Source estimation for wavefield-reconstruction inversion

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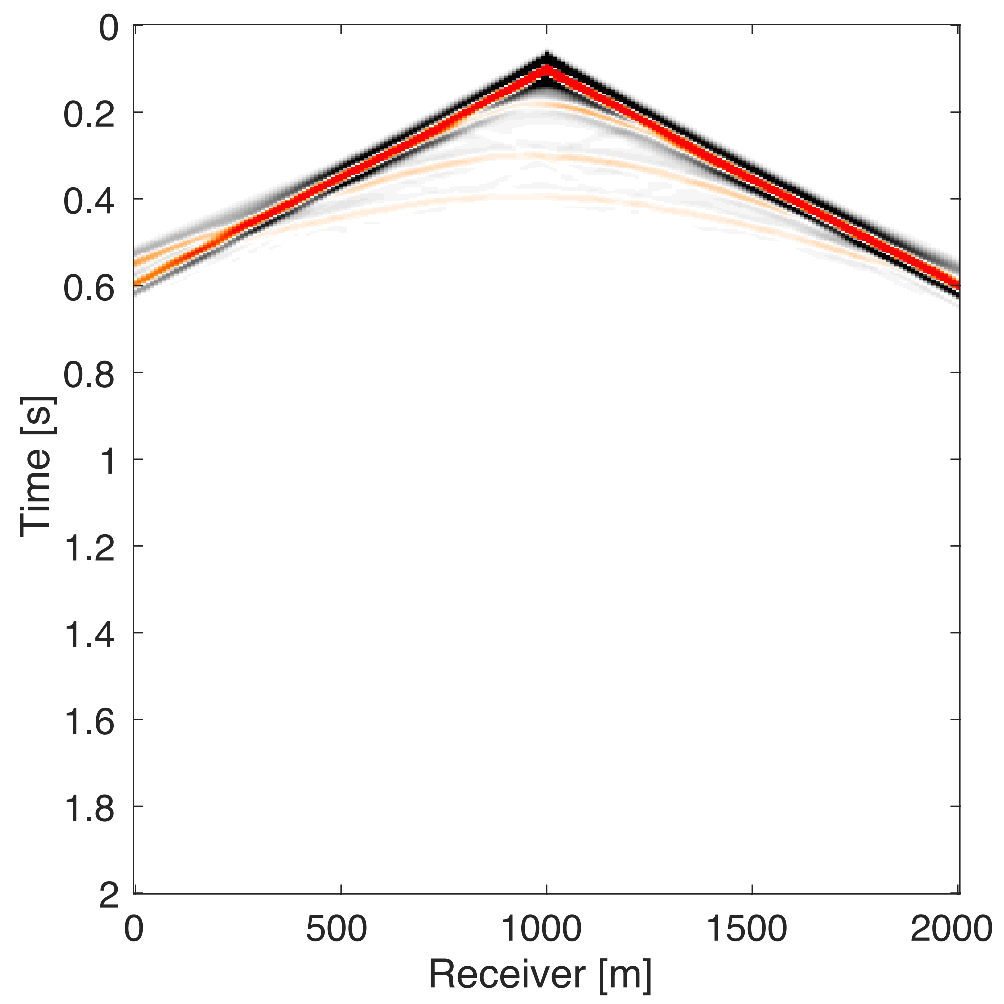
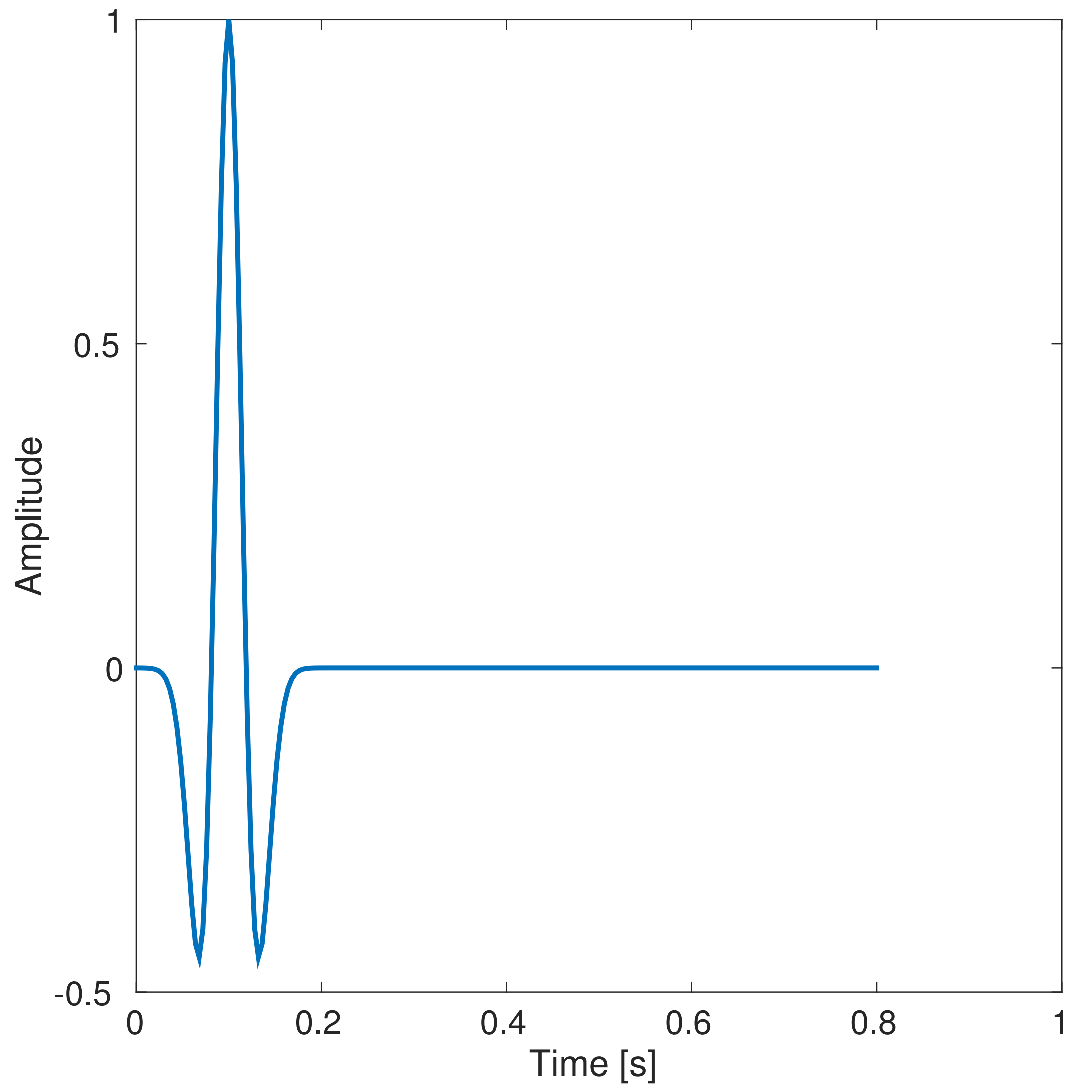
Motivation



