

# Nonlinear FWI: formulation and examples

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Dec 2<sup>nd</sup>, 2016

Banff, AB

# Outline

- Introduction
  - General principle of FWI
  - Nonlinearity of FWI
- Theory
  - Sensitivities: from linear to nonlinear
  - Determining model perturbation before n+1th iteration
  - Two-iteration nonlinear FWI
- Examples
- Conclusions
- Acknowledgements

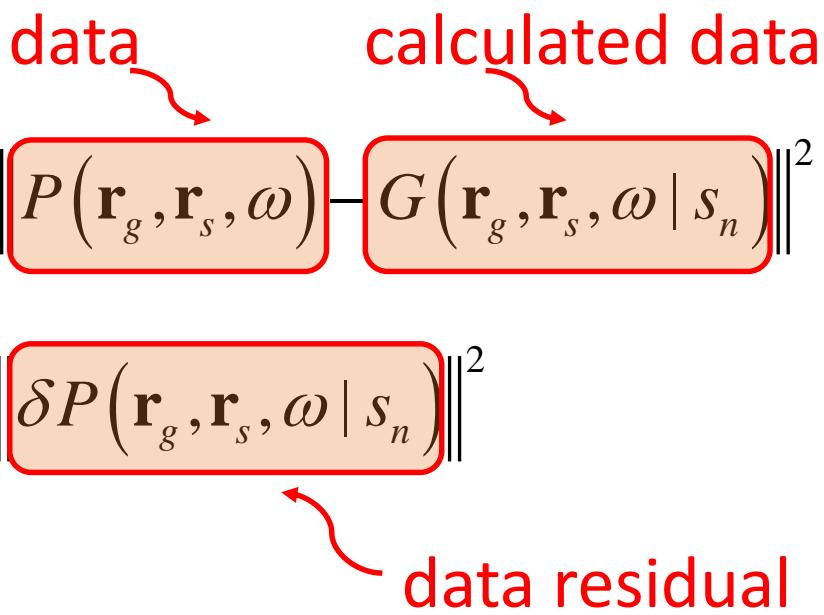
# General principle of FWI

- Forward modeling

$$[\omega^2 s(\mathbf{r}) + \nabla^2] P(\mathbf{r}, \mathbf{r}_s, \omega) = -\delta(\mathbf{r} - \mathbf{r}_s)$$

- Misfit function

$$\begin{aligned}\phi(s) &= \frac{1}{2} \sum_{\mathbf{r}_s} \sum_{\mathbf{r}_g} \sum_{\omega} \left| P(\mathbf{r}_g, \mathbf{r}_s, \omega) - G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n) \right|^2 \\ &= \frac{1}{2} \sum_{\mathbf{r}_s} \sum_{\mathbf{r}_g} \sum_{\omega} \left| \delta P(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n) \right|^2\end{aligned}$$

data                      calculated data  


# General principle of FWI

- Gradient of misfit

$$g_n(\mathbf{r}) = -\sum_{\mathbf{r}_s} \sum_{\mathbf{r}_g} \sum_{\omega} \operatorname{Re} \left( \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)}{\partial s(\mathbf{r})} \delta P^*(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n) \right)$$

sensitivity

- Update model iteratively

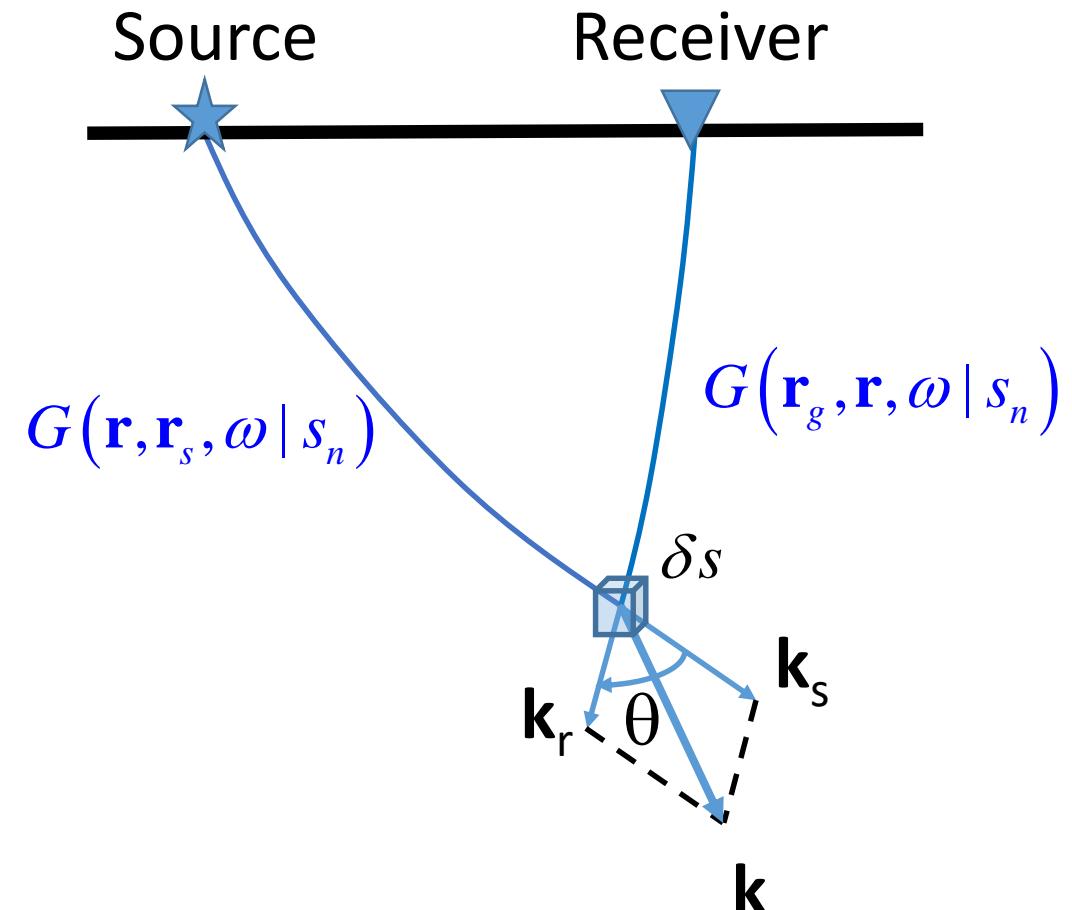
$$s_{n+1}(\mathbf{r}) = s_n(\mathbf{r}) - \mu_n g_n(\mathbf{r})$$

