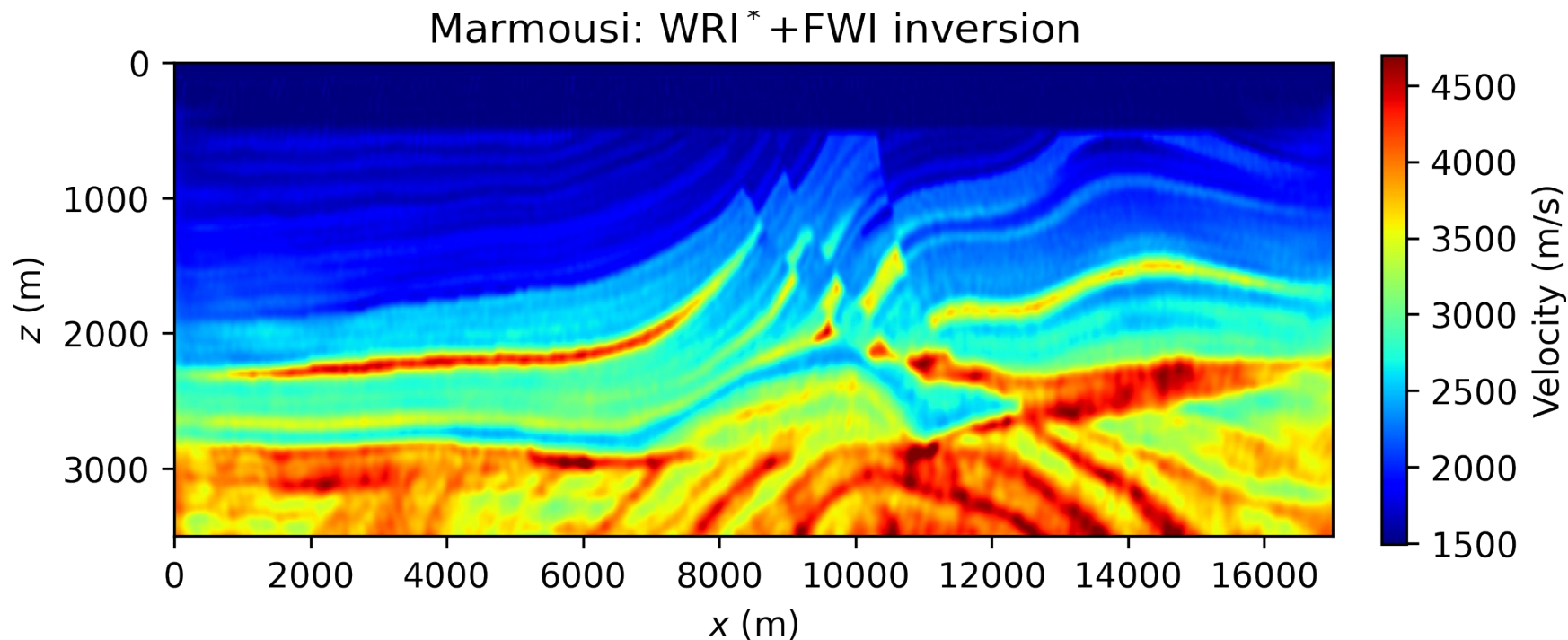
Numerical examples: Marmousi (freq. domain)