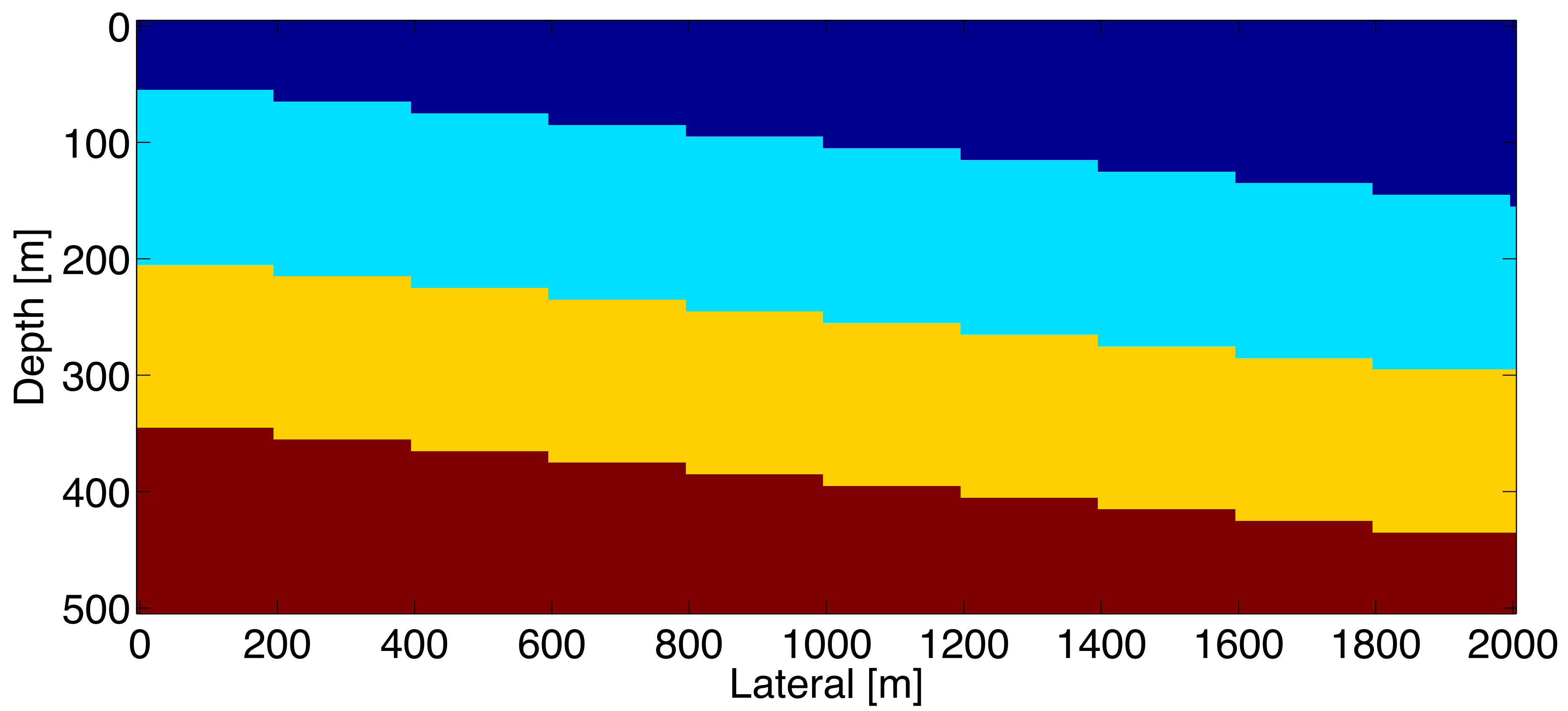
Data

=



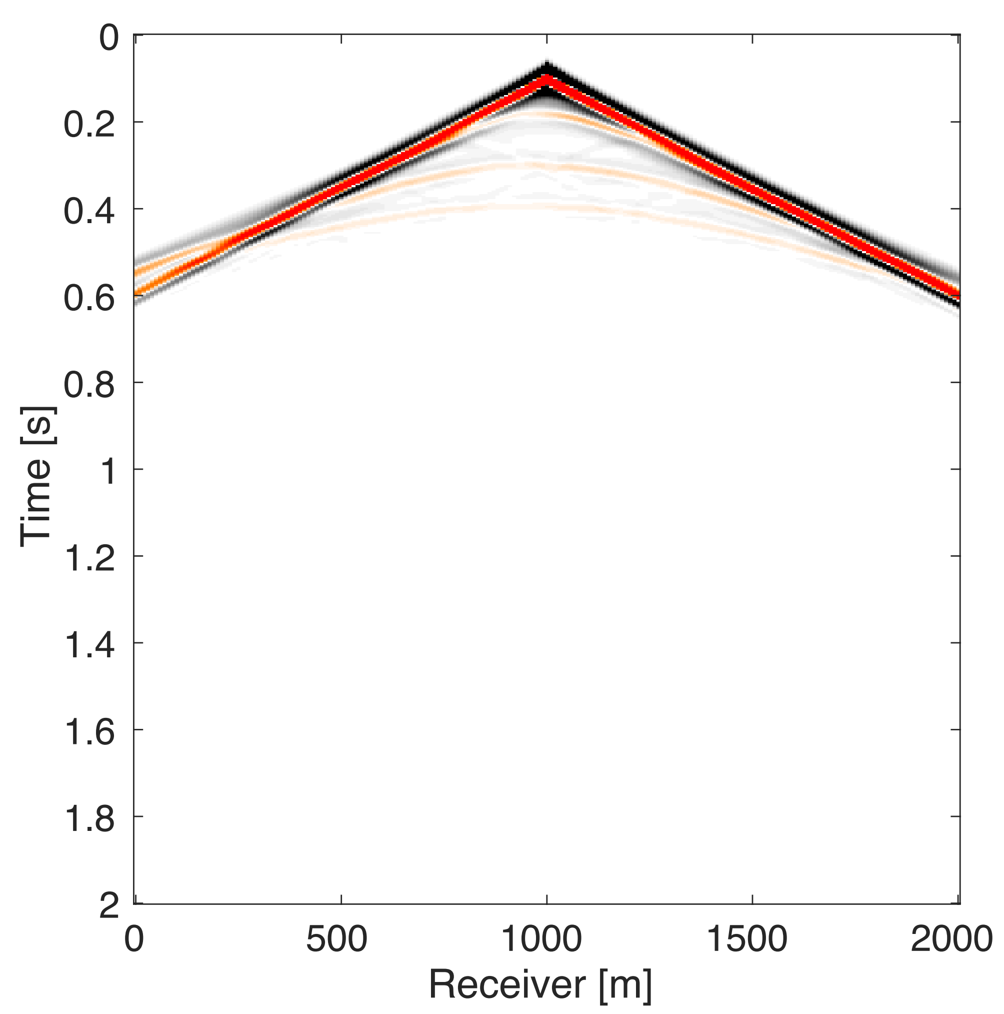
Source wavelet

*



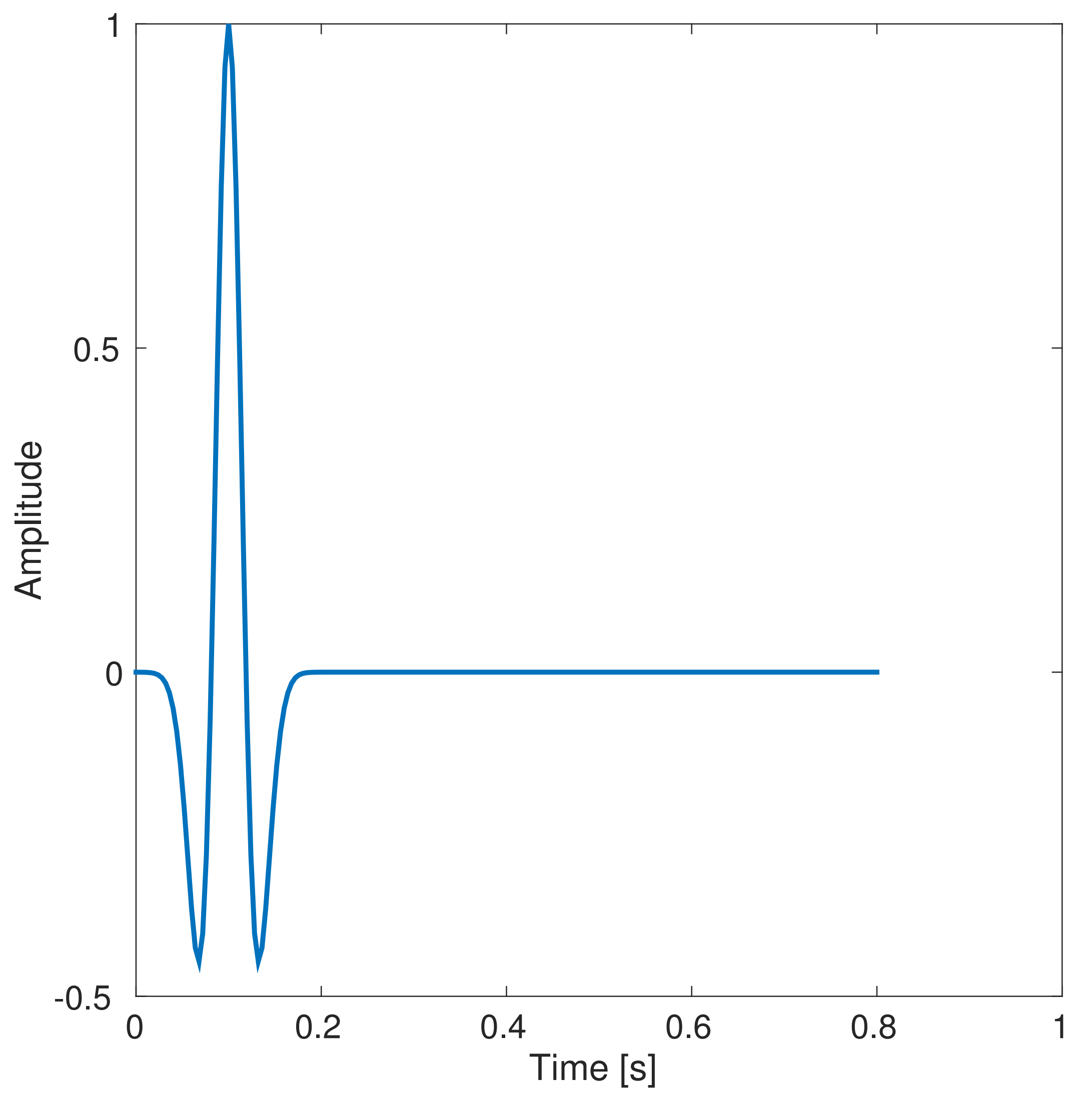
Velocity model

Motivation



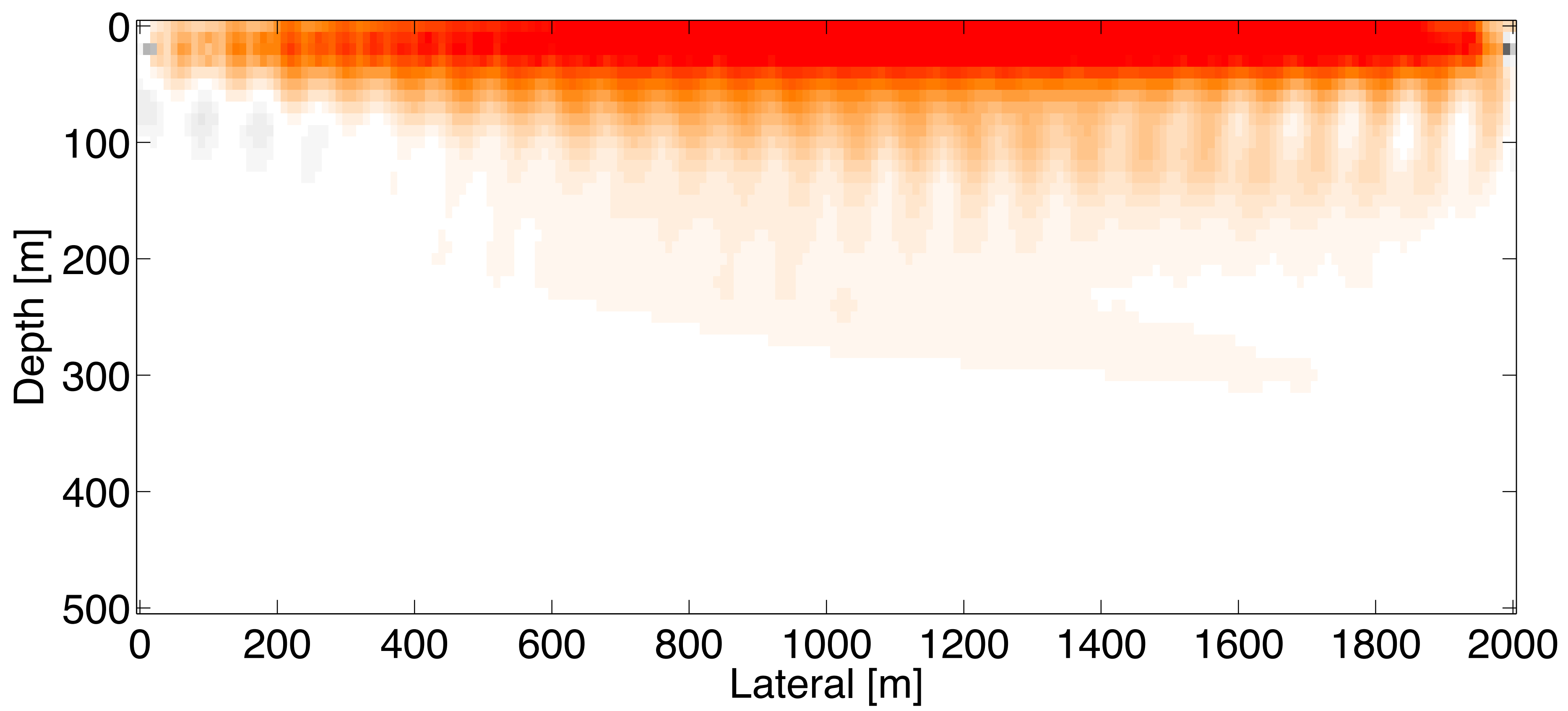
Data

+



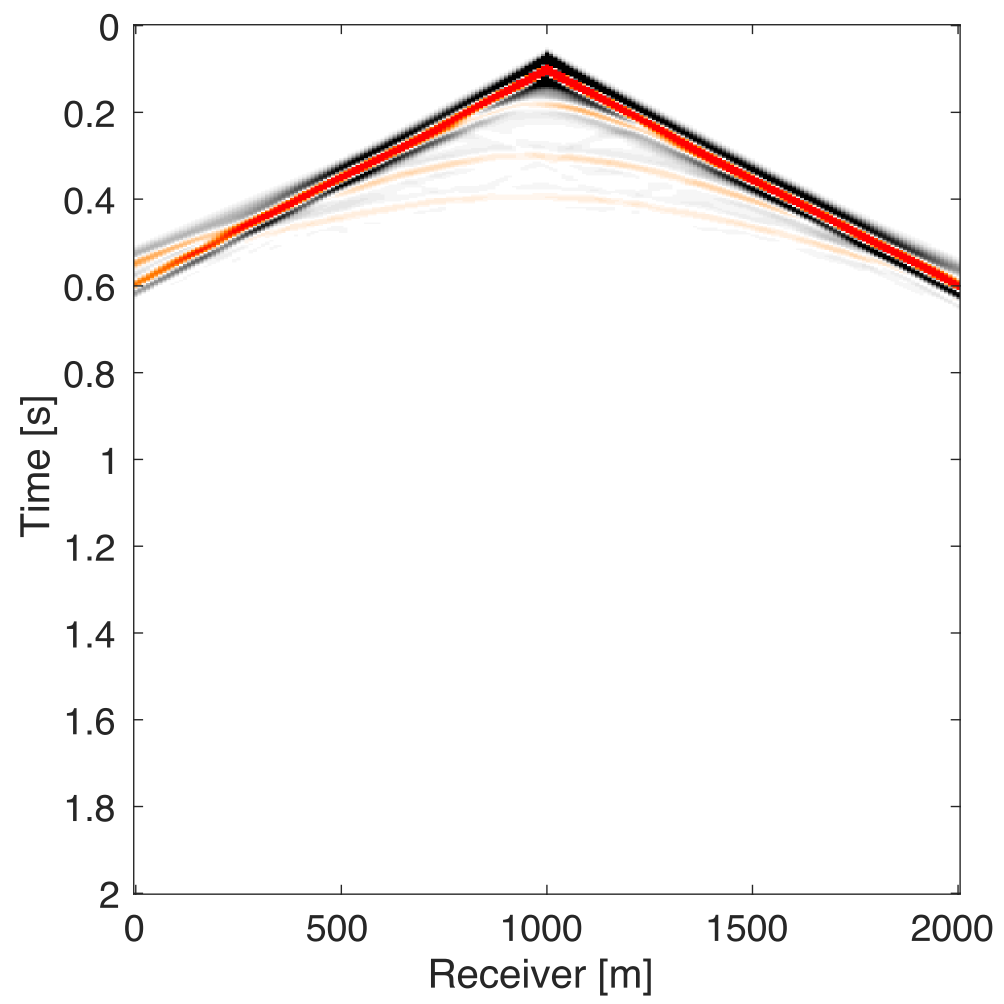
Correct source wavelet

→



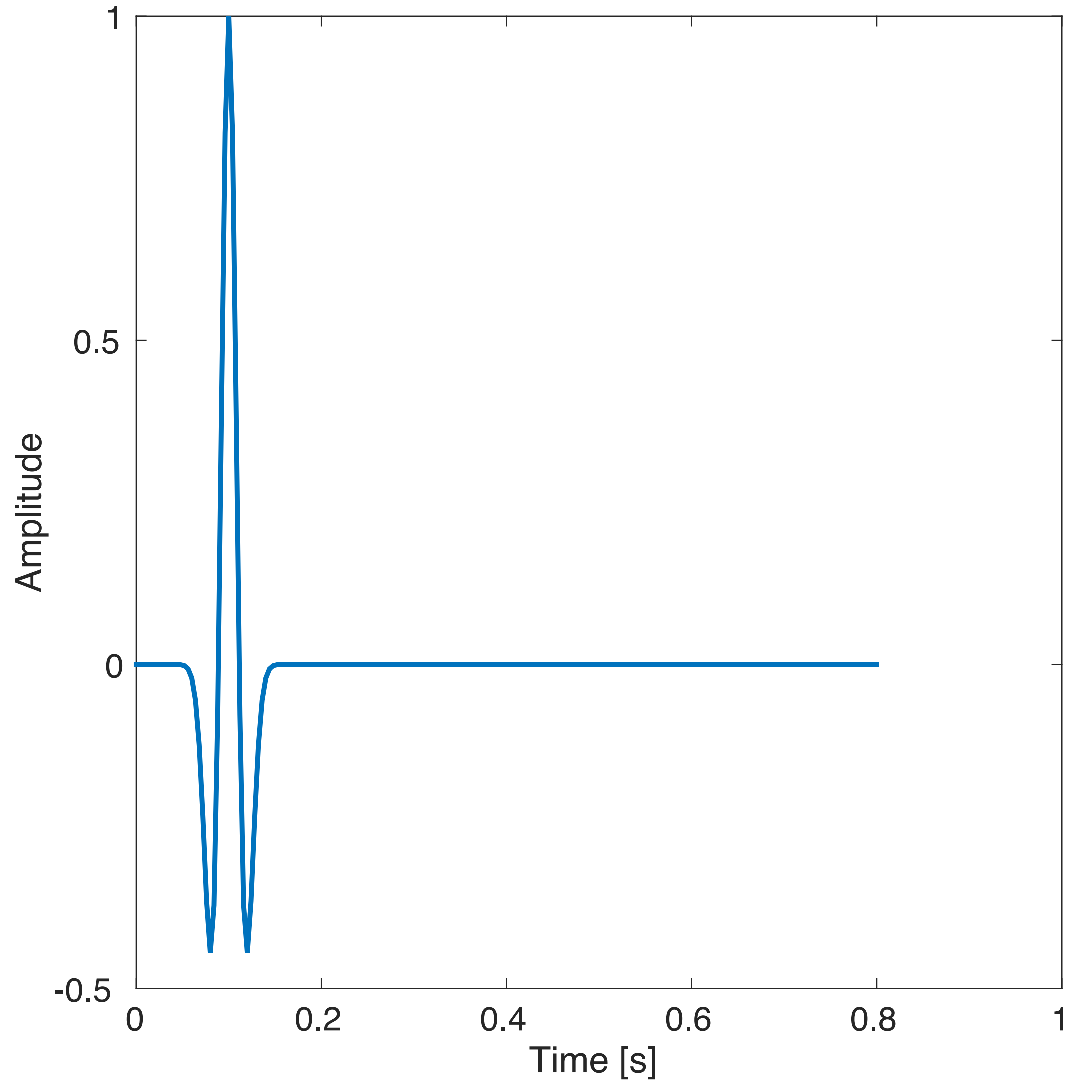
Correct gradient

Motivation

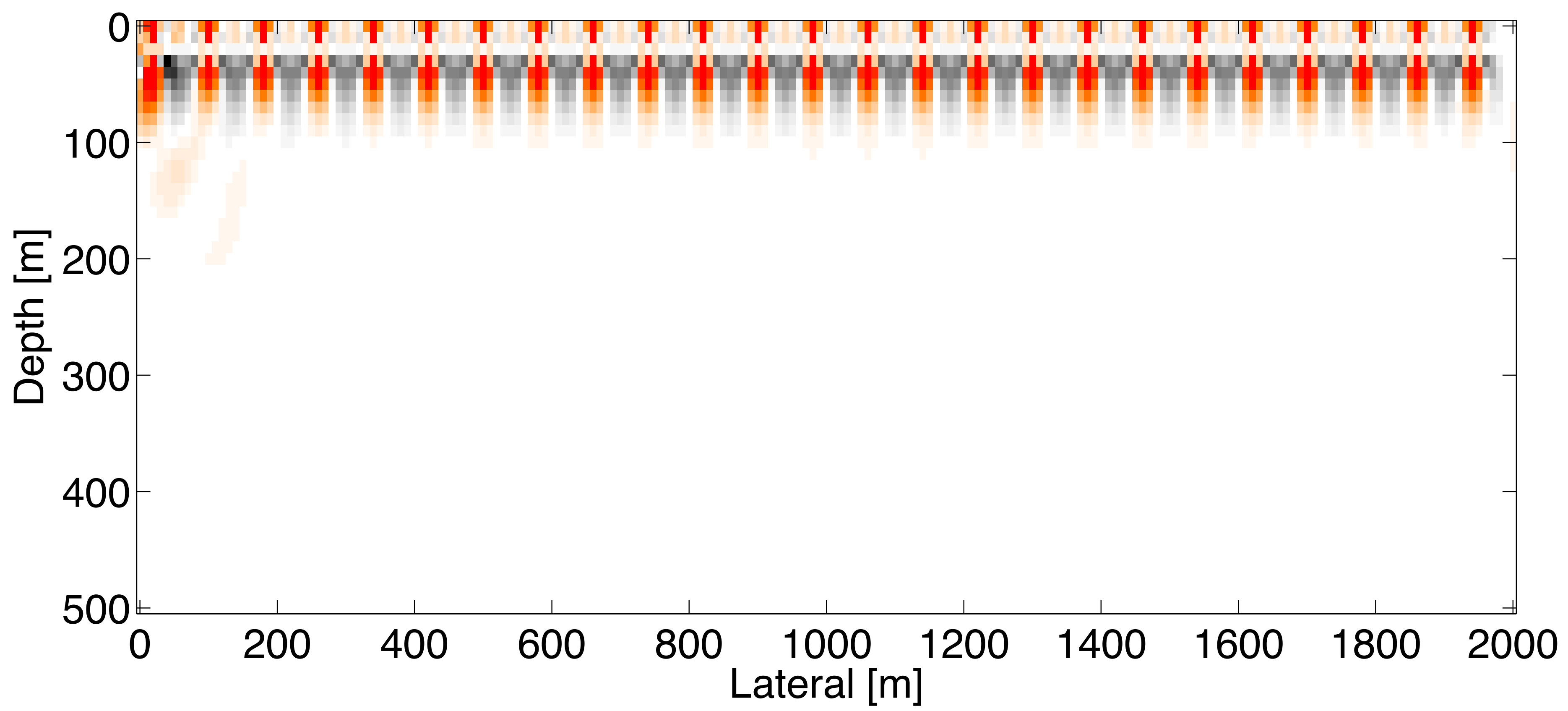


Data

+



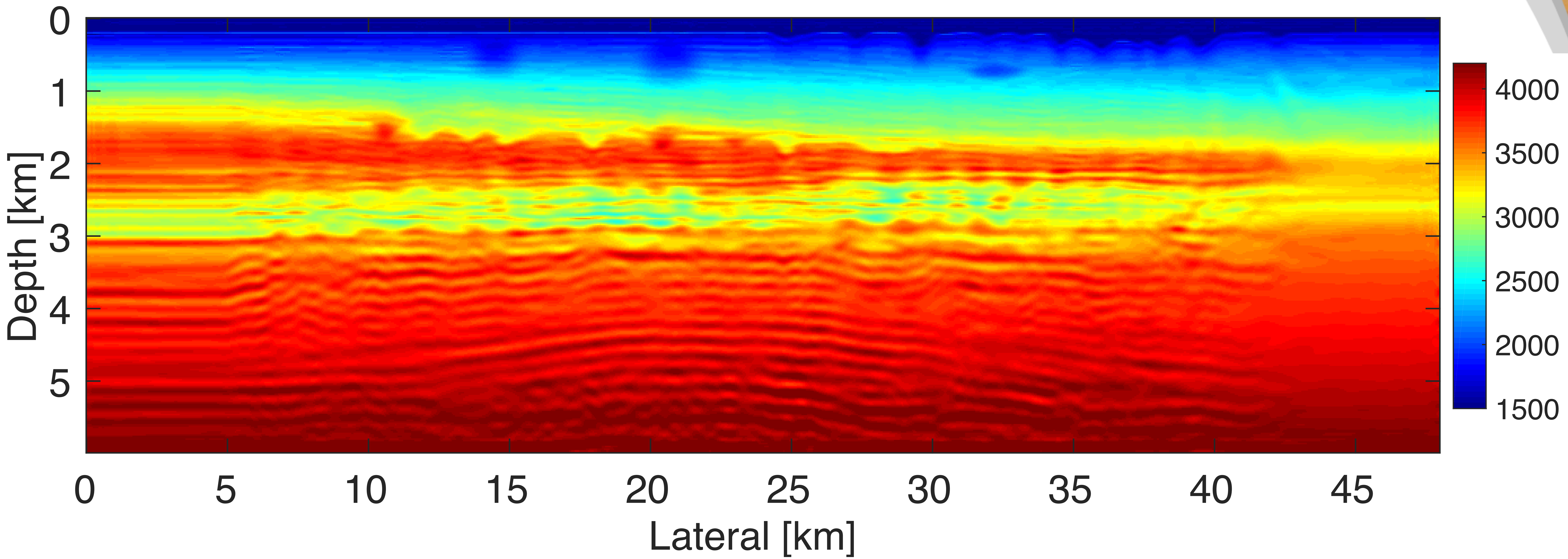
Wrong source wavelet



Wrong gradient

Chevron blind test data

— Wavefield-reconstruction inversion with source estimation



[Tarantola, 1984]

[J. Virieux and S. Operto, 2009]

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Full-waveform inversion

Original problem:

$$\underset{\mathbf{u},\mathbf{m}}{\text{minimize}} \sum_{k,l} \|\mathbf{P}_k \mathbf{u}_{k,l} - \mathbf{d}_{k,l}\|_2^2$$

$$\text{subject to } \mathbf{A}_{k,l}(\mathbf{m})\mathbf{u}_{k,l} = \mathbf{q}_{k,l},$$

where,

$\mathbf{u}_{k,l}$ – Wavefield of the kth shot at lth frequency

$\mathbf{d}_{k,l}$ – Observed data of the kth shot at lth frequency

$\mathbf{q}_{k,l}$ – Source of the kth shot at lth frequency

$\mathbf{A}_{k,l}$ – Helmholtz of the kth shot at lth frequency

$\mathbf{P}_k$ – Receiver projection operator of the kth shot

$\mathbf{m}$ – Squared-slowness

[Tarantola, 1984]

[J. Virieux and S. Operto, 2009]

Full-waveform inversion

Reduced/adjoint-state method:

$$\underset{\mathbf{m}}{\text{minimize}} \sum_{k,l} \|\mathbf{P}_k \mathbf{A}_{k,l}(\mathbf{m})^{-1}\mathbf{q}_{k,l} - \mathbf{d}_{k,l}\|_2^2$$

with the gradient given by

$$\mathbf{g} = \sum_{k,l} \mathbf{u}_{k,l}^* \frac{\partial \mathbf{A}_{k,l}^*}{\partial \mathbf{m}} \mathbf{v}_{k,l}$$

$$\mathbf{u}_{k,l} = \mathbf{A}_{k,l}(\mathbf{m})^{-1}\mathbf{q}_{k,l}$$

$$\mathbf{v}_{k,l} = \mathbf{A}_{k,l}^{-*}(\mathbf{m})\mathbf{P}_k^*\mathbf{r}_{k,l}$$

$$\mathbf{r}_{k,l} = \mathbf{P}_k \mathbf{A}_{k,l}(\mathbf{m})^{-1}\mathbf{q}_{k,l} - \mathbf{d}_{k,l}$$

2 PDE solves are required !

[van Leeuwen, T and Herrmann, F J , 2013]
[Peters, B, Herrmann, F J and van Leeuwen, T, 2014]
[Golub, G H and Pereyra, V , 1973]

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Wavefield-reconstruction inversion

Joint optimization problem:

$$\underset{\mathbf{u},\mathbf{m}}{\text{minimize}} \sum_{k,l} \|\mathbf{P}_k \mathbf{u}_{k,l} - \mathbf{d}_{k,l}\|_2^2 + \lambda^2 \|\mathbf{A}_{k,l}(\mathbf{m})\mathbf{u}_{k,l} - \mathbf{q}_{k,l}\|_2^2$$

Eliminating $\mathbf{u}$ w/ variable projection:

$$\bar{\mathbf{u}} = \arg\min_{\mathbf{u}} \sum_{k,l} \|\mathbf{P}_k \mathbf{u}_{k,l} - \mathbf{d}_{k,l}\|_2^2 + \lambda^2 \|\mathbf{A}_{k,l}(\mathbf{m})\mathbf{u}_{k,l} - \mathbf{q}_{k,l}\|_2^2$$

[van Leeuwen, T and Herrmann, F J , 2013]

[Golub, G and Pereyra, V, 1973]

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Wavefield-reconstruction inversion

Corresponds to solving the following augmented system:

$$\begin{pmatrix} \lambda \mathbf{A}_{k,l} \\ \mathbf{P}_k \end{pmatrix} \overline{\mathbf{u}}_{k,l} = \begin{pmatrix} \lambda \mathbf{q}_{k,l} \\ \mathbf{d}_{k,l} \end{pmatrix}$$

with the gradient

1 augmented system solves is required !

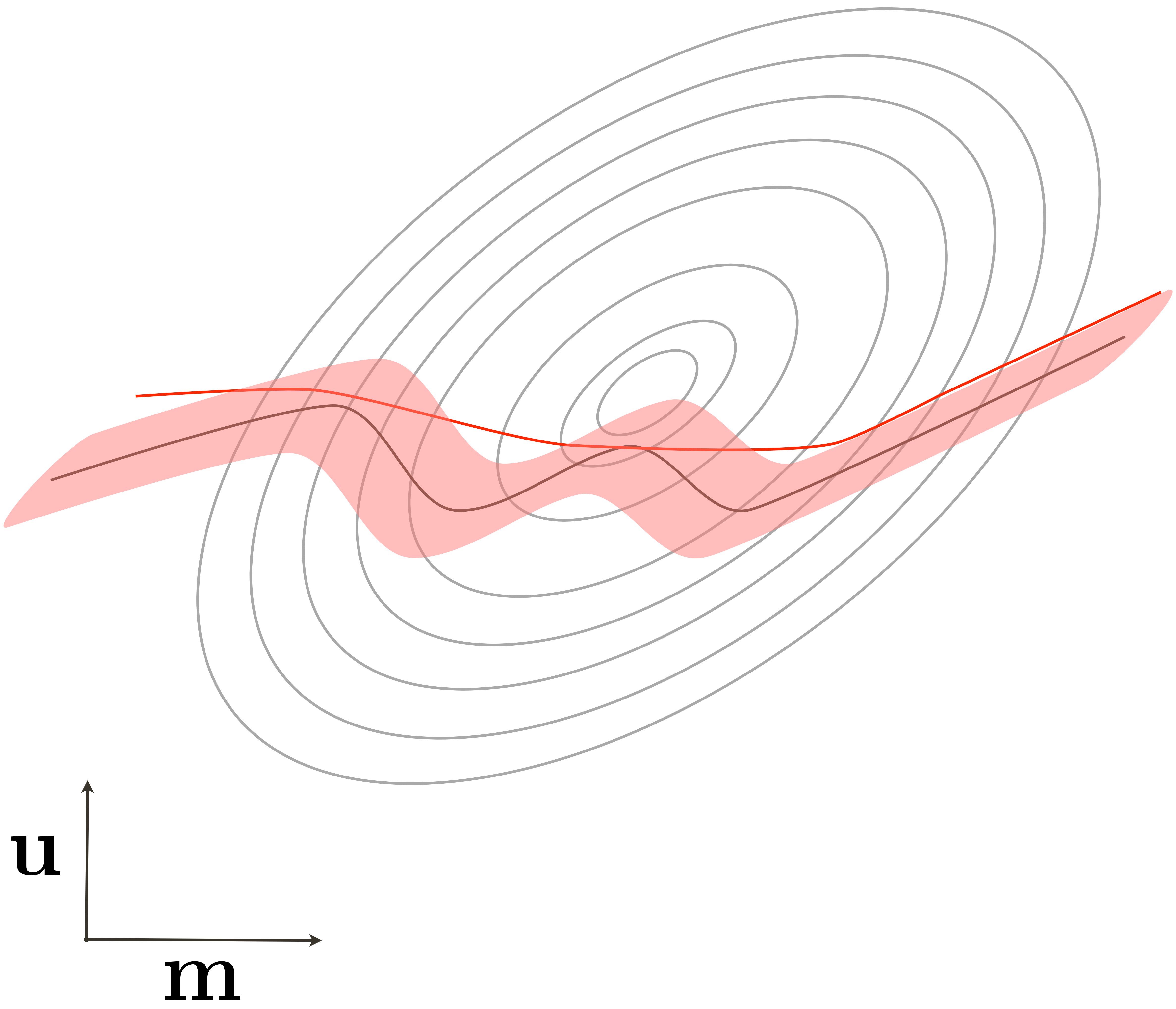
$$\mathbf{g} = \sum_{k,l} \overline{\mathbf{u}}_{k,l}^* \frac{\partial \mathbf{A}_{k,l}^*}{\partial \mathbf{m}} \overline{\mathbf{v}}_{k,l}$$

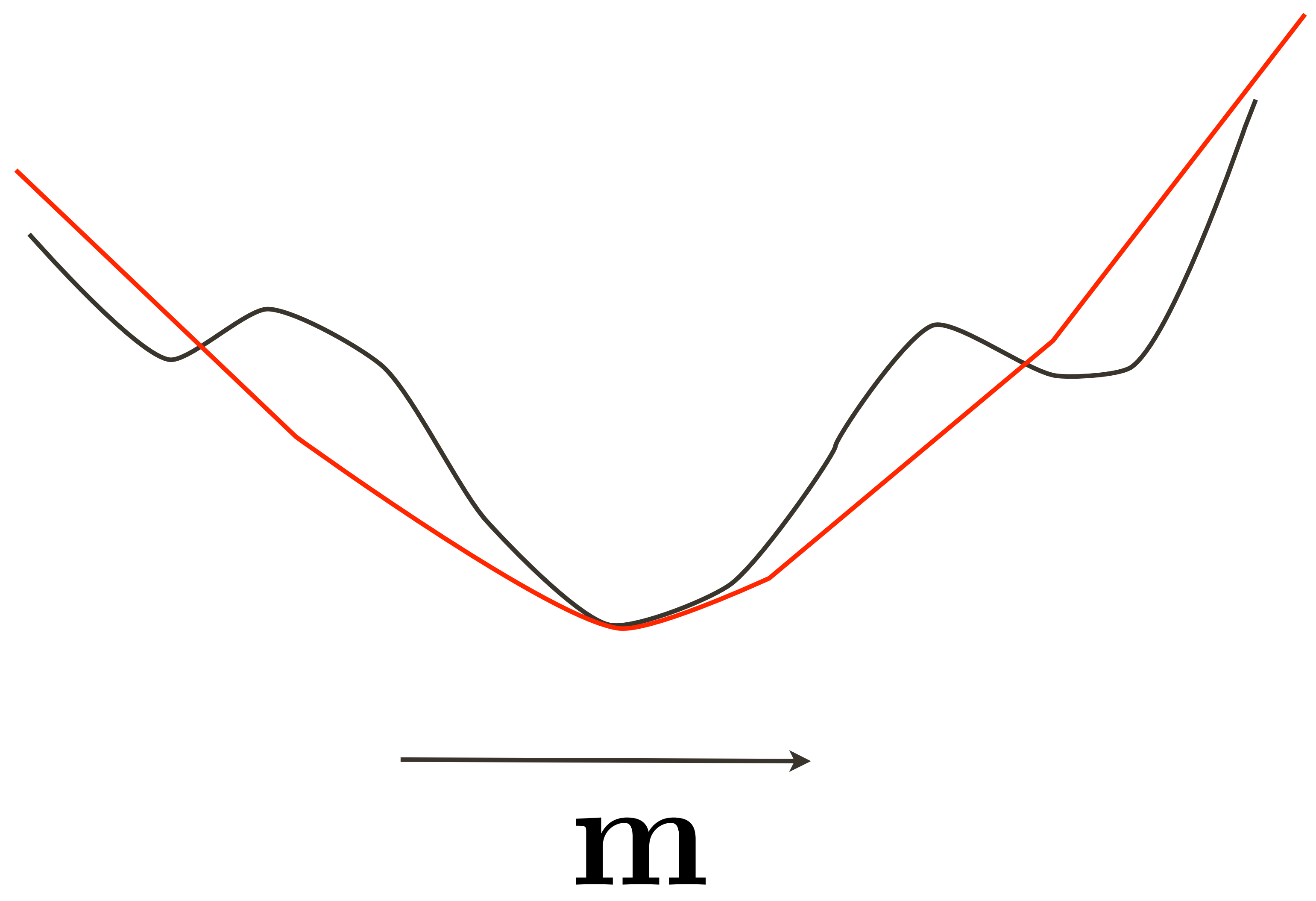
$$\overline{\mathbf{v}}_{k,l} = \mathbf{A}_{k,l}(\mathbf{m}) \overline{\mathbf{u}}_{k,l} - \mathbf{q}_{k,l}$$

WRI vs. FWI

[van Leeuwen, T and Herrmann, F J , 2013]

[Peters, B, Herrmann, F J and van Leeuwen, T, 2014]

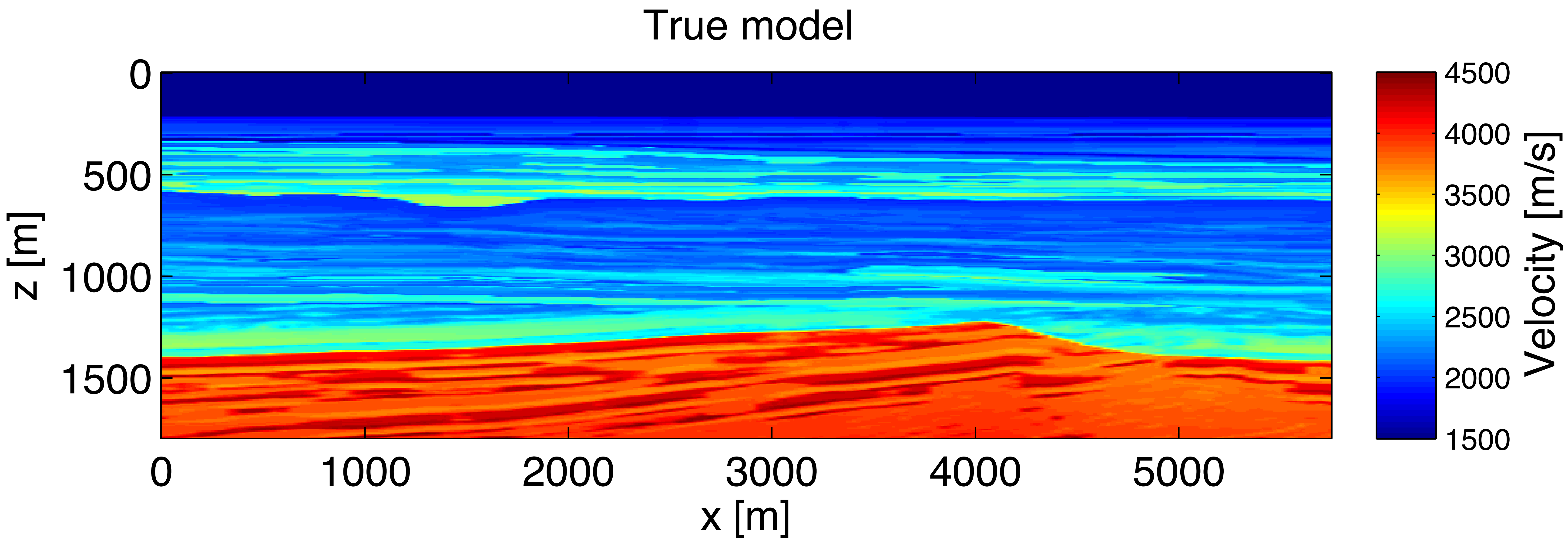


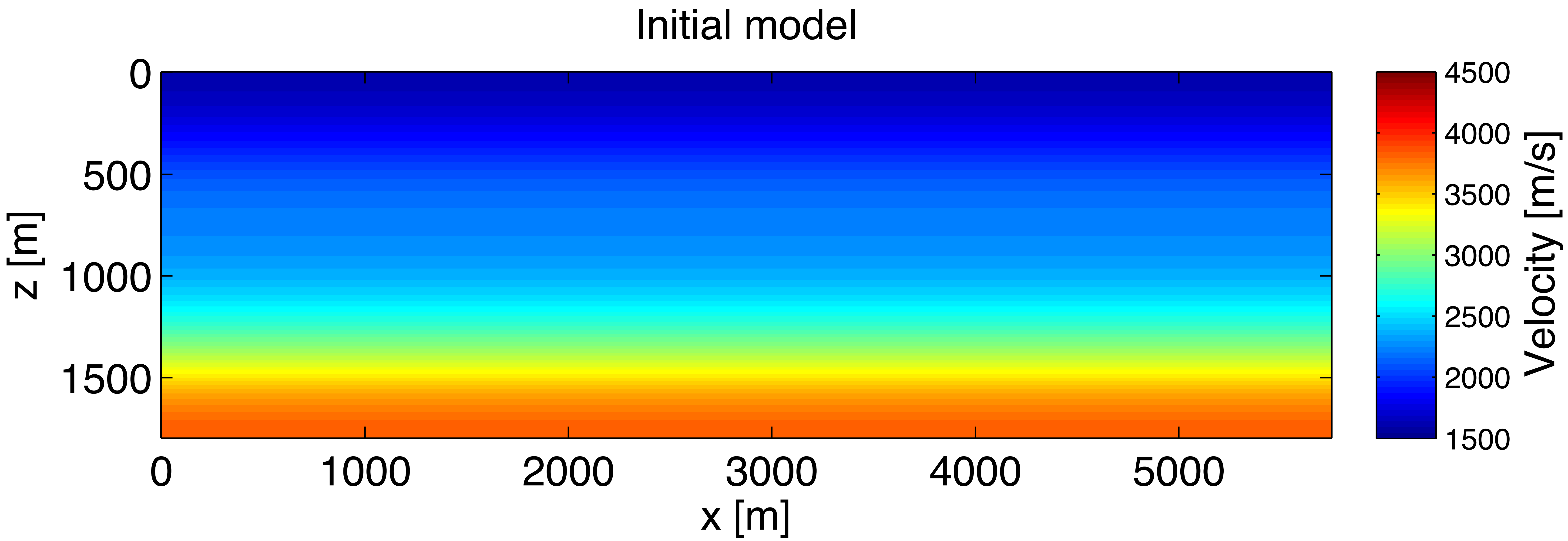


True & initial model

[van Leeuwen, T and Herrmann, F J , 2013]

[Peters, B, Herrmann, F J and van Leeuwen, T, 2014]

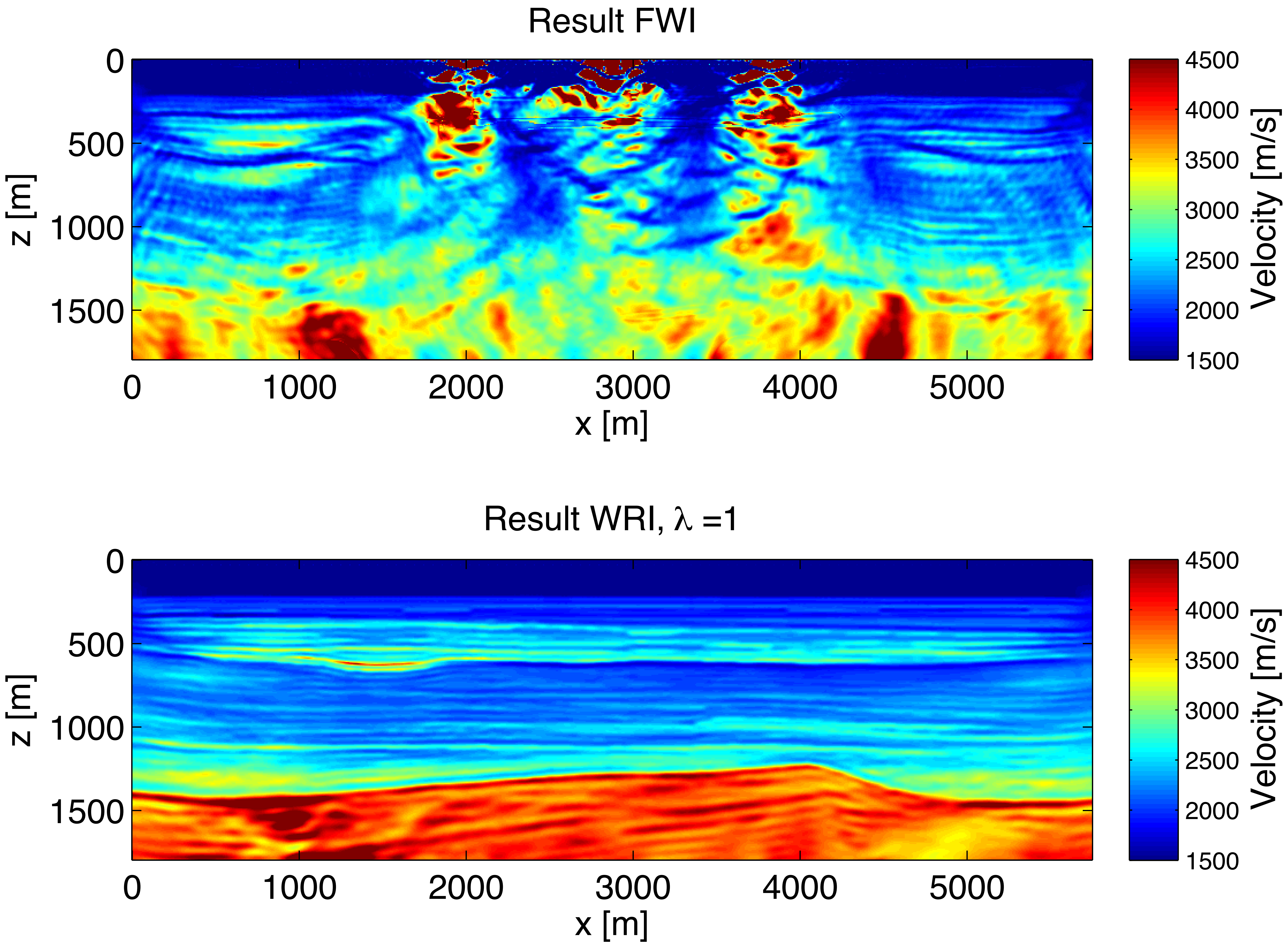




FWI vs WRI

[van Leeuwen, T and Herrmann, F J , 2013]

[Peters, B, Herrmann, F J and van Leeuwen, T, 2014]



WRI with source estimation

Triple parameters optimization problem:

$$\underset{\mathbf{u},\mathbf{m},\alpha}{\text{minimize}} \sum_{k,l} \|\mathbf{P}_k \mathbf{u}_{k,l} - \mathbf{d}_{k,l}\|_2^2 + \lambda^2 \|\mathbf{A}_{k,l}(\mathbf{m})\mathbf{u}_{k,l} - \alpha_{k,l}\mathbf{e}_{k,l}\|_2^2$$

[Aravkin, A Y, van Leeuwen, T, Calandra, H, and Herrmann, F J, 2012]
[Li, M, Rickett, J, and Abubakar, A, 2013]

FWI with source estimation

Joint optimization problem:

$$\underset{\mathbf{m},\alpha}{\text{minimize}} \sum_{k,l} \|\mathbf{P}_k \mathbf{A}_{k,l}(\mathbf{m})^{-1} \alpha_{k,l} \mathbf{e}_{k,l} - \mathbf{d}_{k,l}\|_2^2$$

Eliminate α w/ variable projection:

$$\overline{\alpha} = \arg\min_{\alpha} \sum_{k,l} \|\mathbf{P}_k \mathbf{A}_{k,l}(\mathbf{m})^{-1} \alpha_{k,l} \mathbf{e}_{k,l} - \mathbf{d}_{k,l}\|_2^2$$

WRI with source estimation

Triple parameters optimization problem:

$$\underset{\mathbf{u},\mathbf{m},\alpha}{\text{minimize}} \sum_{k,l} \|\mathbf{P}_k \mathbf{u}_{k,l} - \mathbf{d}_{k,l}\|_2^2 + \lambda^2 \|\mathbf{A}_{k,l}(\mathbf{m})\mathbf{u}_{k,l} - \alpha_{k,l}\mathbf{e}_{k,l}\|_2^2$$

WRI with source estimation

Triple parameters optimization problem:

$$\underset{\mathbf{u},\mathbf{m},\alpha}{\text{minimize}} \sum_{k,l} \|\mathbf{P}_k \mathbf{u}_{k,l} - \mathbf{d}_{k,l}\|_2^2 + \lambda^2 \|\mathbf{A}_{k,l}(\mathbf{m})\mathbf{u}_{k,l} - \alpha_{k,l}\mathbf{e}_{k,l}\|_2^2$$

Eliminate $\mathbf{u}$ and α jointly w/ variable projection:

$$[\overline{\mathbf{u}}, \overline{\alpha}] = \arg\min_{\mathbf{u},\alpha} \sum_{k,l} \|\mathbf{P}_k \mathbf{u}_{k,l} - \mathbf{d}_{k,l}\|_2^2 + \lambda^2 \|\mathbf{A}_{k,l}(\mathbf{m})\mathbf{u}_{k,l} - \alpha_{k,l}\mathbf{e}_{k,l}\|_2^2$$

WRI with source estimation

Corresponds to solving the following augmented system:

$$\begin{pmatrix} \lambda \mathbf{A}_{k,l} & -\lambda \mathbf{e}_{k,l} \\ \mathbf{P}_k & 0 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{u}}_{k,l} \\ \overline{\alpha}_{k,l} \end{pmatrix} = \begin{pmatrix} 0 \\ \mathbf{d}_{k,l} \end{pmatrix}$$

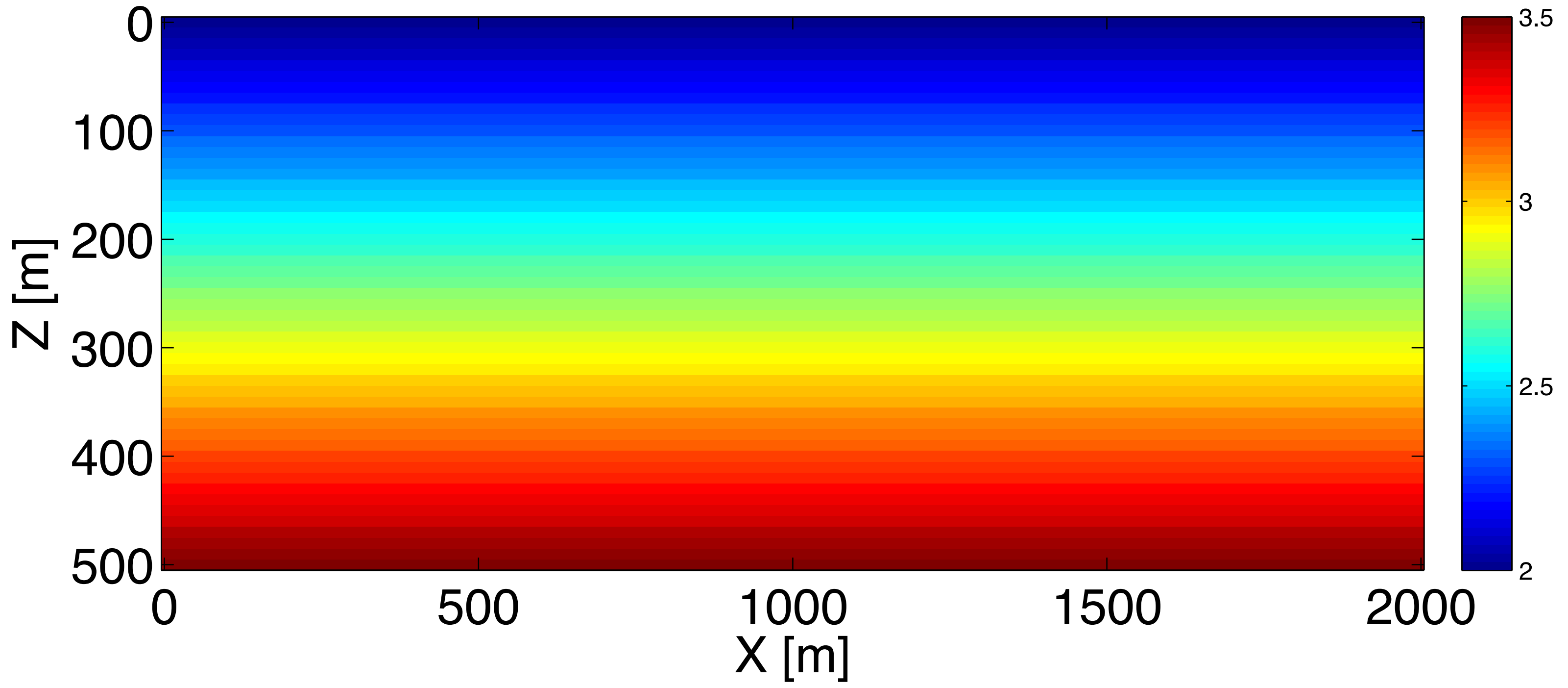
Cf. original augmented system:

Full column rank!

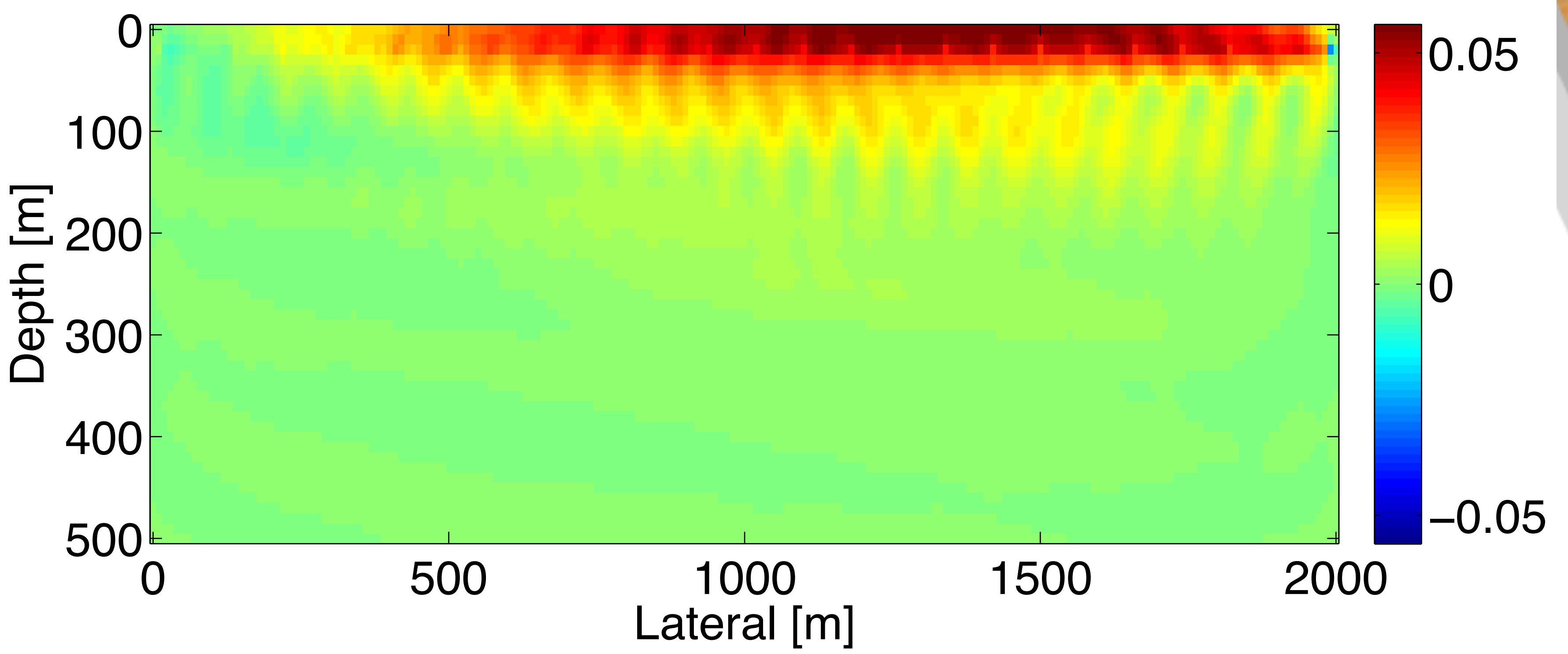
No additional computational cost!

$$\begin{pmatrix} \lambda \mathbf{A}_{k,l} \\ \mathbf{P}_k \end{pmatrix} \overline{\mathbf{u}}_{k,l} = \begin{pmatrix} \lambda \mathbf{q}_{k,l} \\ \mathbf{d}_{k,l} \end{pmatrix}$$

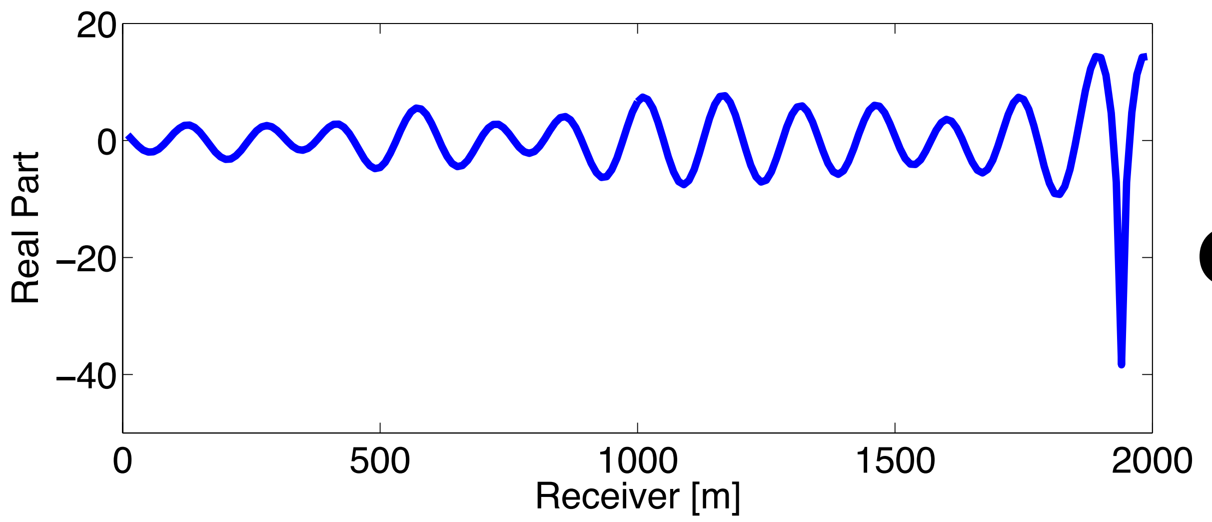
WRI with source estimation



$\mathbf{m}$



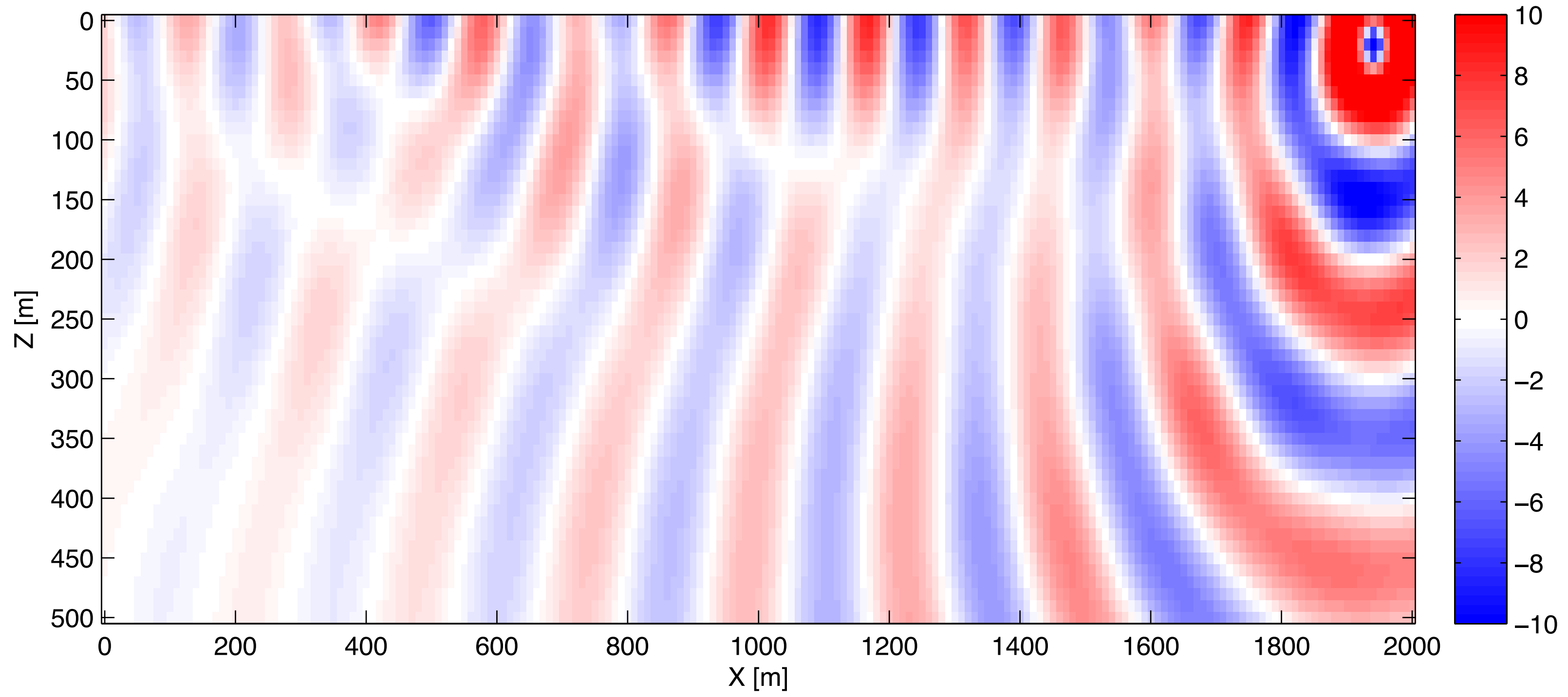
$\mathbf{g}$



$\mathbf{d_{obs}}$

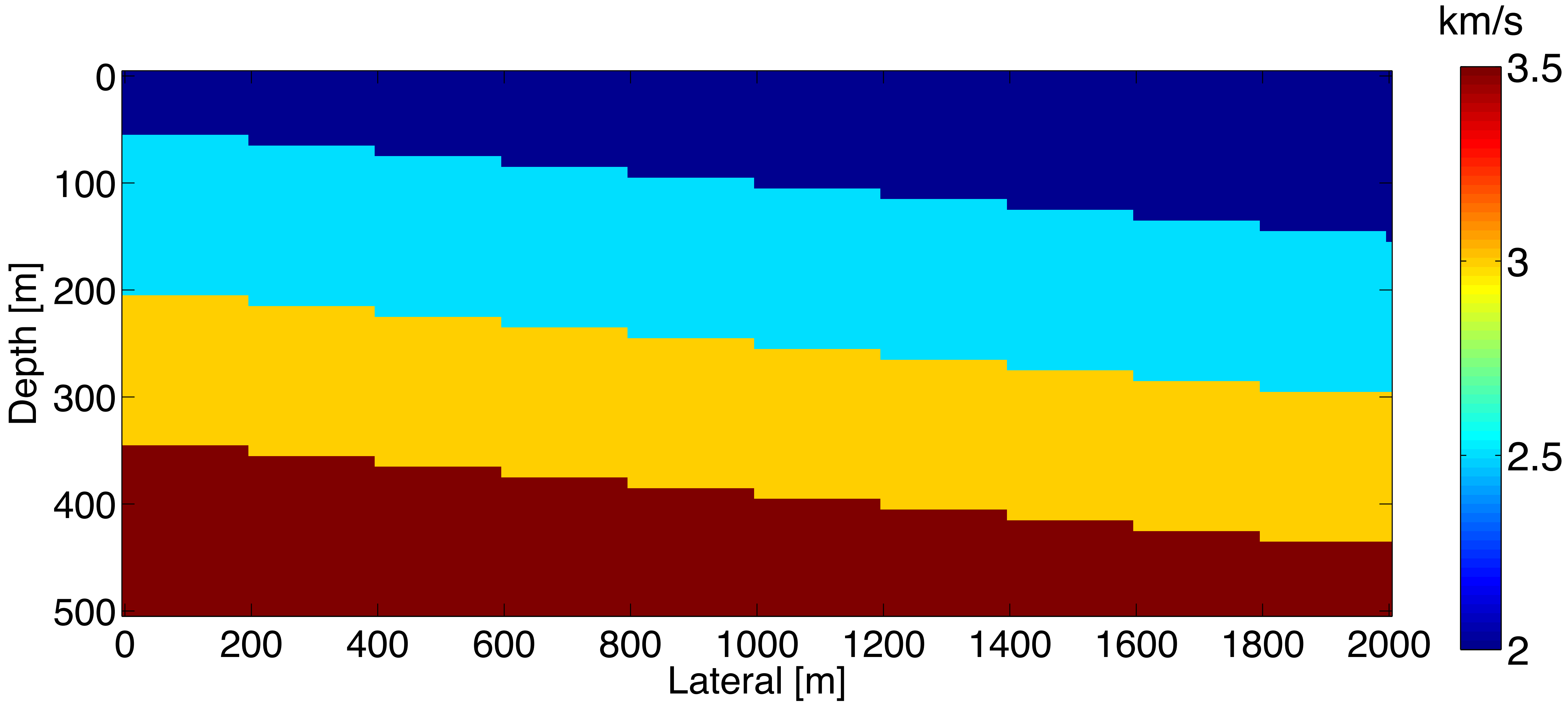
$$\begin{pmatrix} \lambda \mathbf{A} & -\lambda \mathbf{e} \\ \mathbf{P} & 0 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{u}} \\ \overline{\alpha} \end{pmatrix} = \begin{pmatrix} 0 \\ \mathbf{d_{obs}} \end{pmatrix}$$