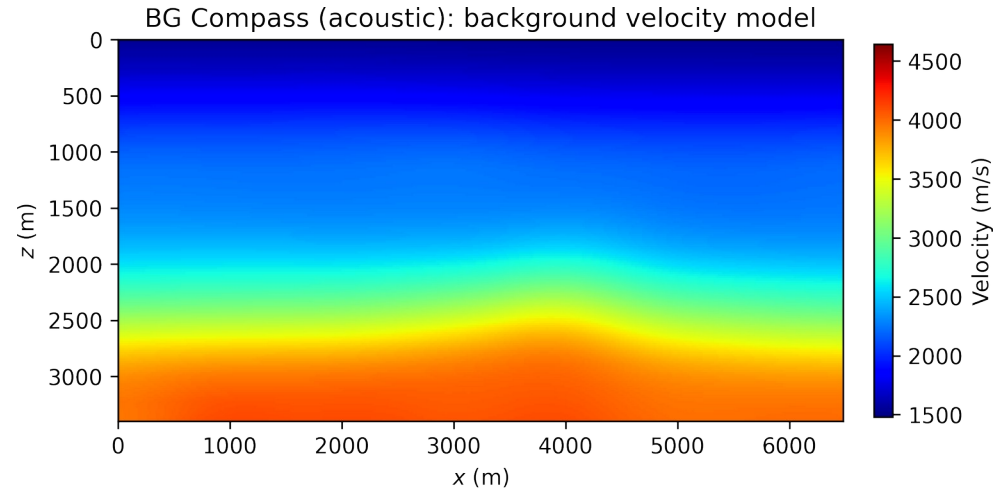
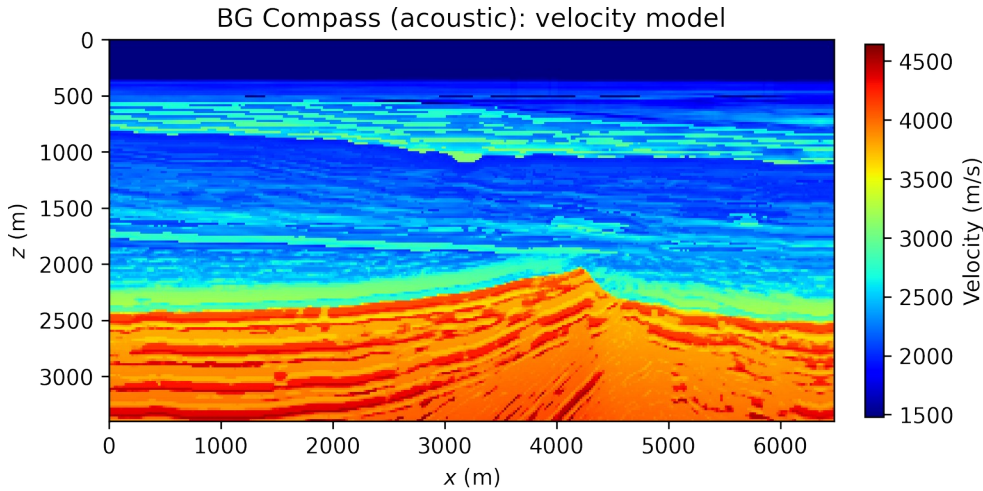
Numerical examples: Marmousi (freq. domain)



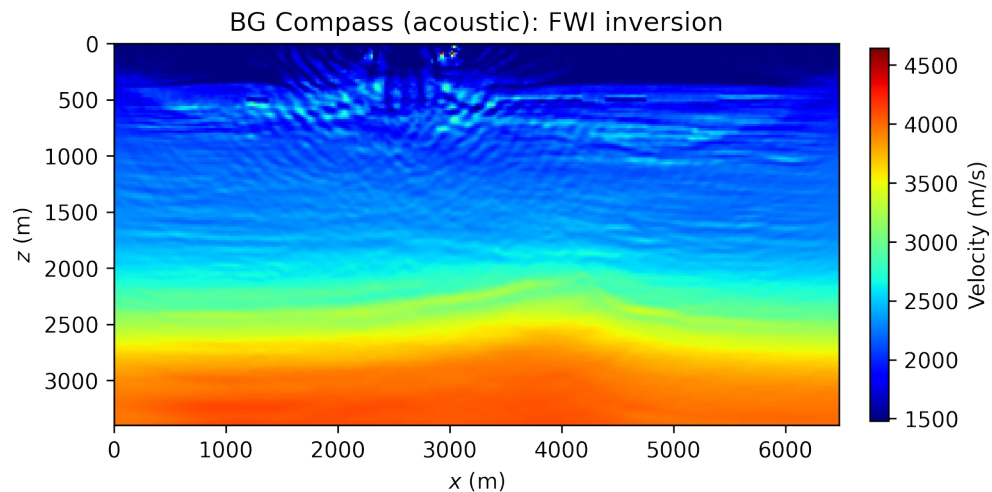
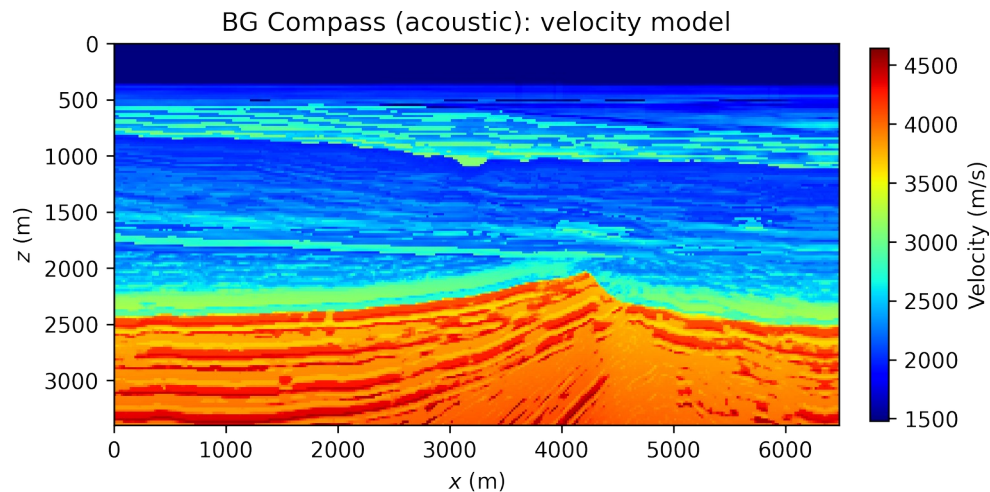
Numerical examples: Marmousi (freq. domain)