$\overline{\alpha}$ **and**

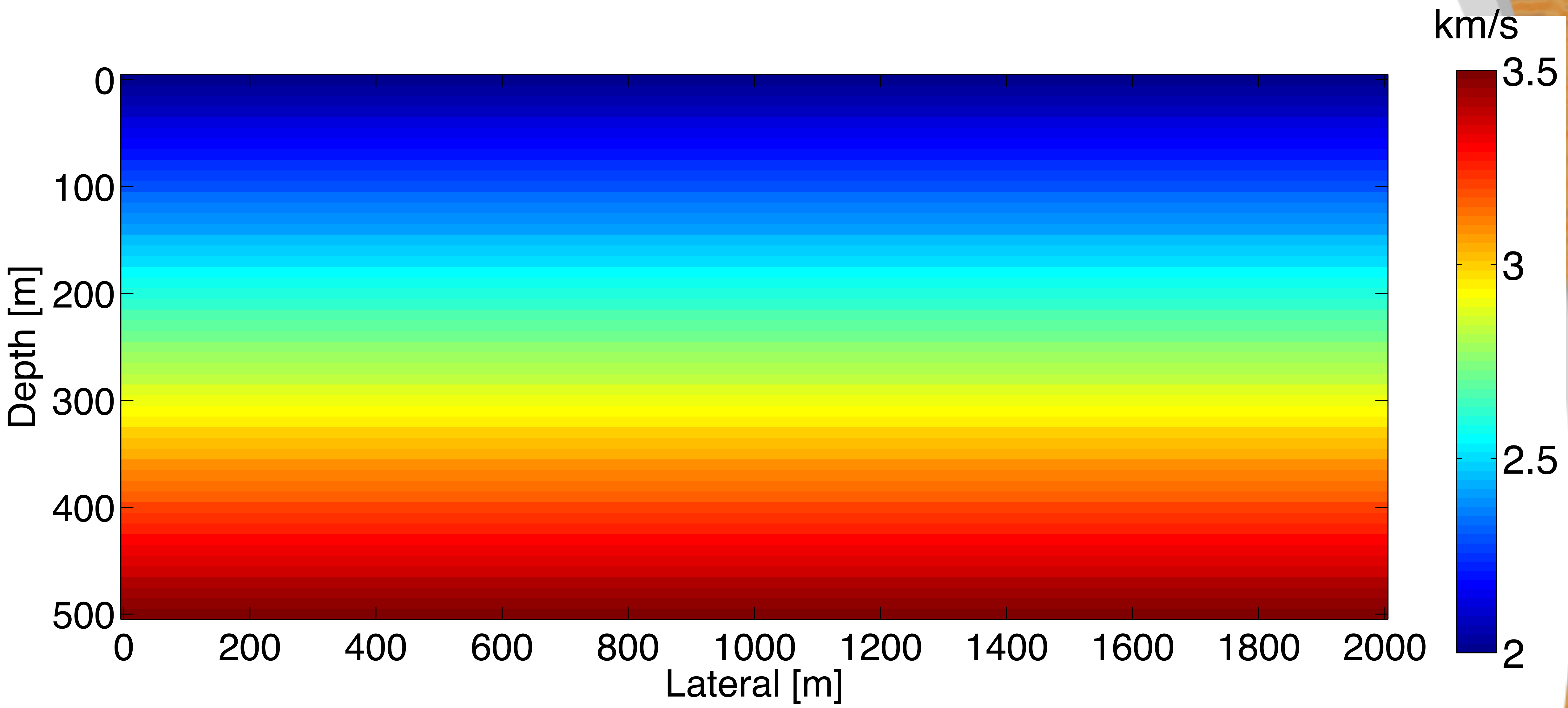


$\overline{\mathbf{u}}$

Synthetic example

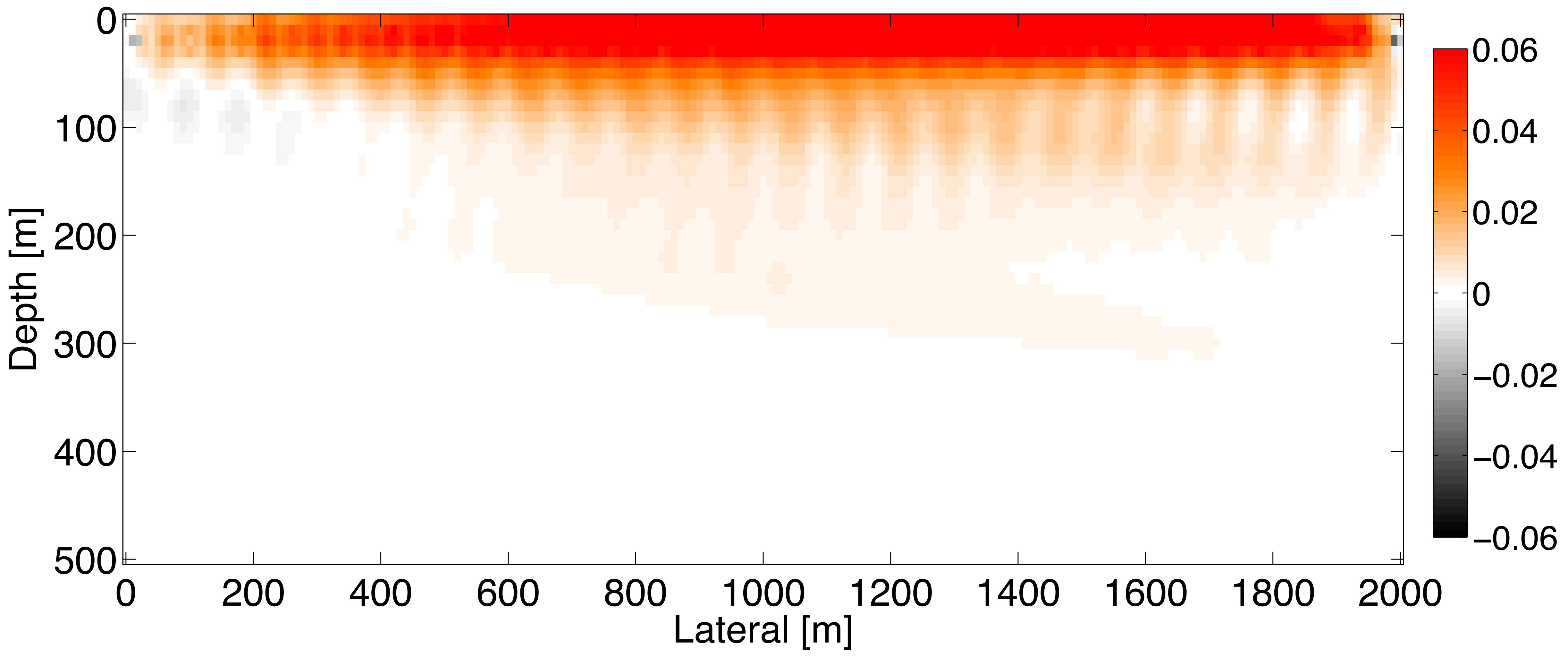


True Model

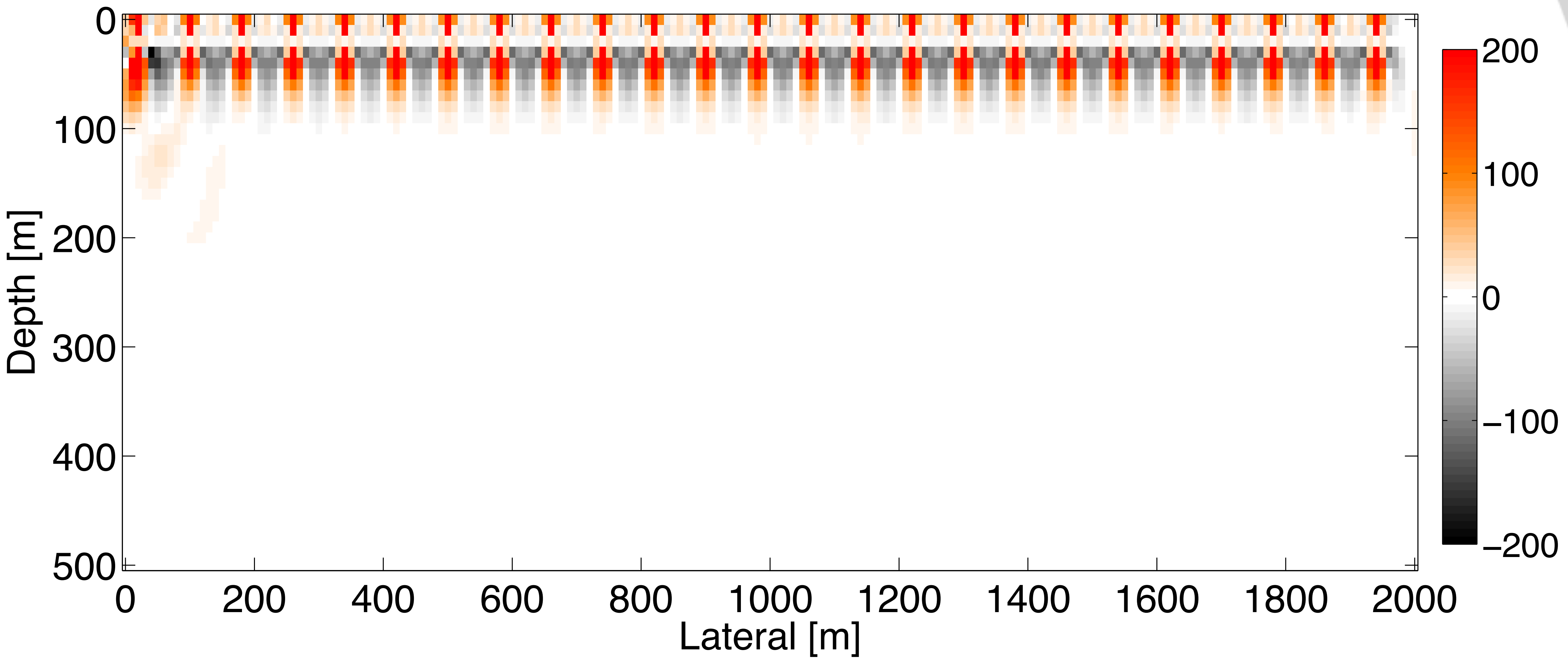


Initial Model

Gradient comparison

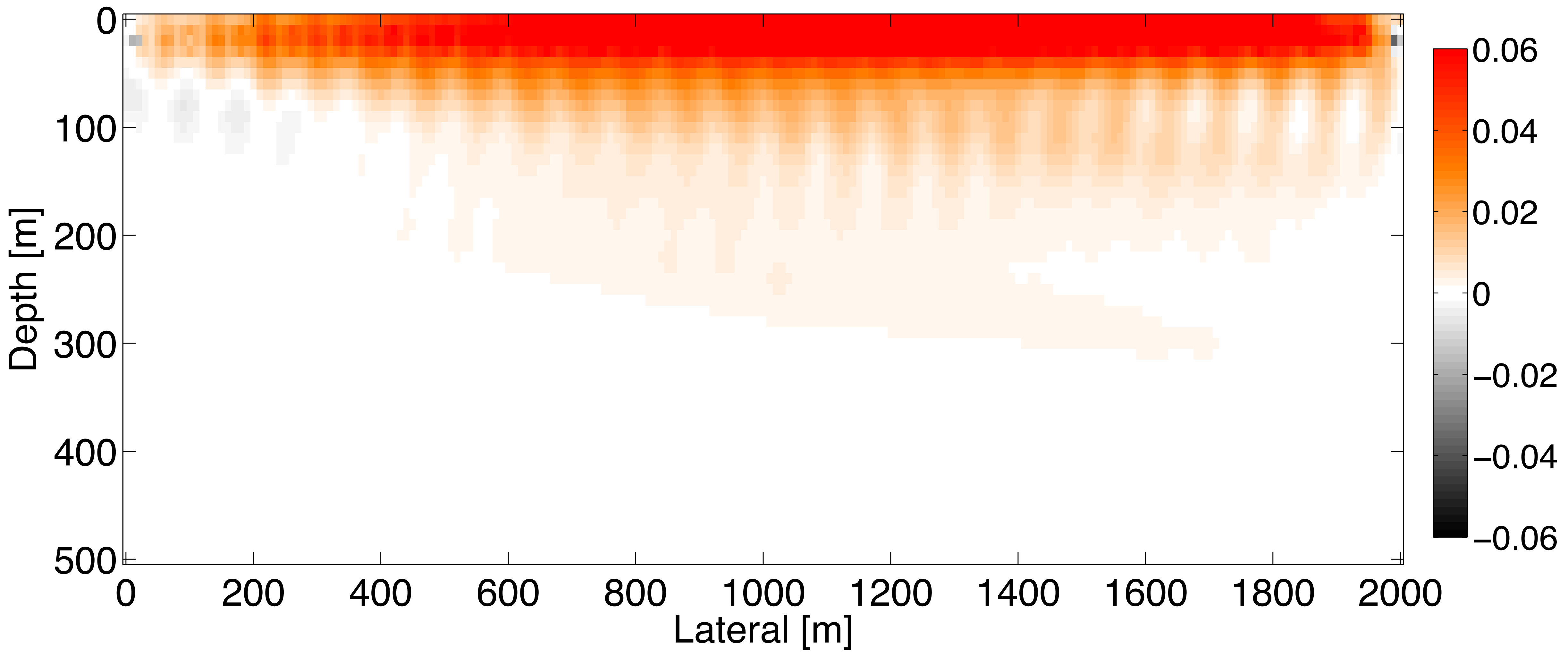


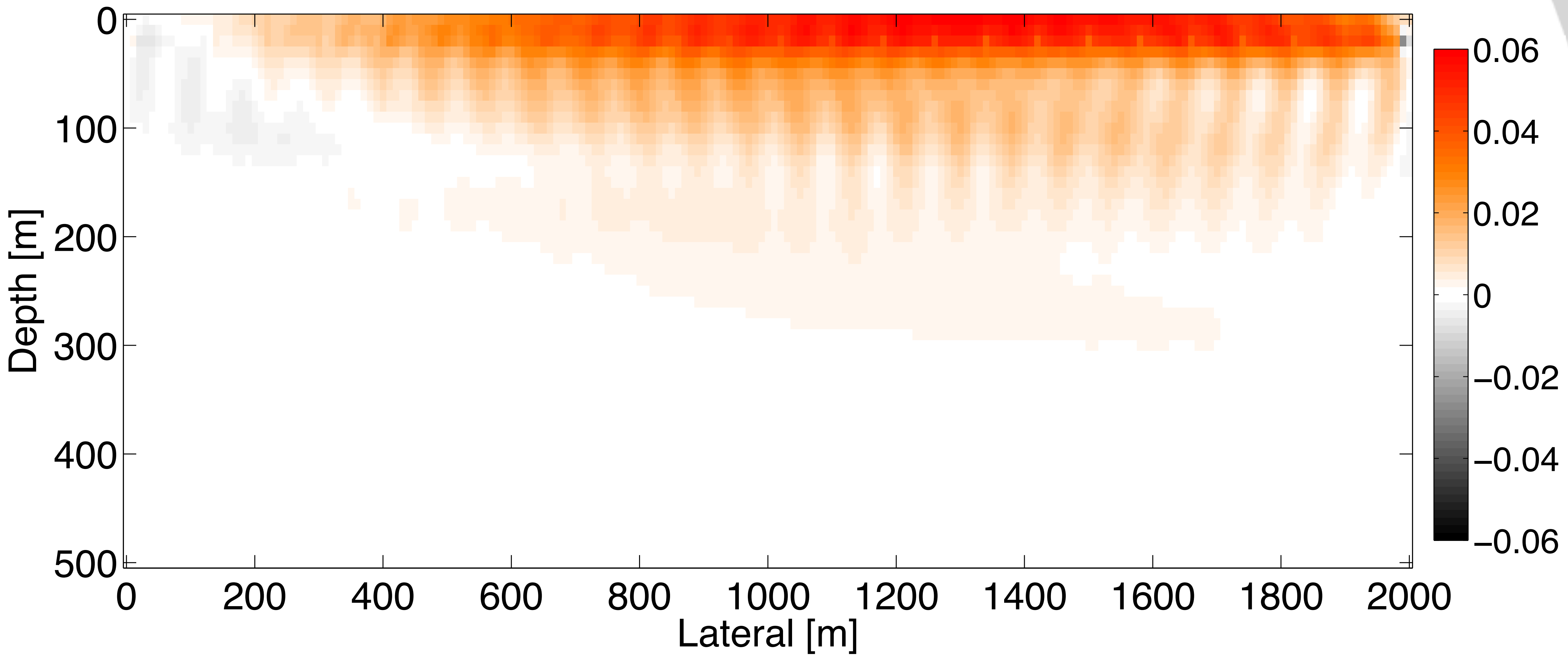
Gradient with true source wavelet



Gradient with wrong source wavelet

Gradient comparison

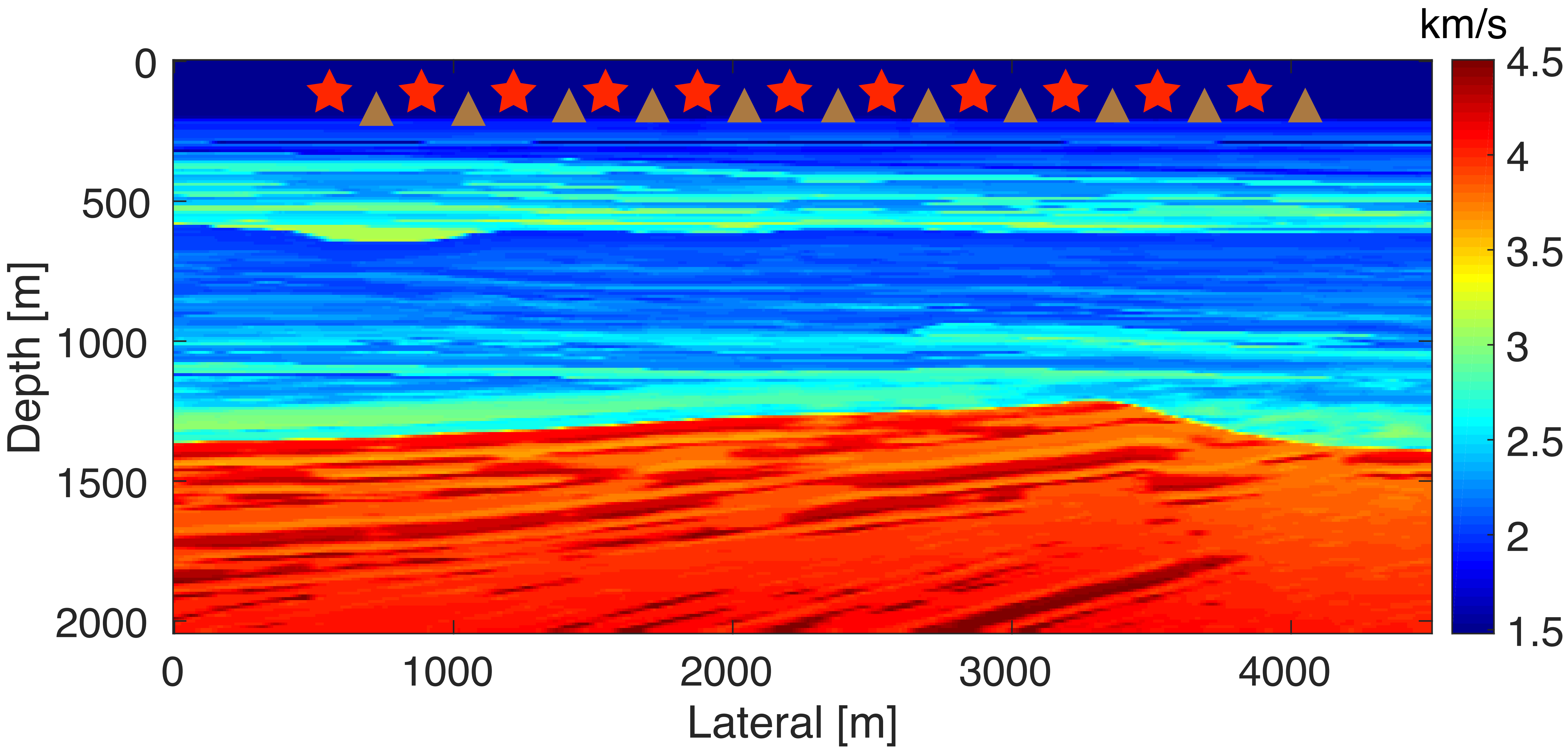




Gradient with true source wavelet

Gradient with estimated source wavelet

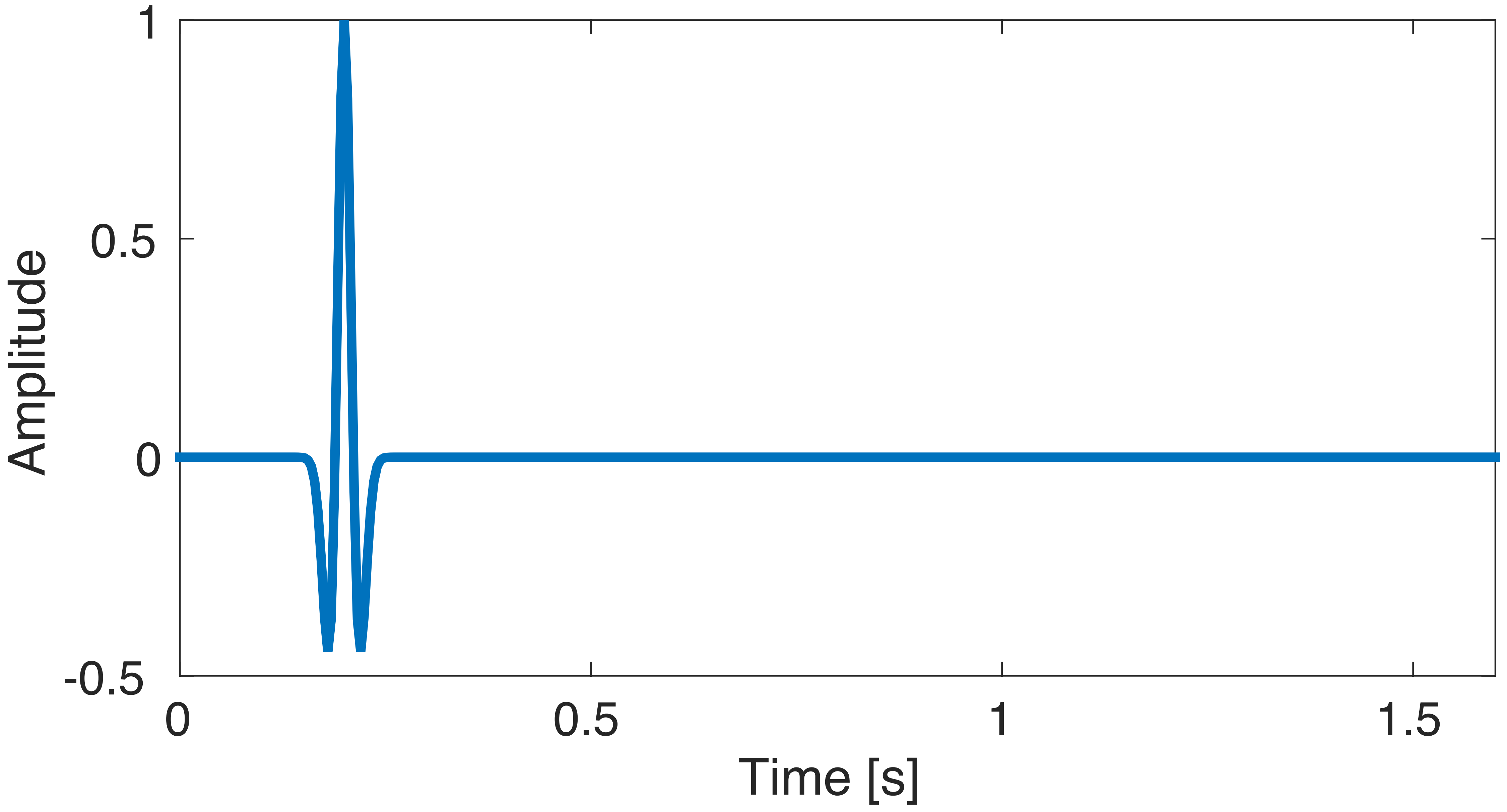
BG model



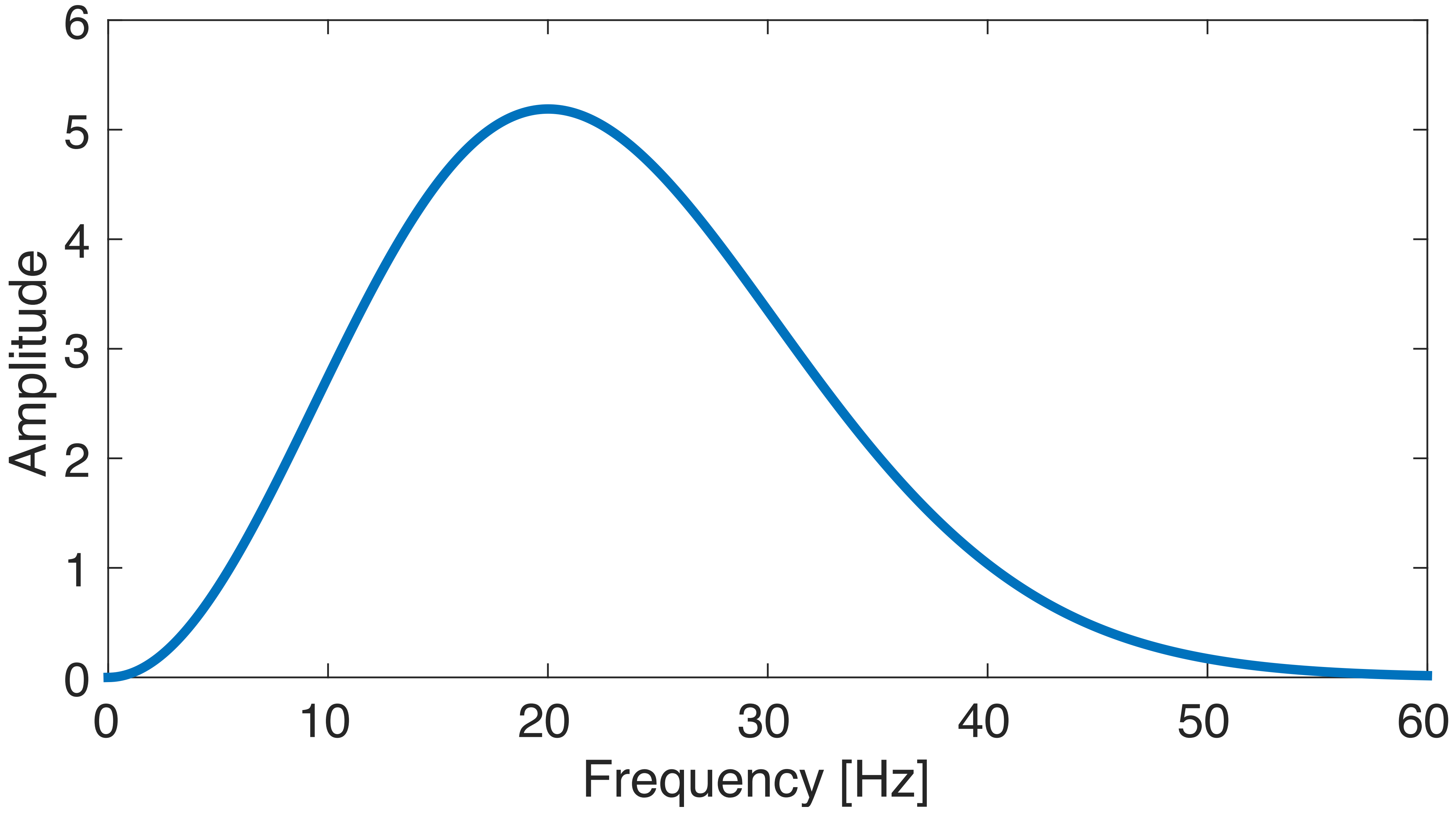
Modeling information:
Model size: 2000m x 4500m
Source spacing: 50m
Receiver spacing: 10m
Fixed spread 4.5km
Frequency : 2~31 Hz