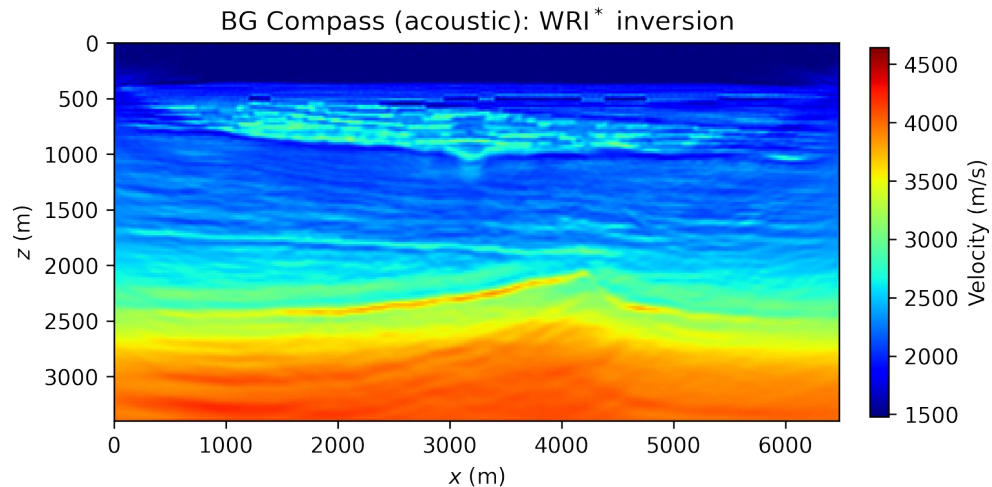
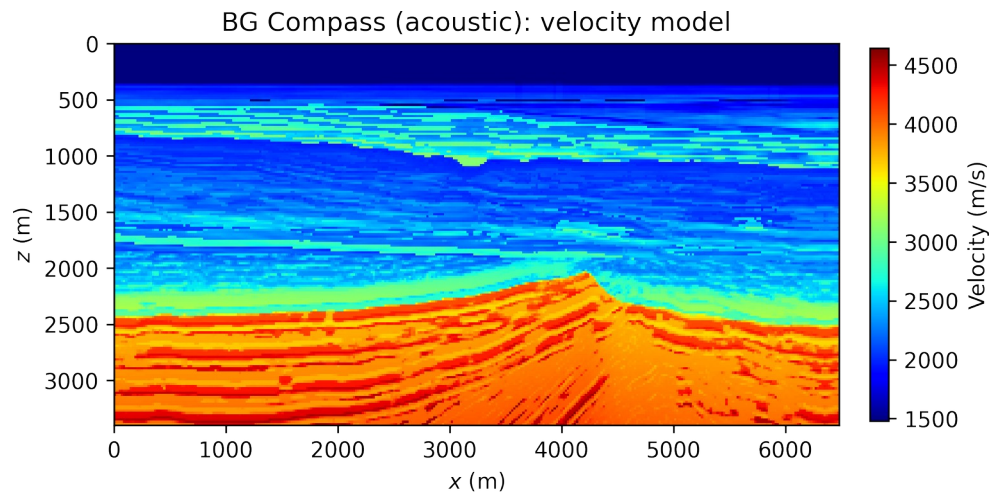
Numerical examples: BG Compass (acoustic)



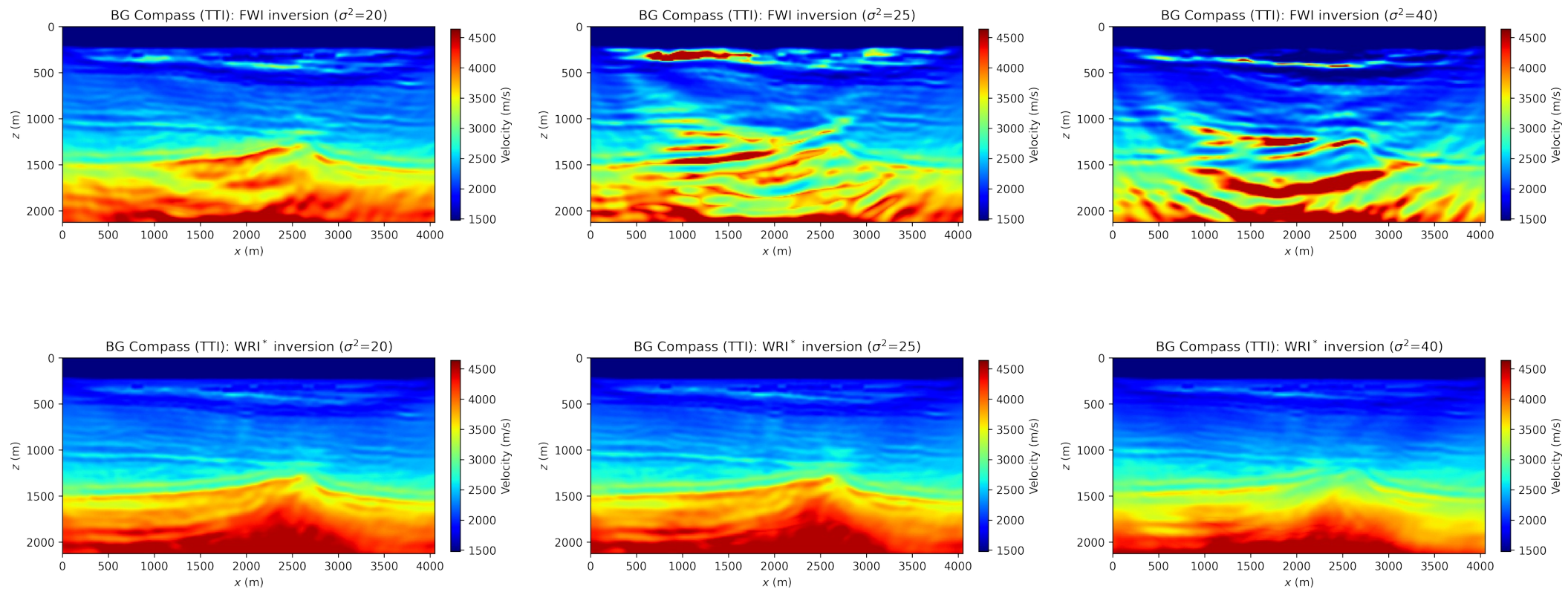
Numerical examples: BG Compass (acoustic)