Inversion information:
Optimization Solver: Gauss-Newton
Iterations per frequency band: 21
Batch size: 15

Source wavelet

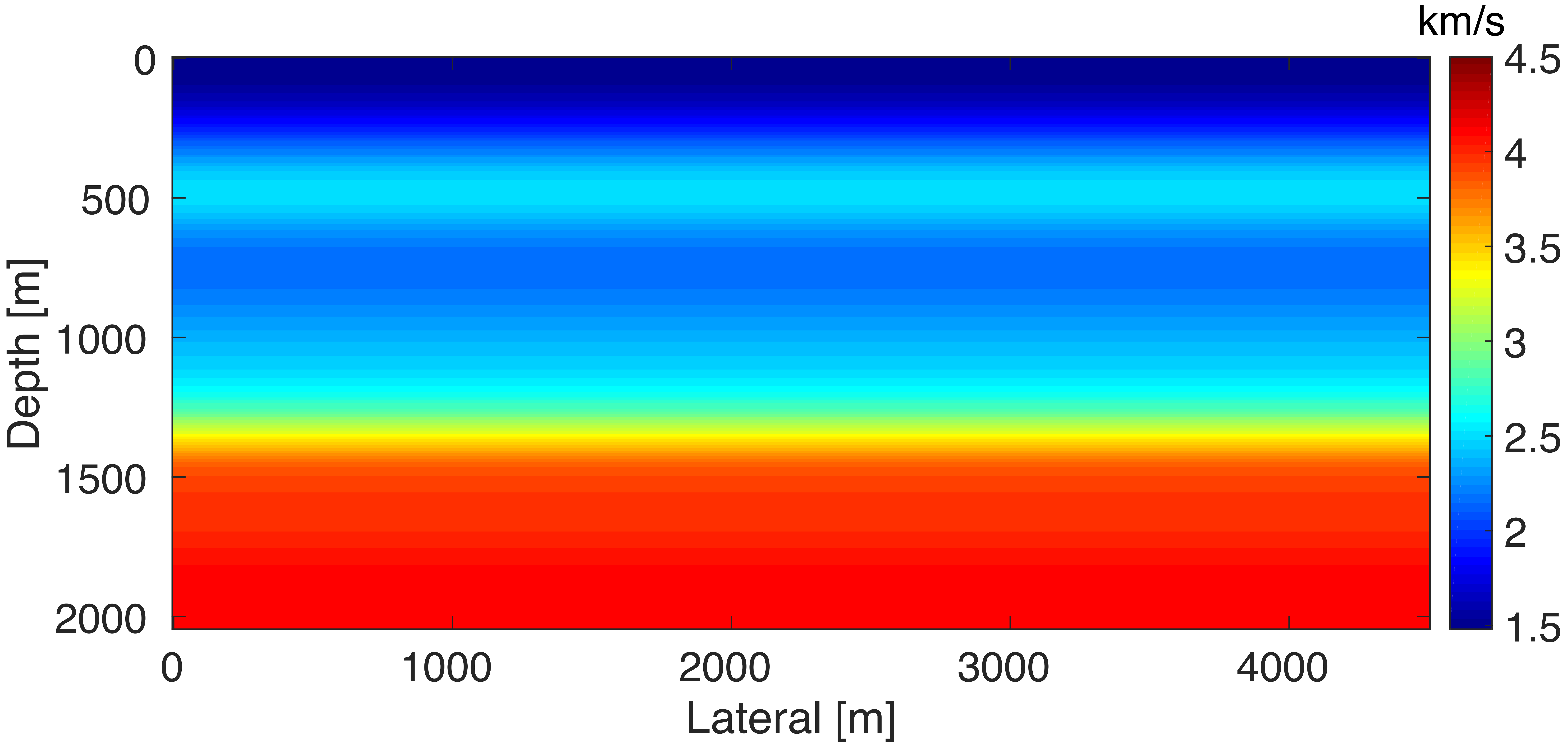


Source Wavelet

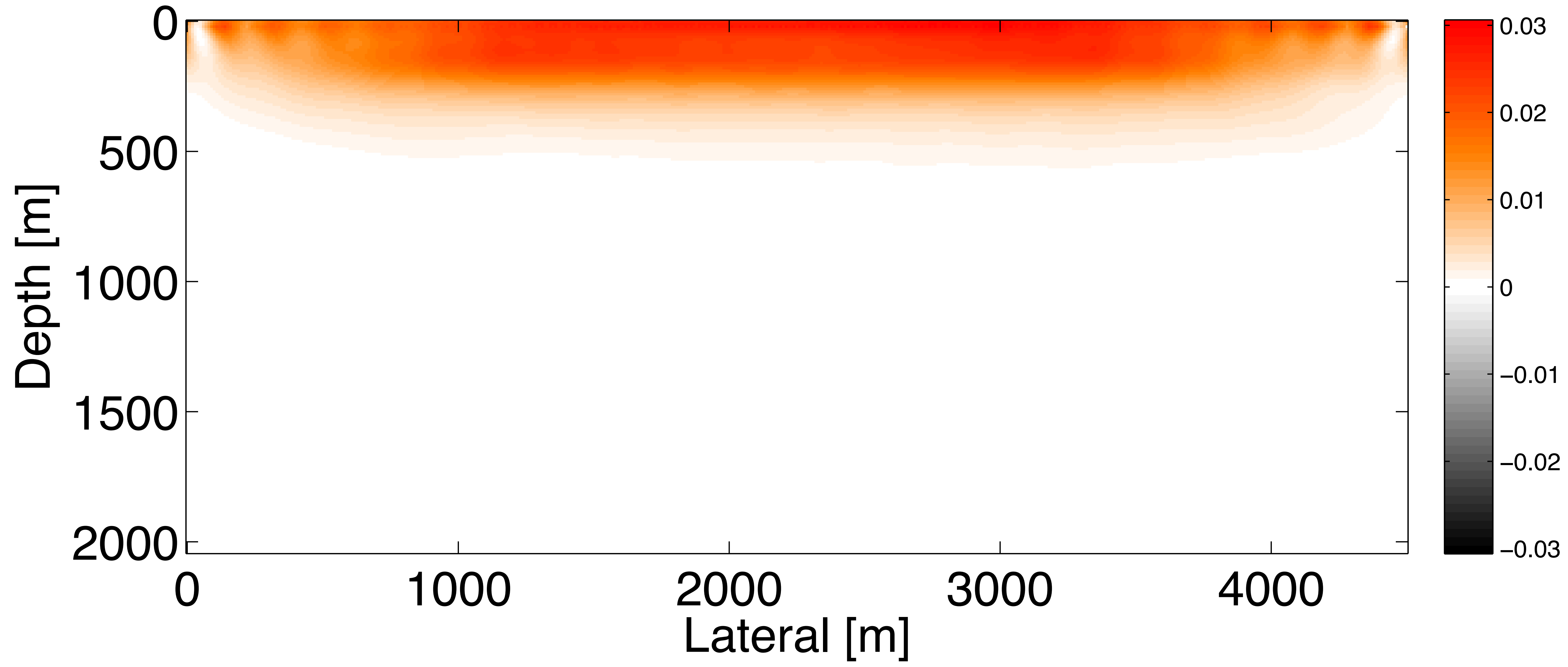


Spectrum

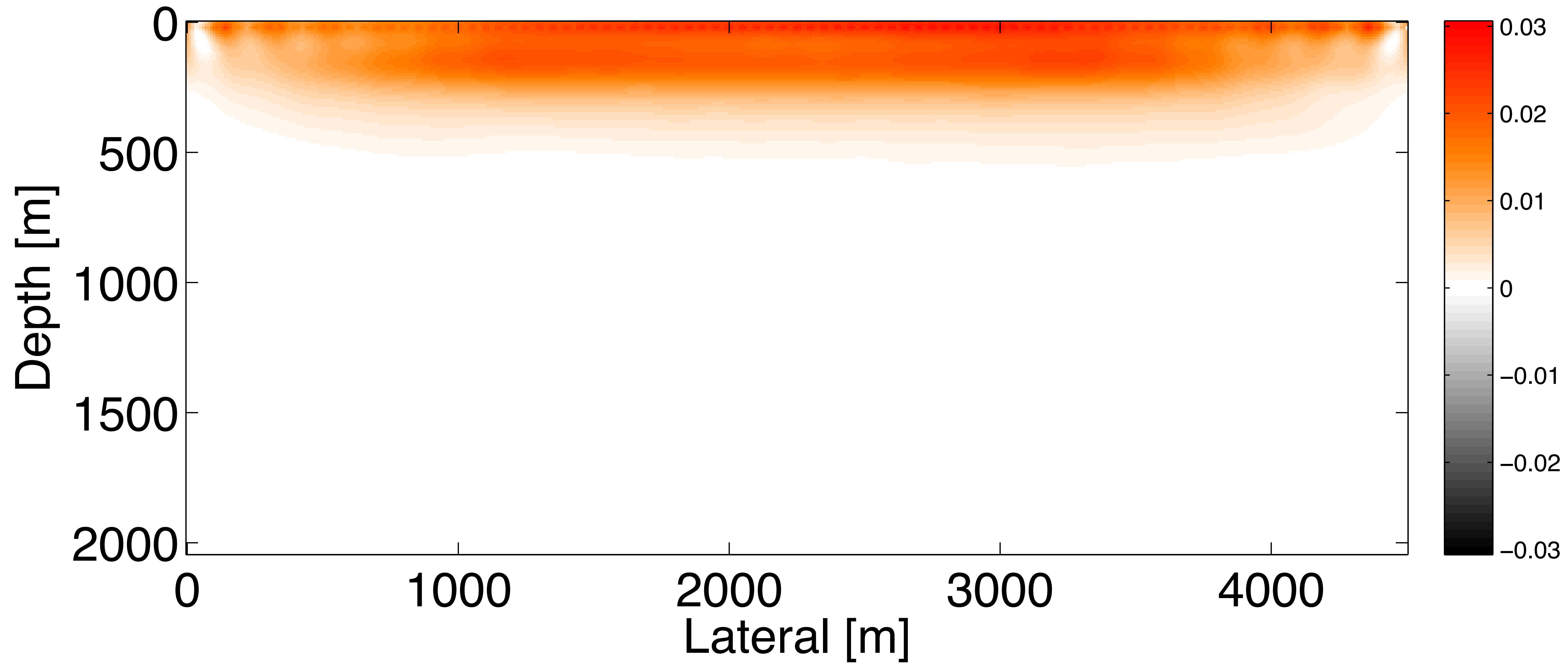
Initial model



First gradient comparison

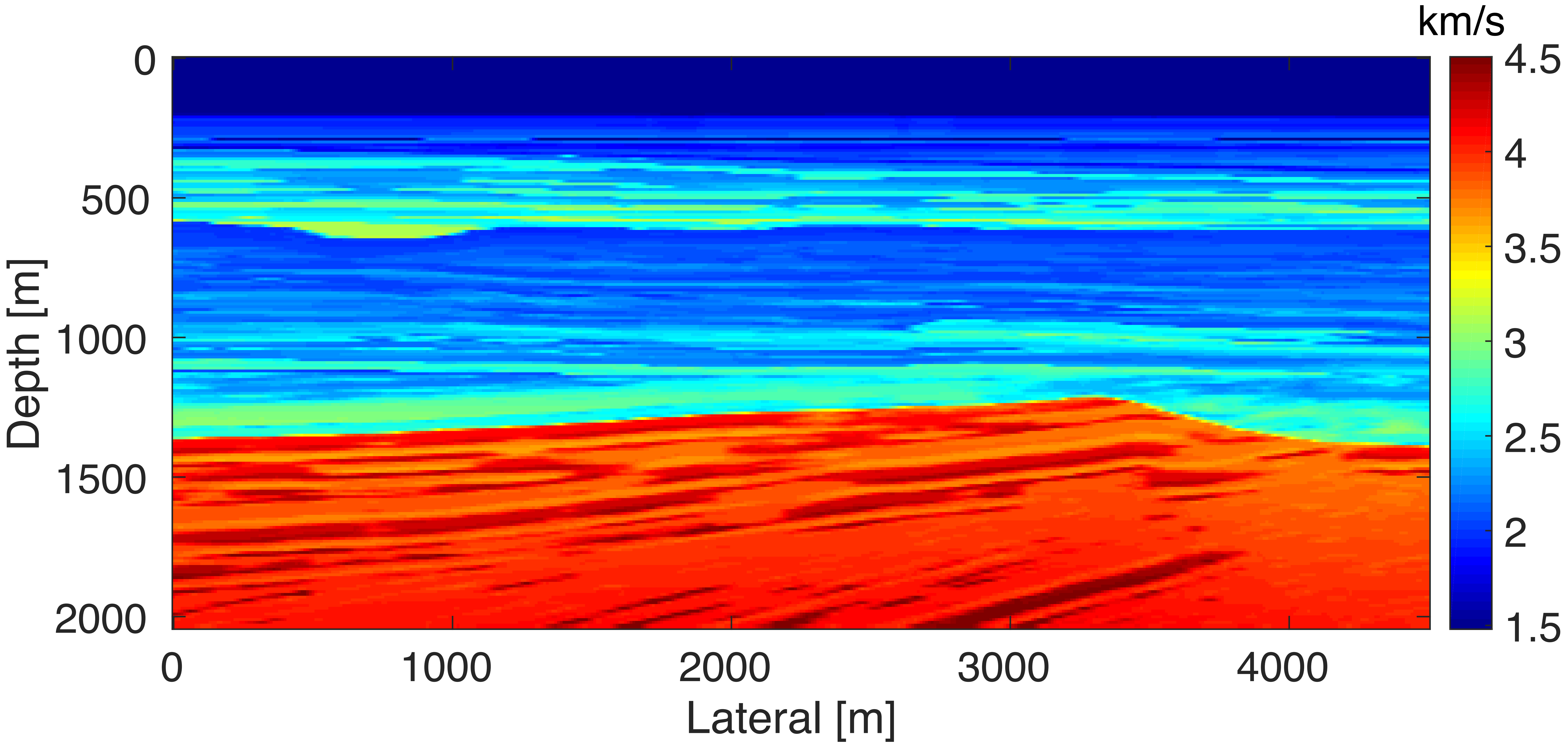


Gradient with true source wavelet

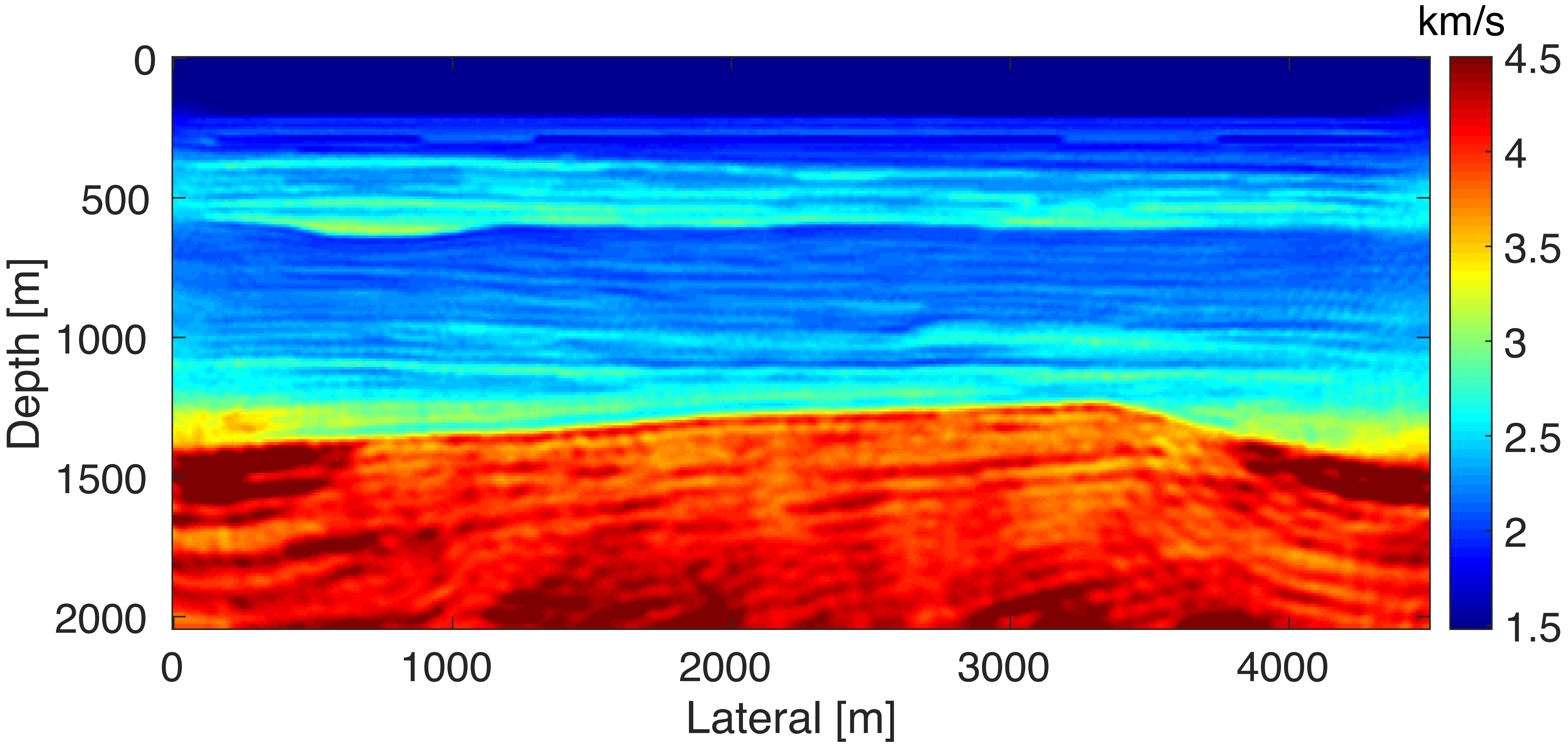


Gradient with estimated source wavelet

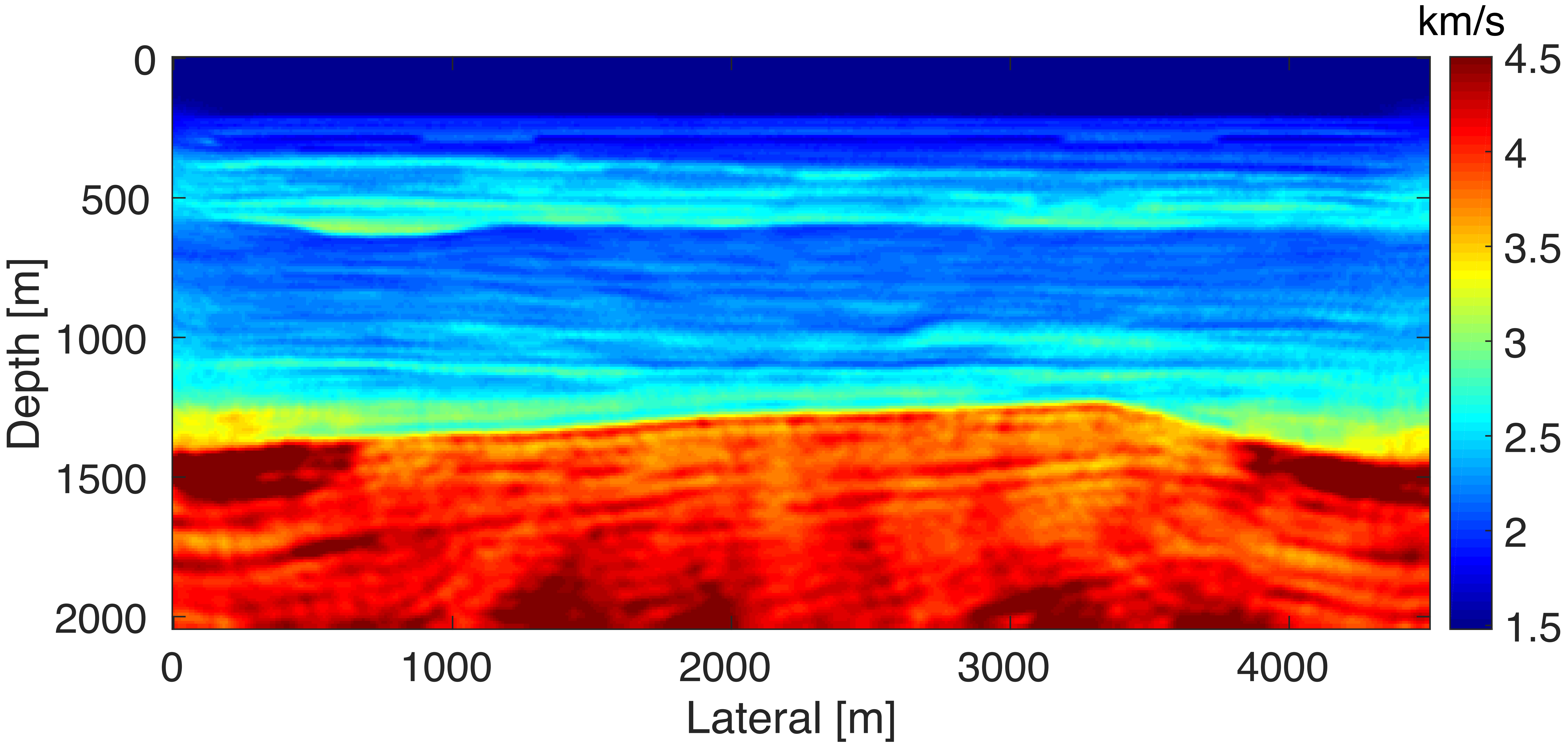
True Model


km/s
4.5
4
3.5
3
2.5
2
1.5
0
500
1000
1500
2000
Depth [m]
0
1000
2000
3000
4000
Lateral [m]

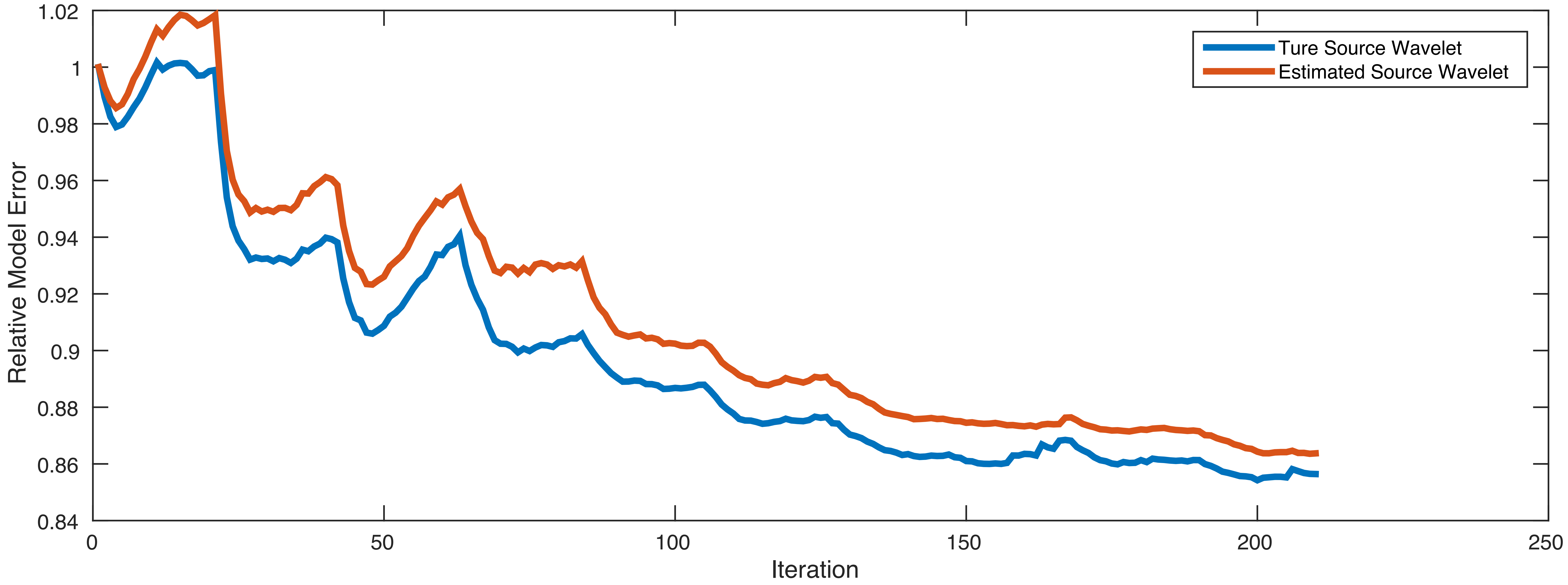
Result with true source wavelet



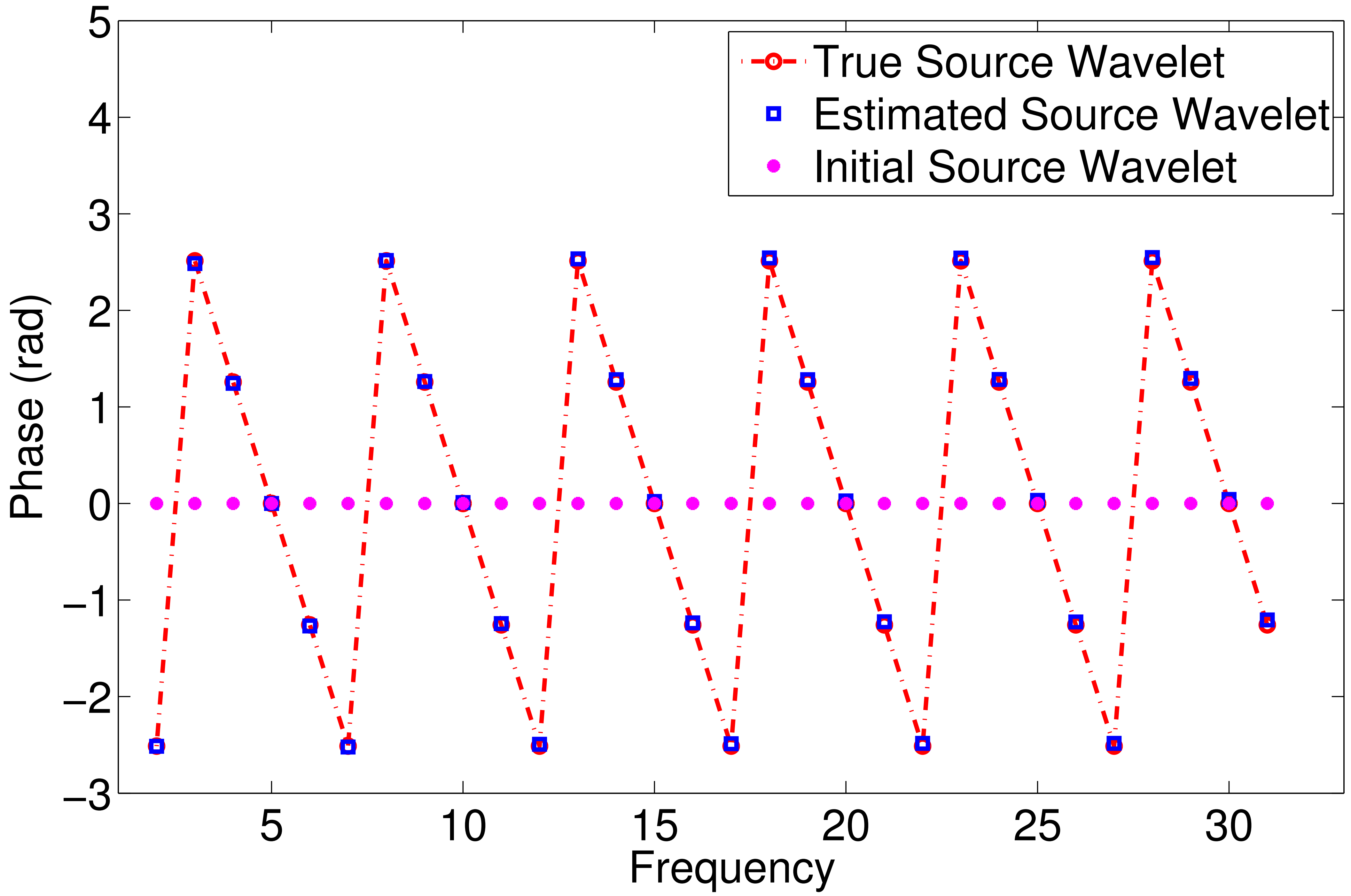
Result with estimated source wavelet



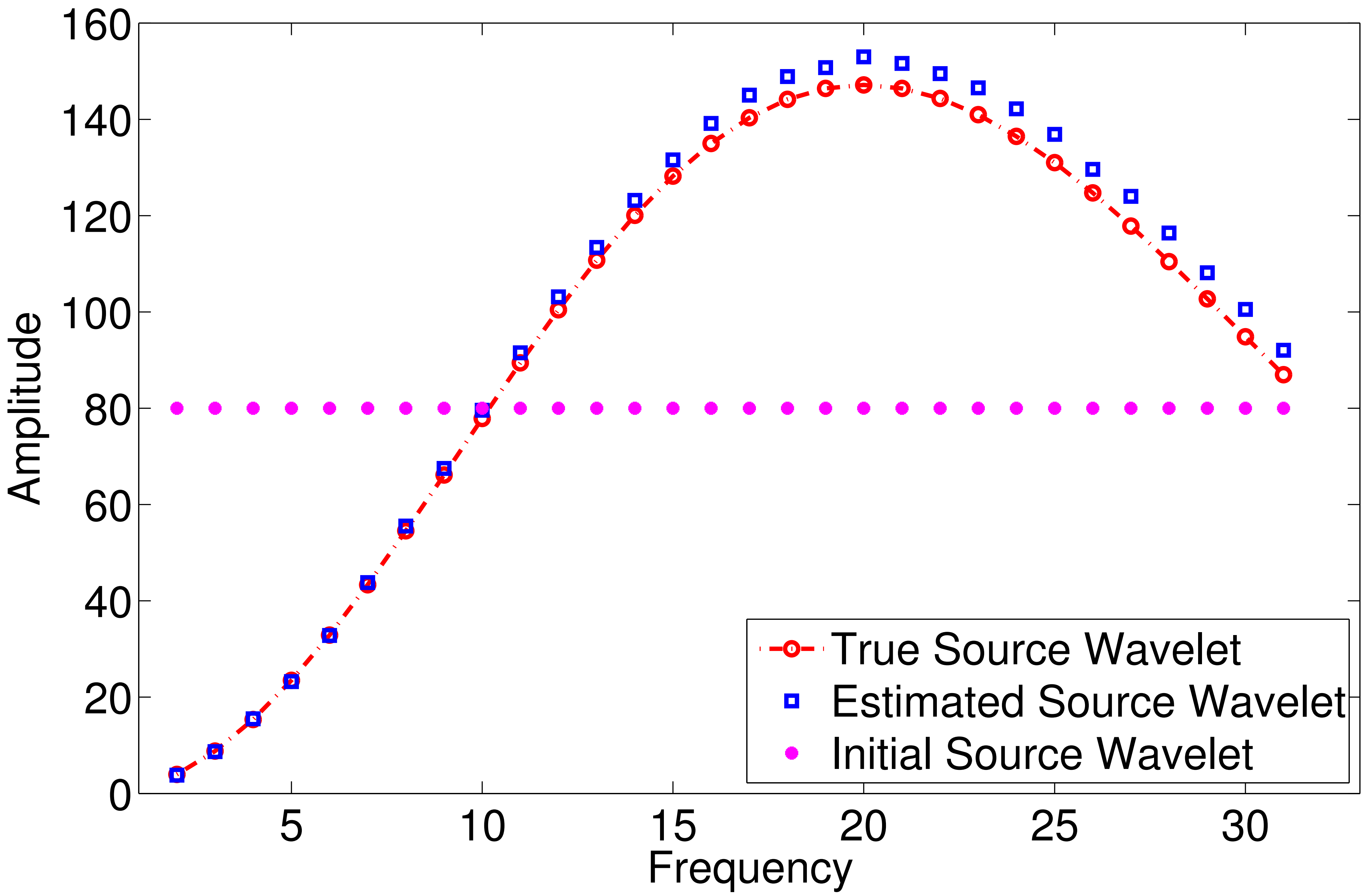
Relative model-error comparison



Source wavelet comparison

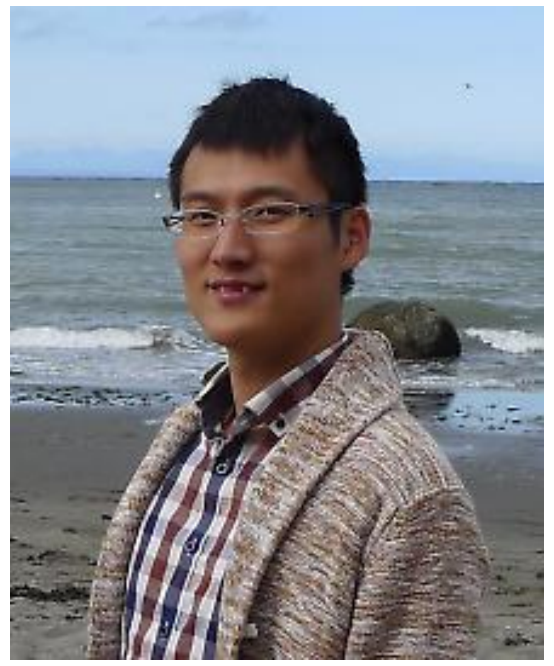


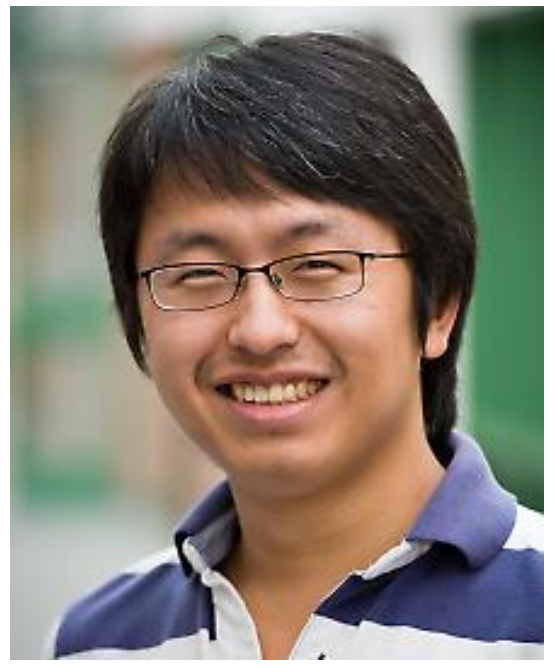
Phase



Amplitude

Chevron blind test data


Zhilong Fang


Xiang Li

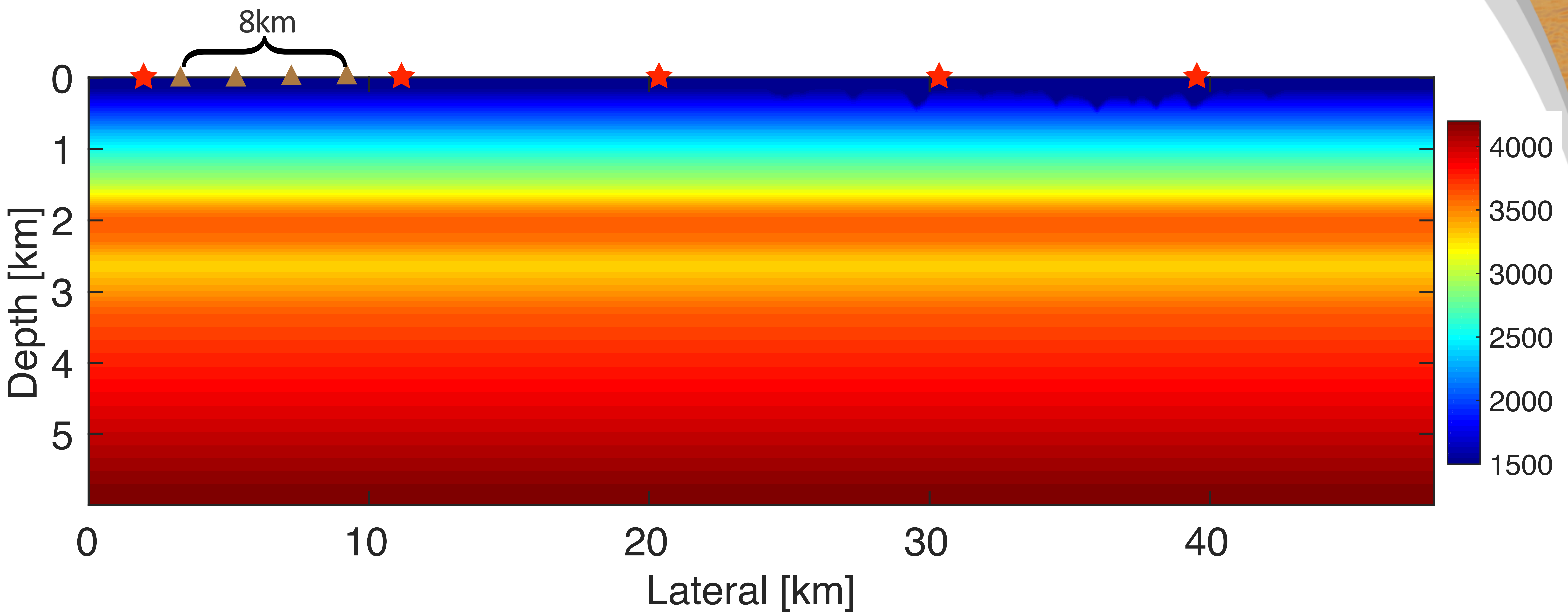

Bas Peters


Brendan Smithyman

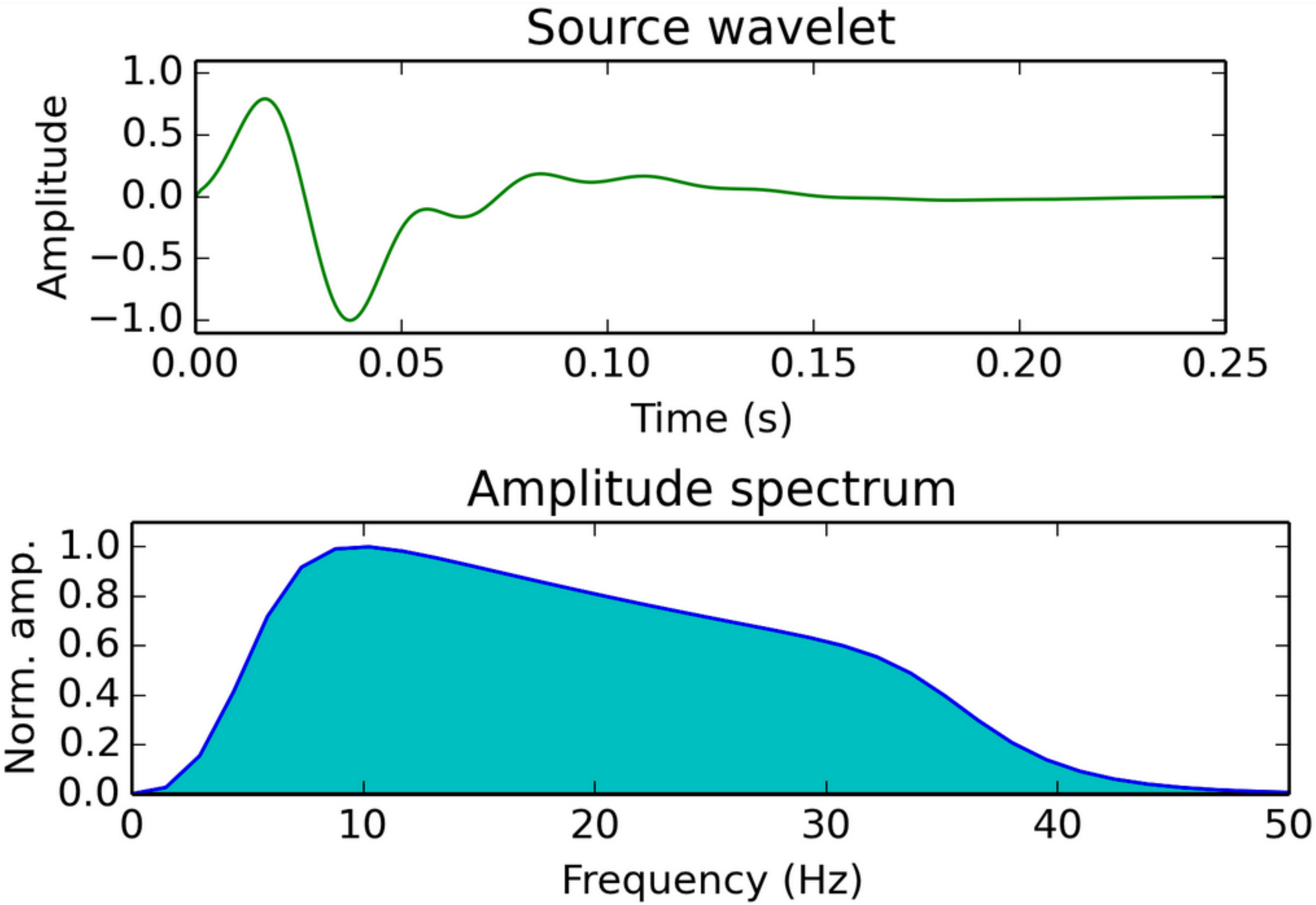

Mengmeng Yang


Felix J. Herrmann

Chevron blind test data



Chevron blind test data



Chevron blind test data

Data-set information:

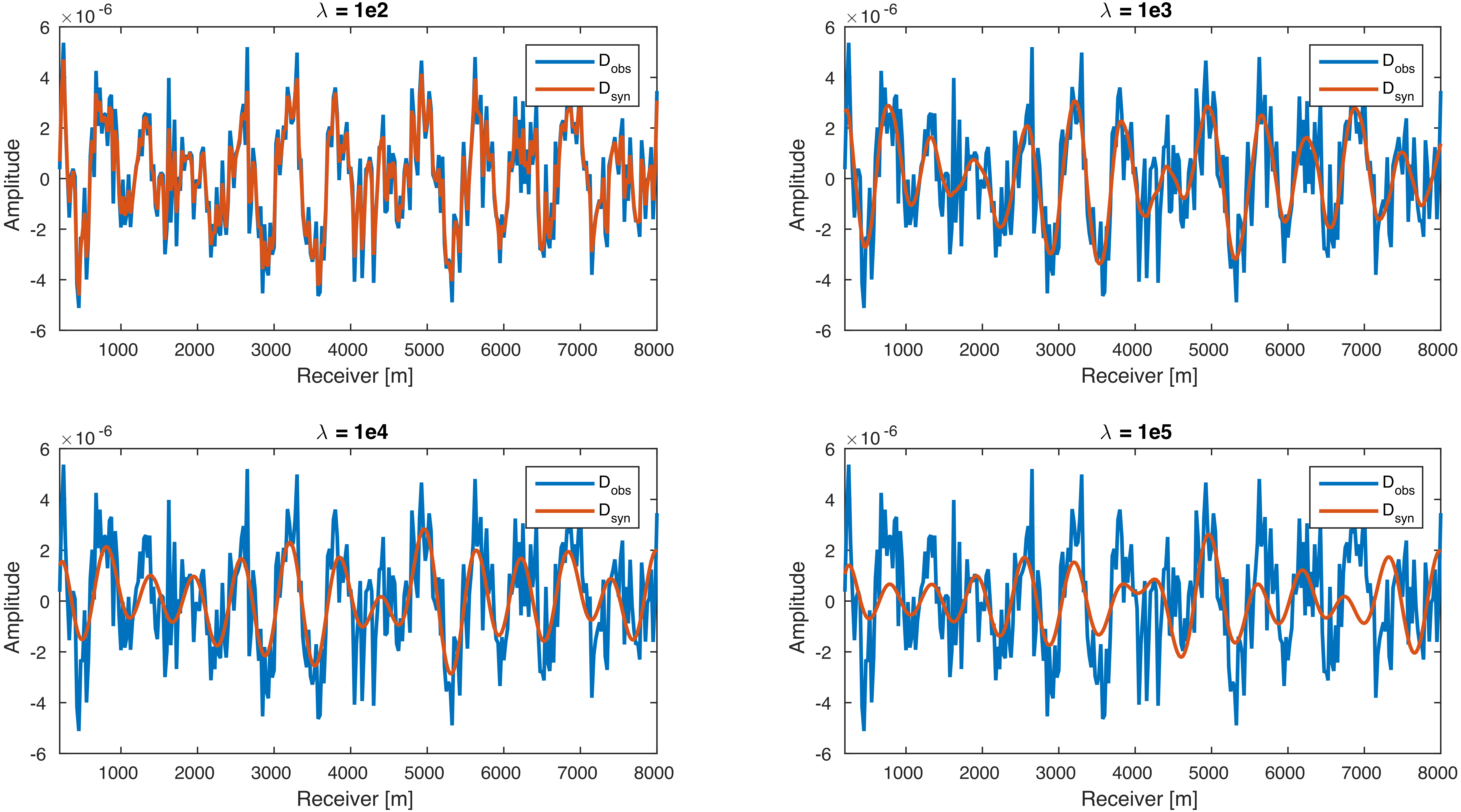
1. 1600 shots: ds = 25 m, Source depth = 15 m;
2. 321 hydrophone recs/shot: dr = 25 m, Receiver depth = 15 m;
3. Maximum offset = 8000 m;
4. Record time = 8.0 s, sample rate 4 ms;
5. Vp water = constant = 1510 m/s;
6. With free surface multiples present in the data;
7. Isotropic Elastic.