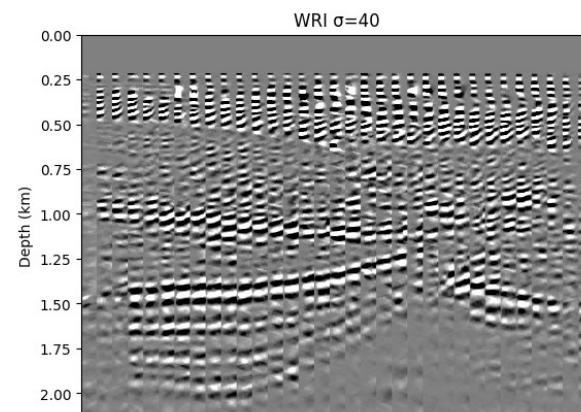
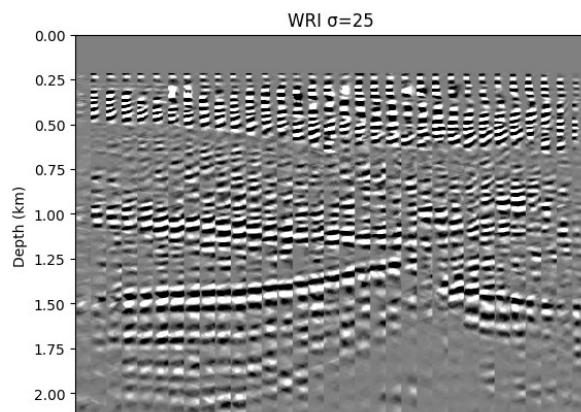
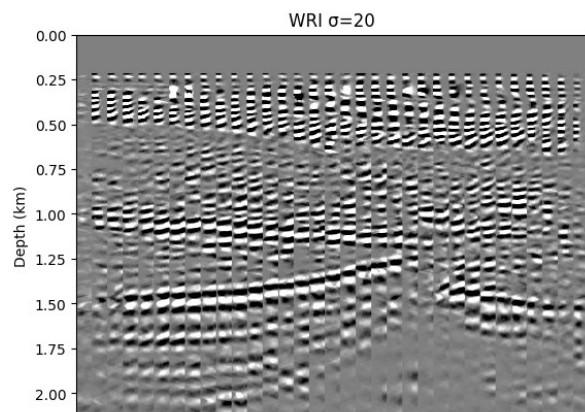
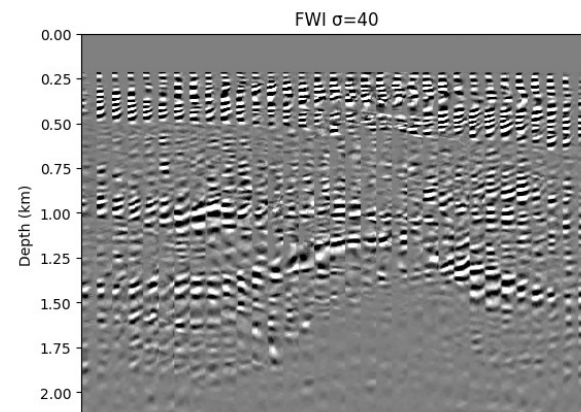
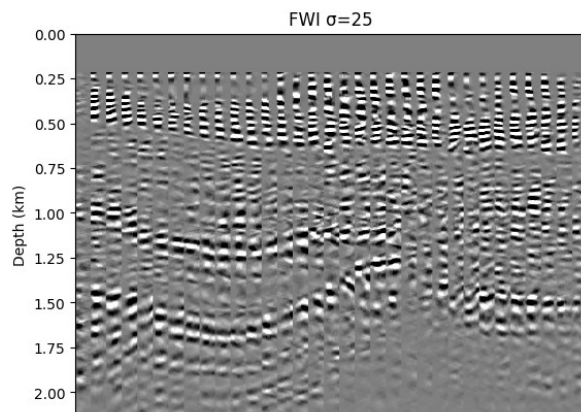
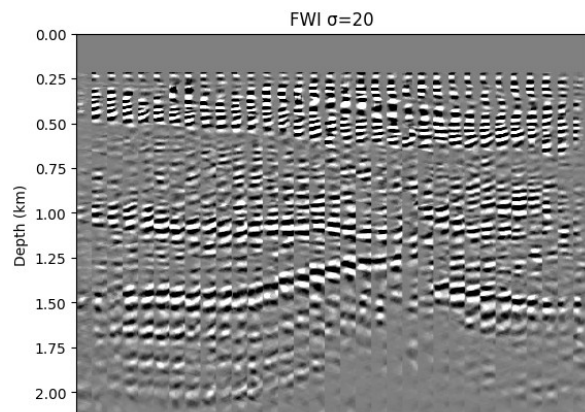
Numerical examples: BG Compass (acoustic)