Chevron blind test data

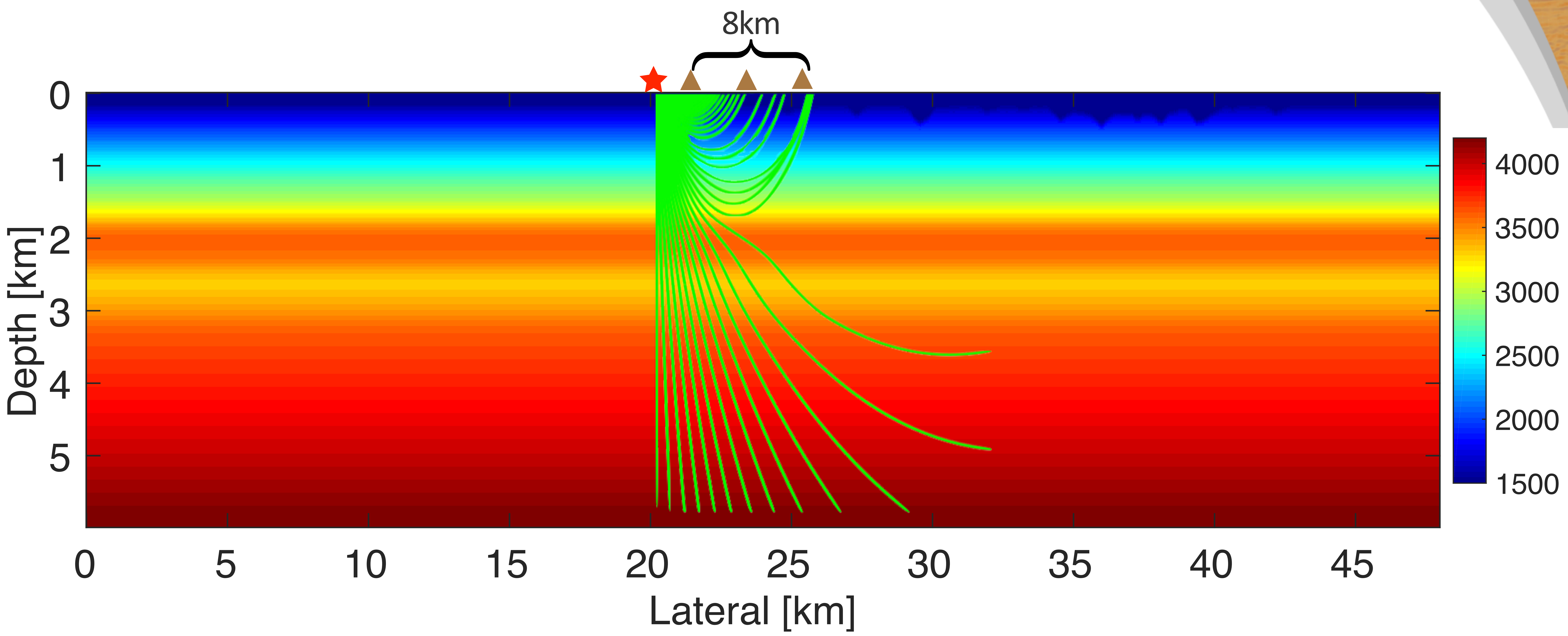
Inversion strategy:

1. Frequency domain WRI with Source estimation;
2. Frequency bands: [3:0.2:5]Hz, [3:0.2:7]Hz, [3:0.2:9]Hz, [3:0.2:11]Hz;
3. Batch sizes of random frequency subsets: 3, 6, 10, 10;
4. Batch size of random source subsets: 300;
5. Optimization solver: l-BFGS with 30 iterations per frequency band;
6. 2passes of WRI from frequency 3-11 Hz;
7. Grid size: 20m;
8. Minimum offset used: 1000m;
9. No pre-processing !!!

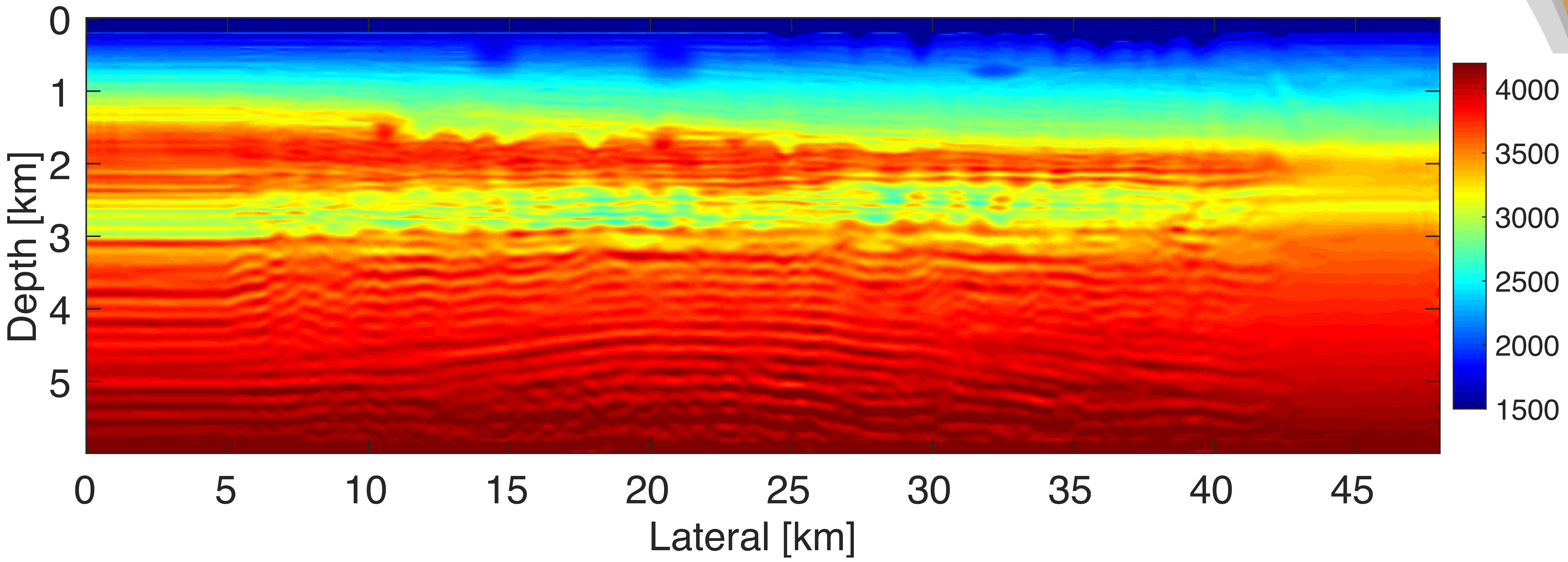
Data comparison
— 3 Hz Data of 800th shot



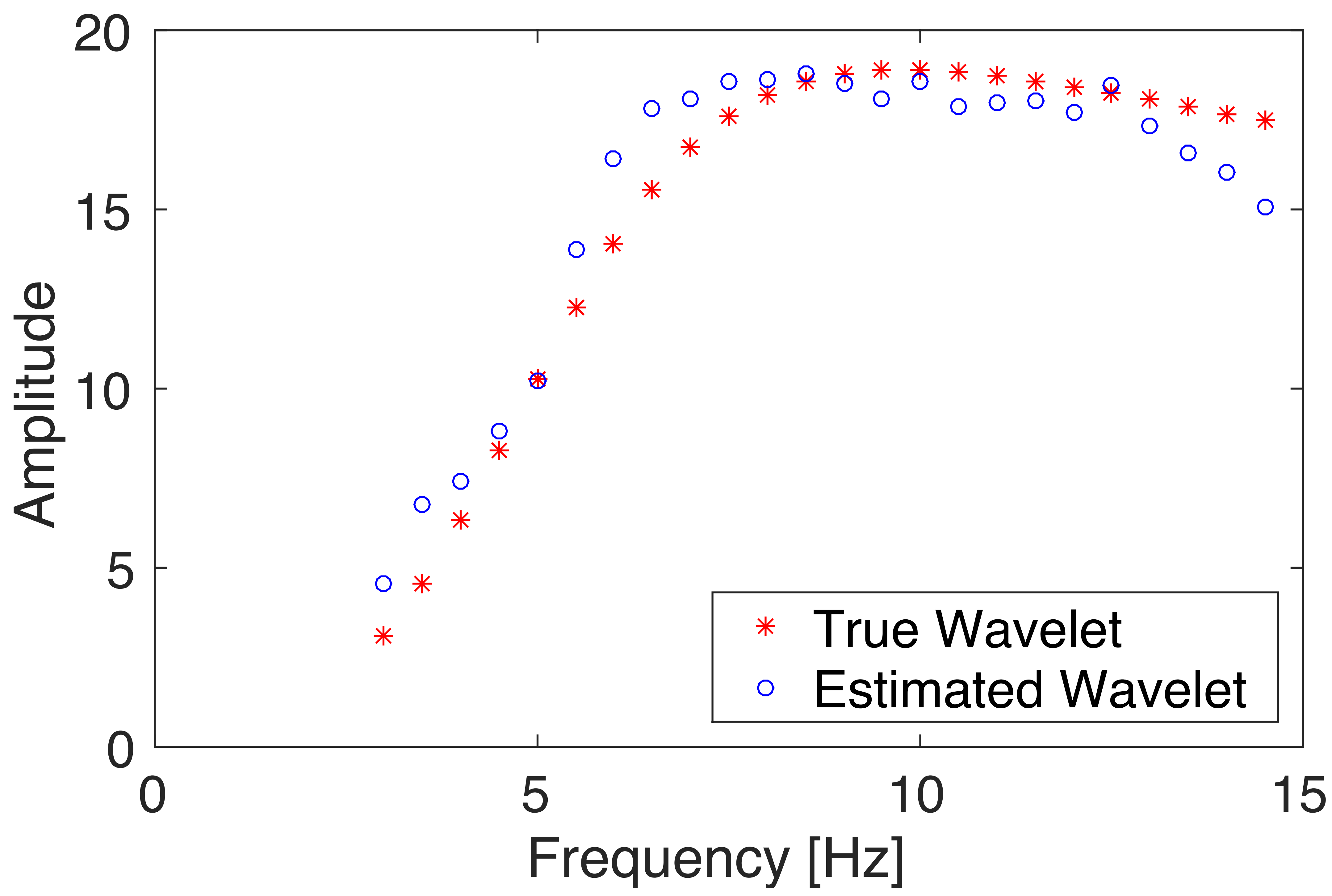
Initial model



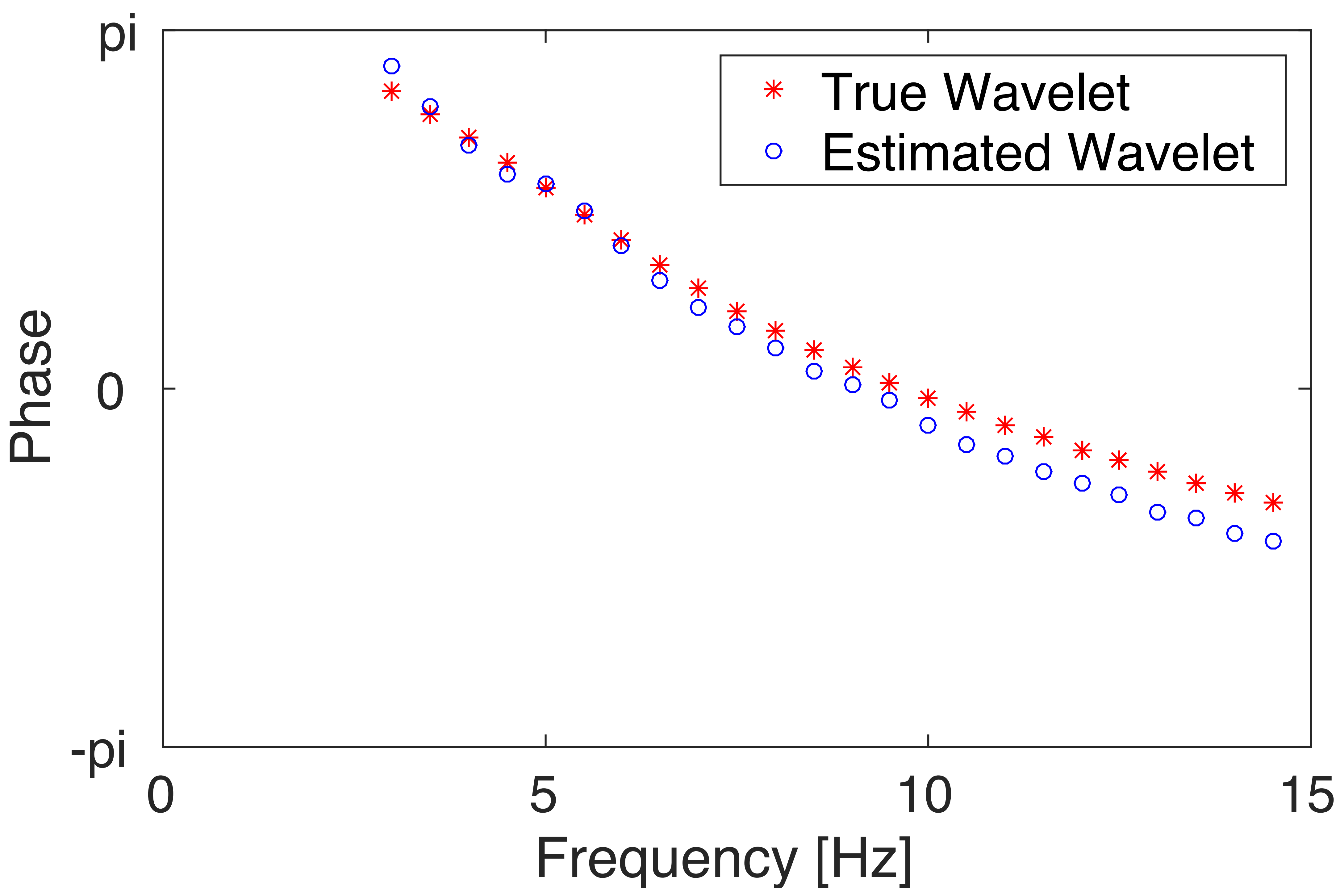
Inversion result



Source wavelet comparison

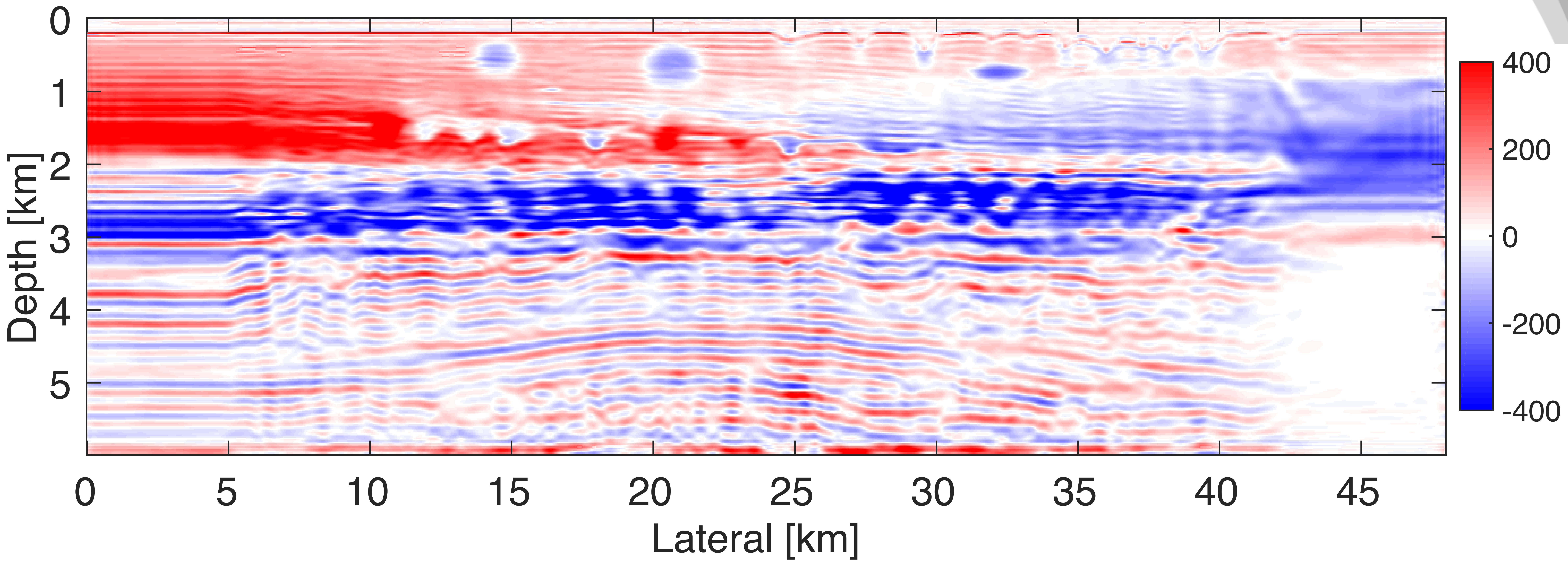


Amplitude

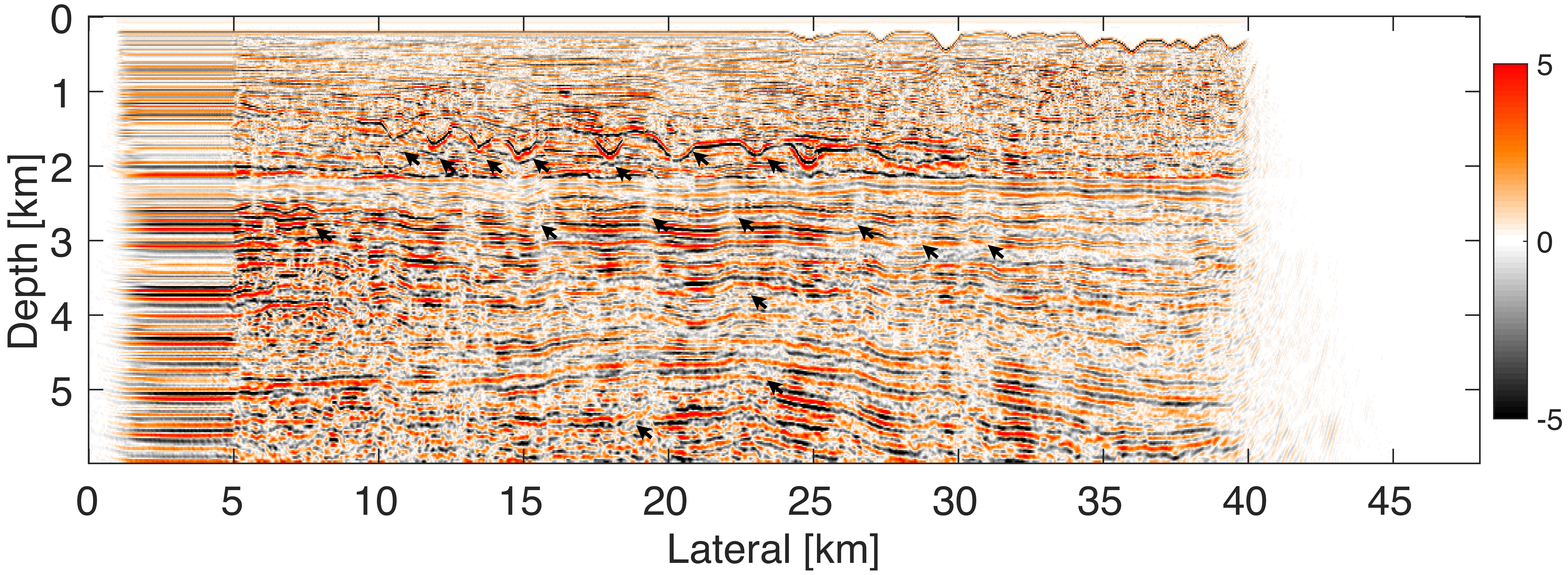


Phase

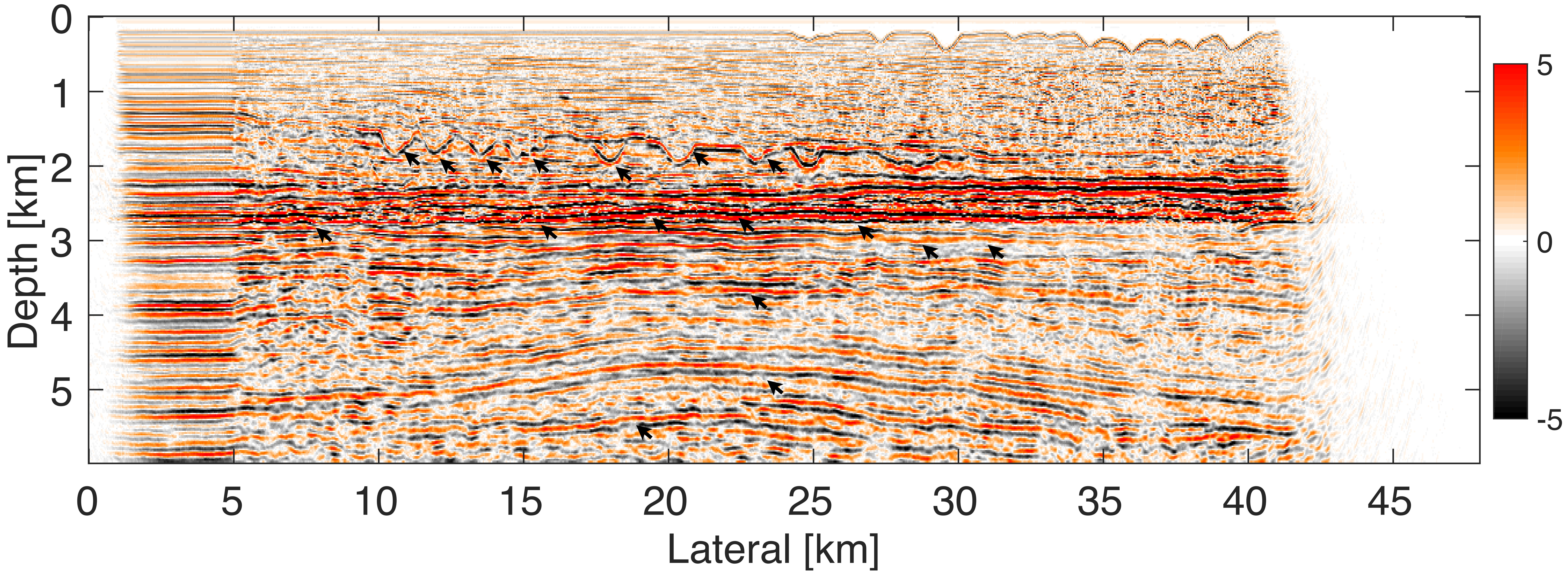
Model update



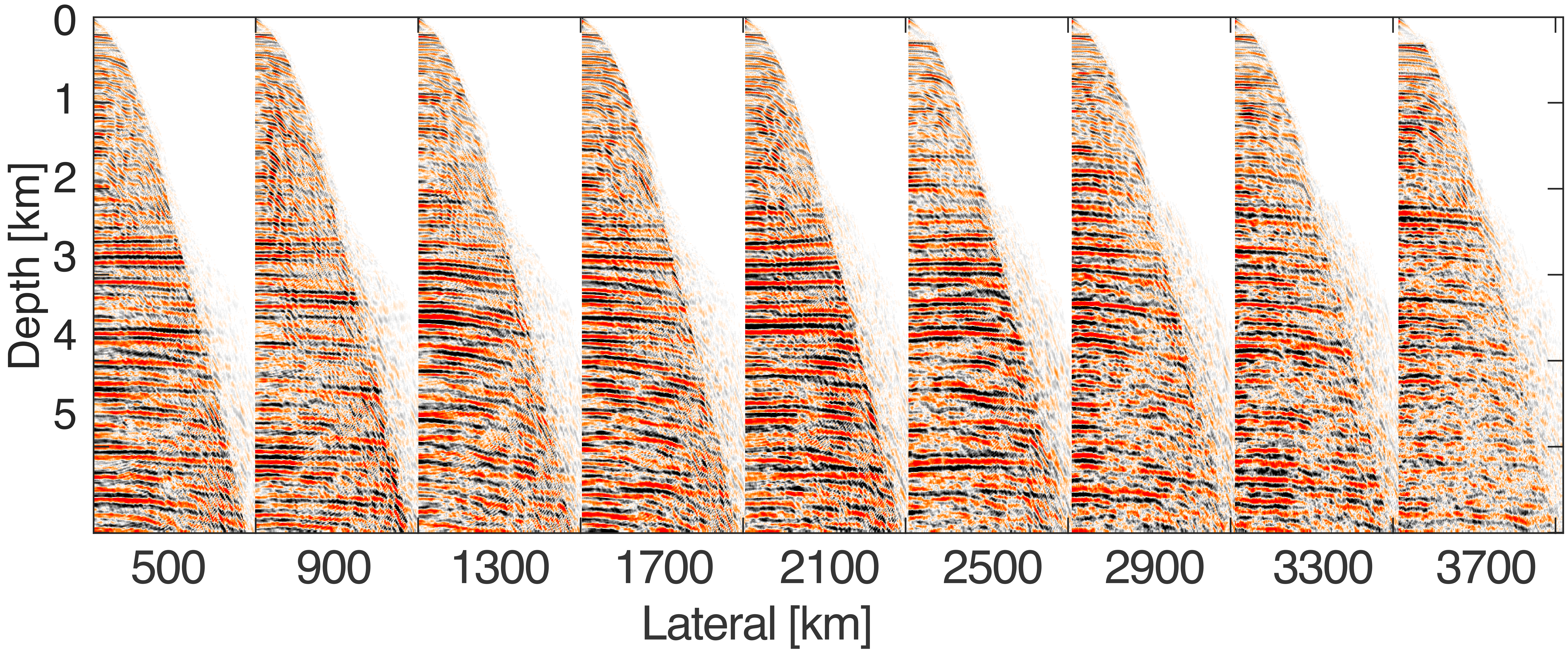
Kirchhoff migration
—Initial model



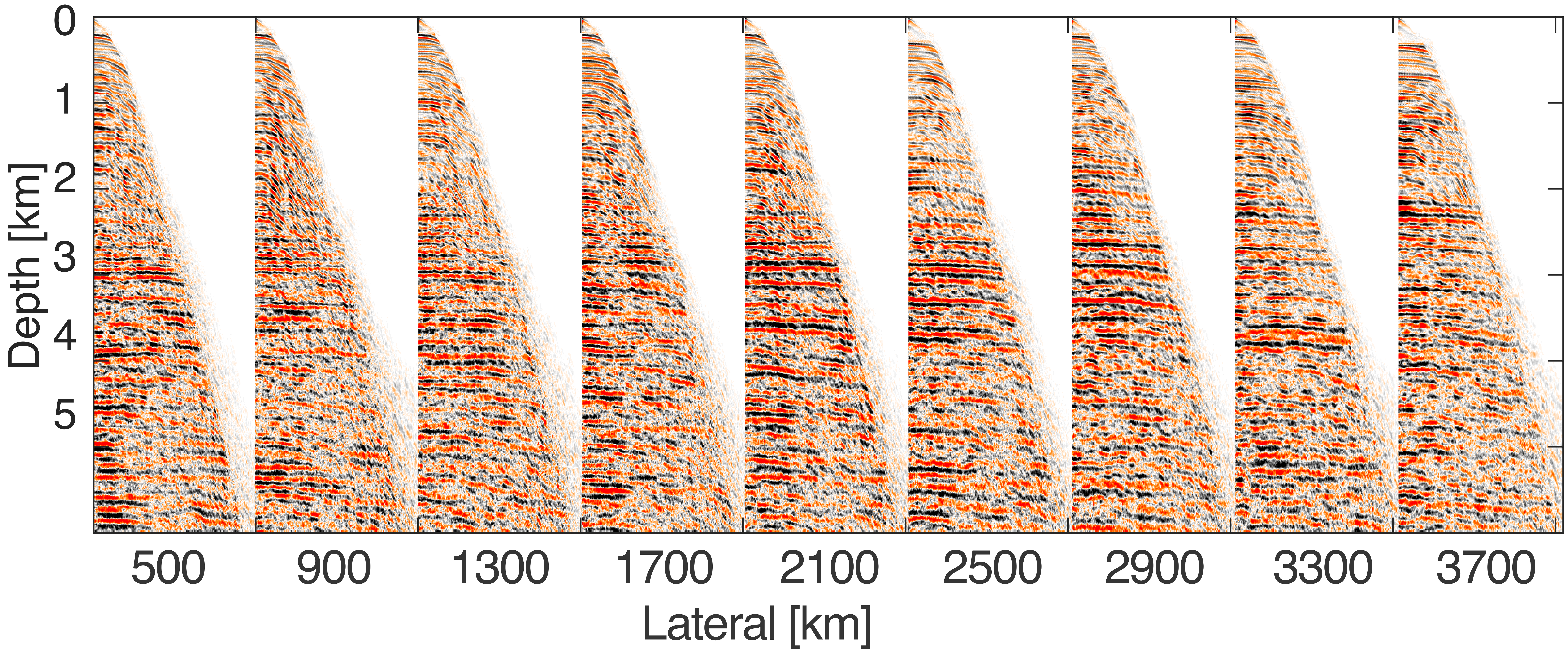
Kirchhoff migration
—Inversion result



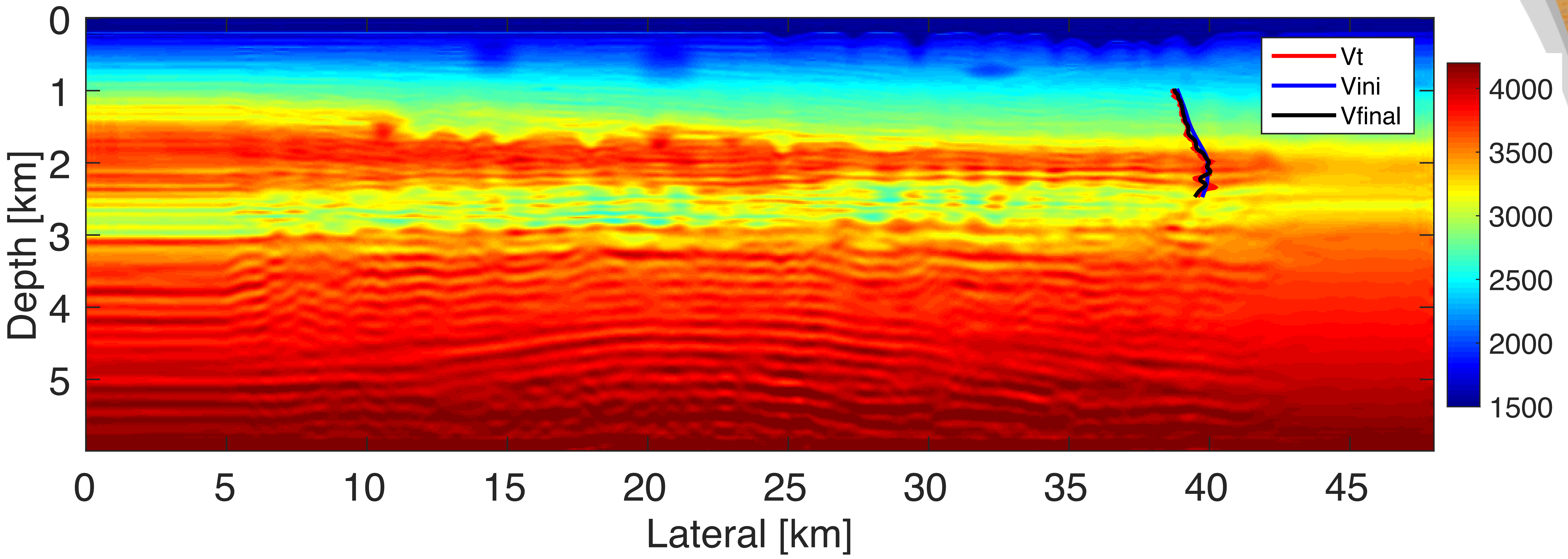
Common Image Gather
—Initial model



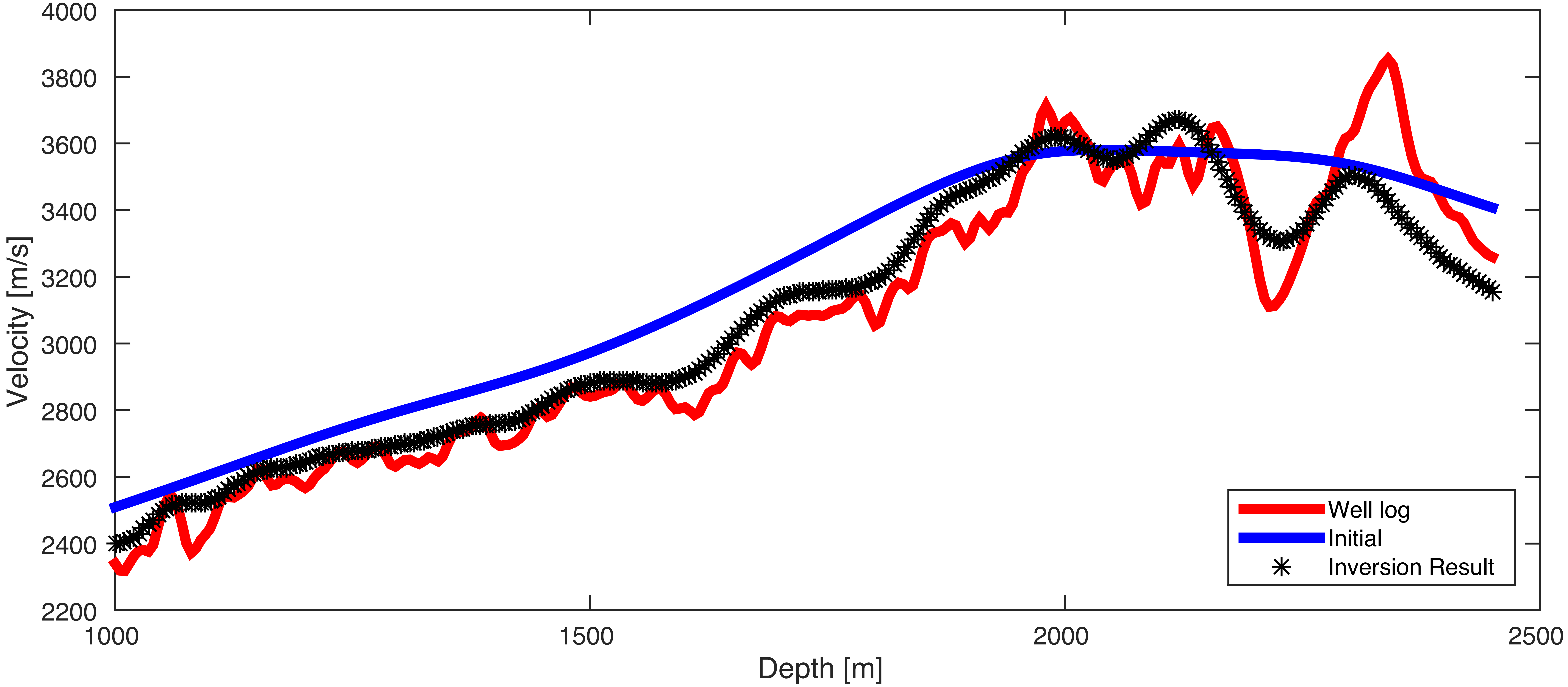
Common Image Gather
—Inversion result



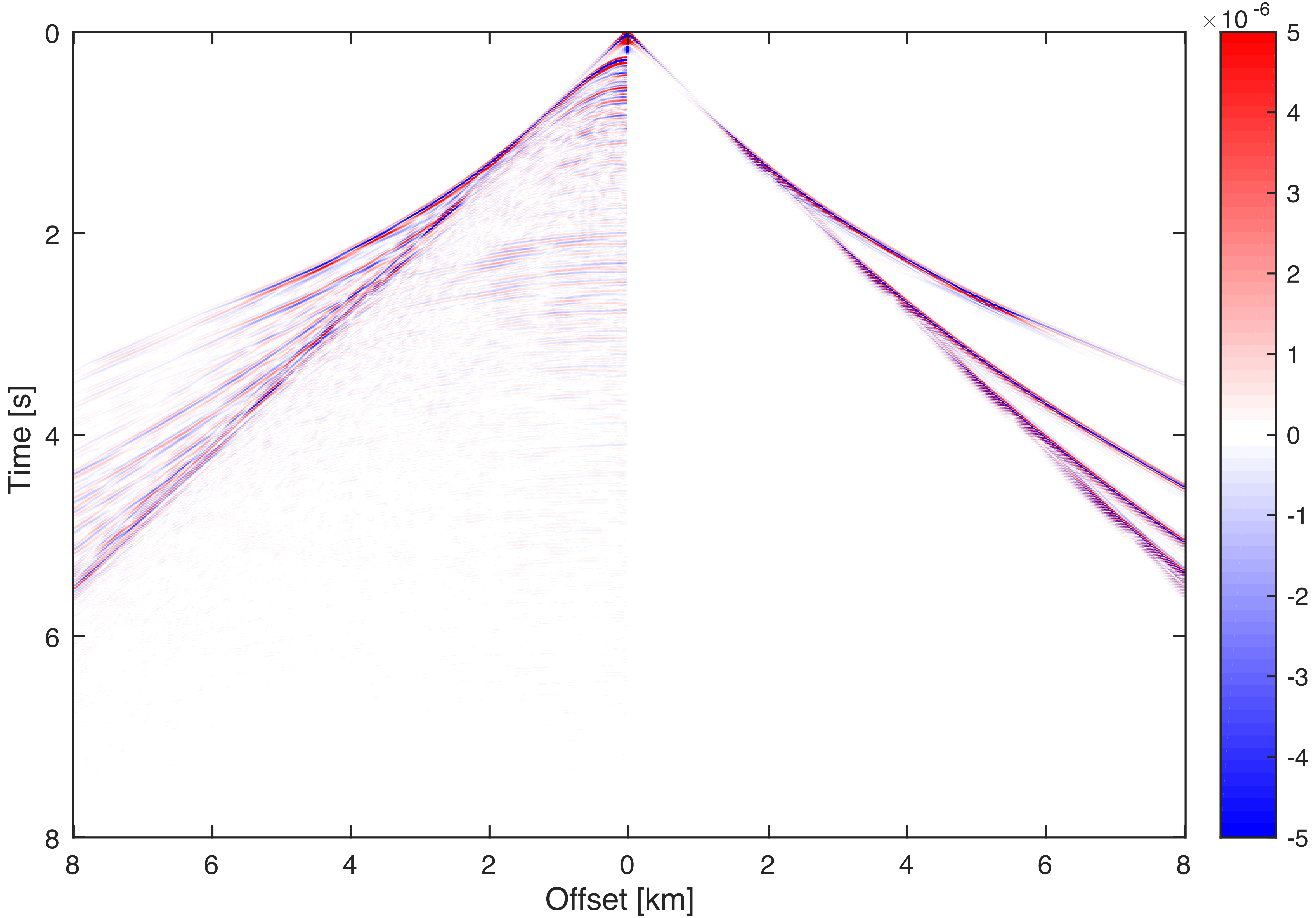
Well-log comparison



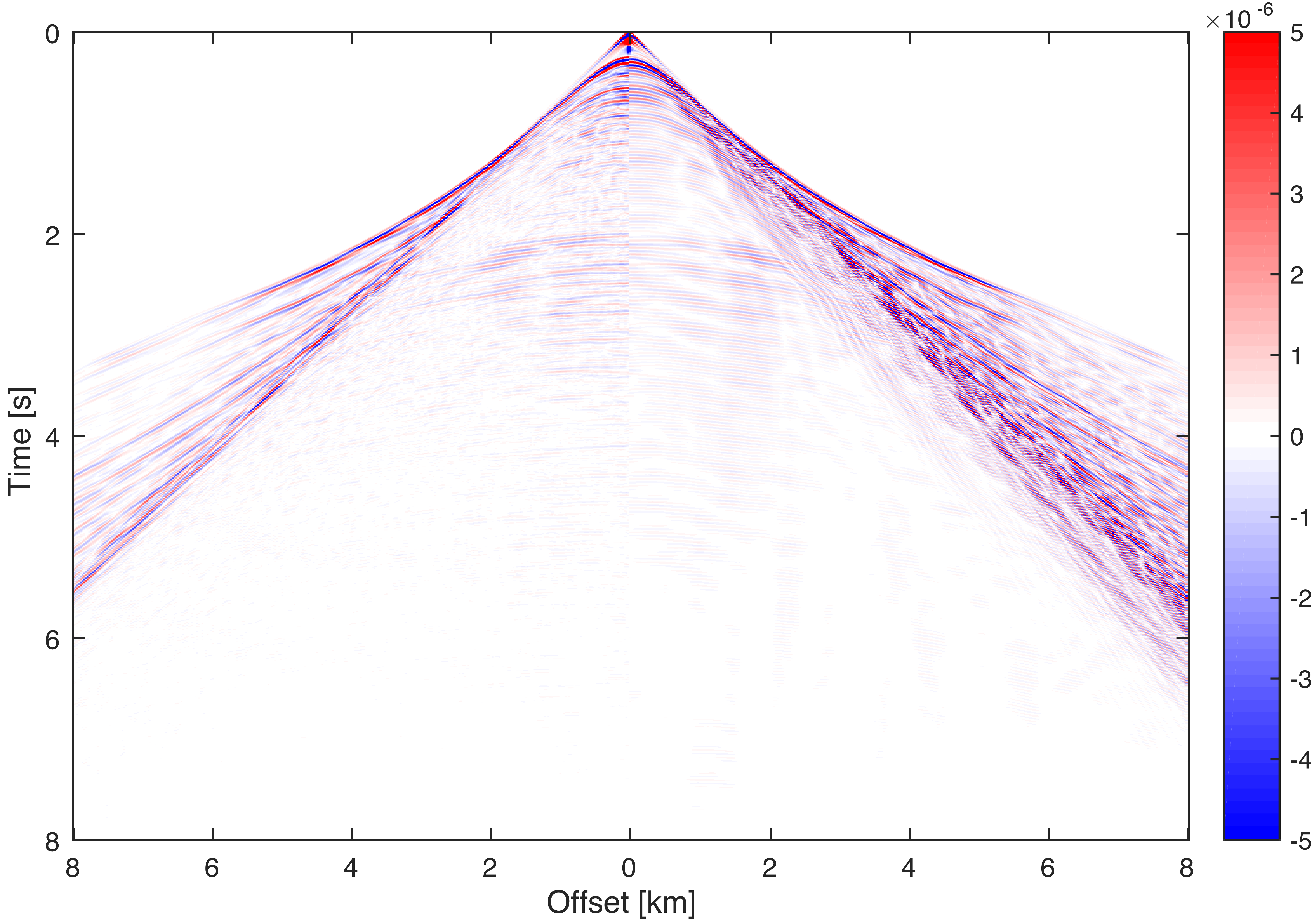
Well-log comparison



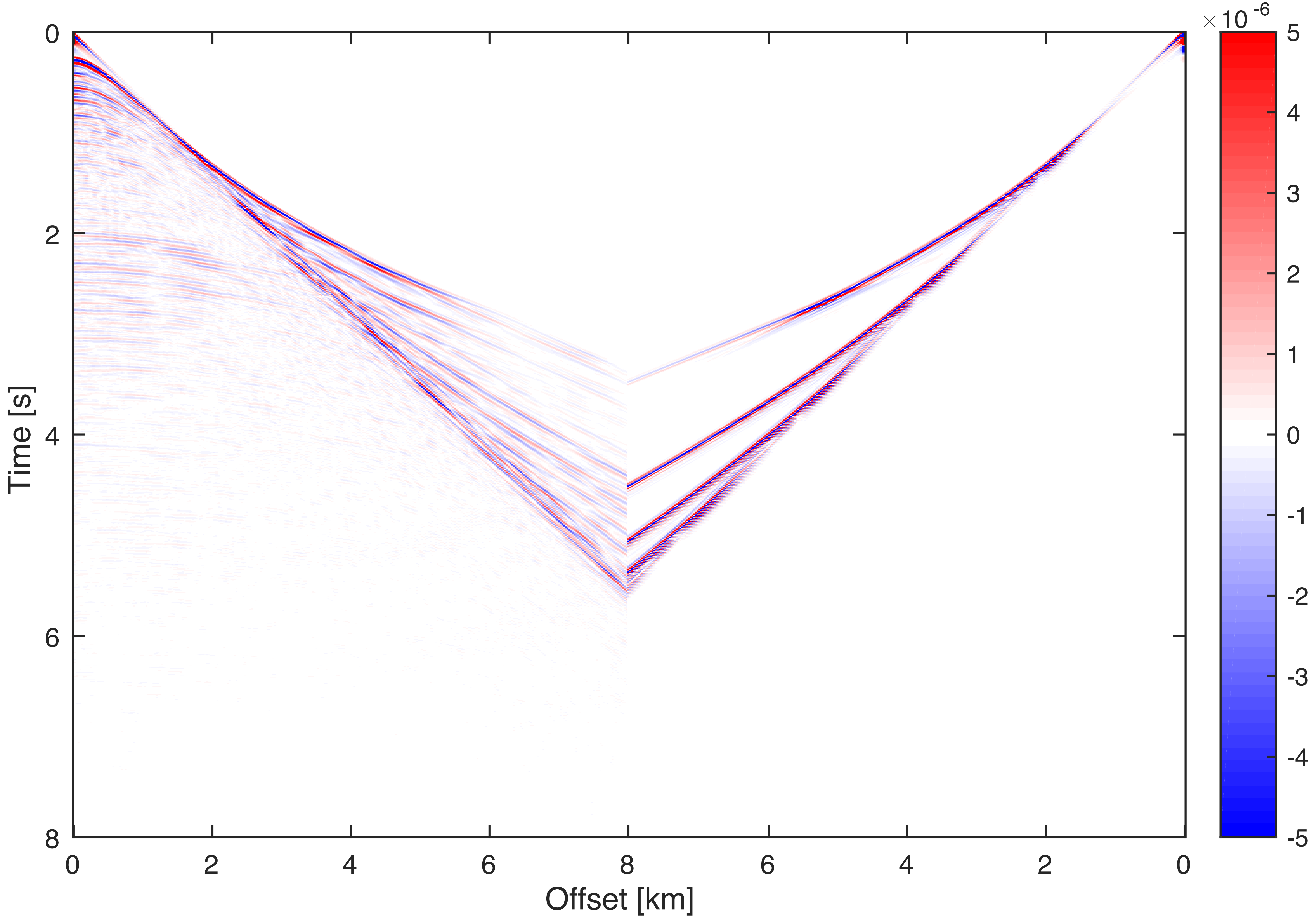
Shot record comparison— Initial model



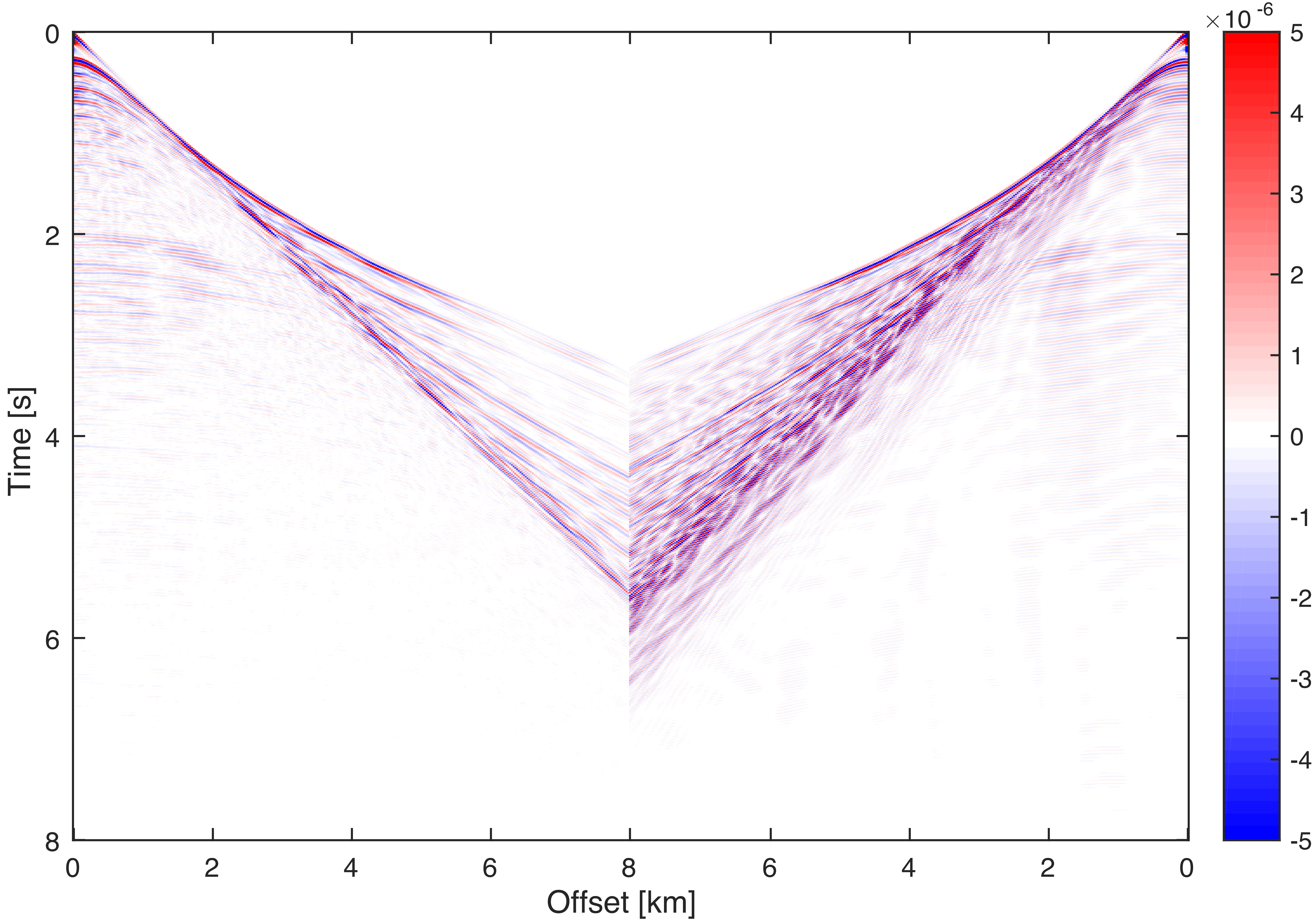
Shot record comparison—— Inversion result



Shot record comparison—— Initial model



Shot record comparison—— Inversion result



Conclusions

1. Using the variable projection method, we can estimate the source wavelet for the WRI.
    - Synthetic BG model

2. Source estimation enhances the robustness of WRI for field seismic data.
    - Chevron blind test data

Acknowledgements

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