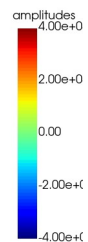
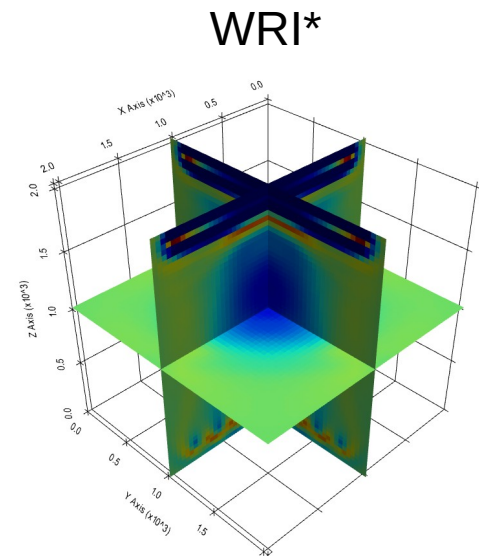
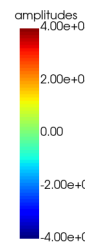
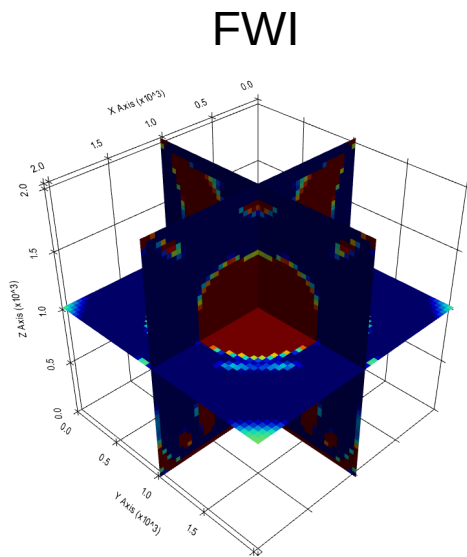
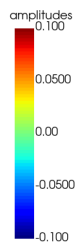
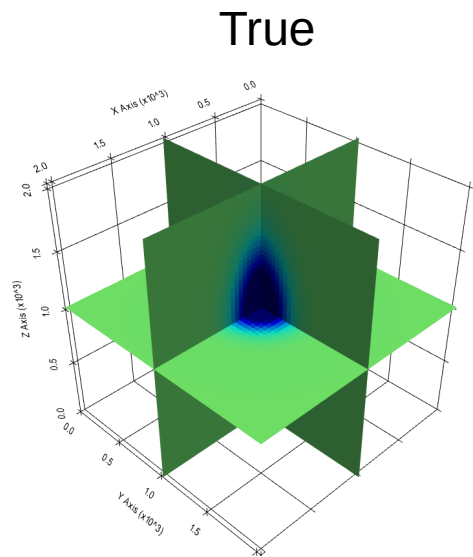
Numerical examples: BG Compass (TTI)



Numerical examples: BG Compass (TTI, CIGs)



Numerical examples: small 3D



Discussion

- WRI*:
 - affordable version of WRI
 - retains robustness of WRI
 - 2nd order methods/hybrid schemes to improve resolution
- Source-focusing annihilator might be necessary to avoid local minima [Symes, W. W., 2020]

Discussion

- Better approximation for augmented variable?
Need for automatic differentiation! [Ablin et al., 2020]

$$\nabla_{\mathbf{m}} \tilde{\mathcal{L}} = \nabla_{\mathbf{m}} \bar{\mathcal{L}} + \frac{\partial \tilde{\mathbf{y}}}{\partial \mathbf{m}}^H \nabla_{\mathbf{y}} \bar{\mathcal{L}}$$

- Some special choice for covariance of data/model error
avoid the need for augmented variable approximation

[van Leeuwen, T., 2019]

Open source implementation



- `github.com/slimgroup/JUDI.jl`
- `devitoproject.org`

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