# Nonlinearity of FWI

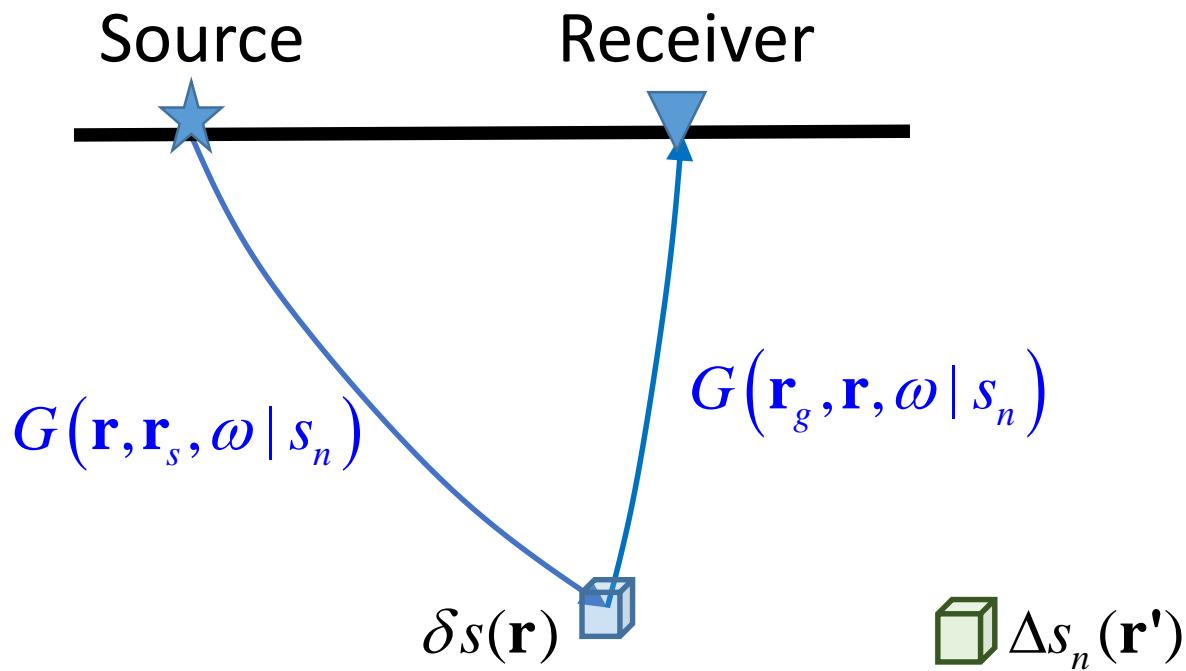
- FWI relies on the diffraction tomography principle

$$\mathbf{k} = \frac{2\omega}{v} \cos\left(\frac{\theta}{2}\right) \mathbf{n}$$

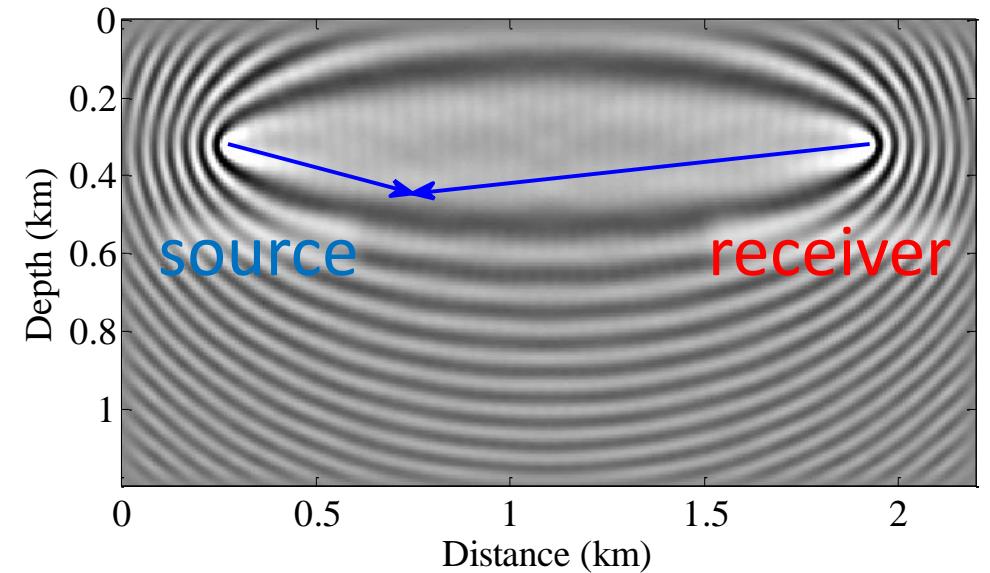
- One frequency and one aperture map one wavenumber in the model space
- Low frequencies and wide apertures data help to recover long-to-intermediate wavelengths structures
- Nonlinearity in FWI
  - Multiscale FWI
  - Build good starting model
  - Nonlinear sensitivities



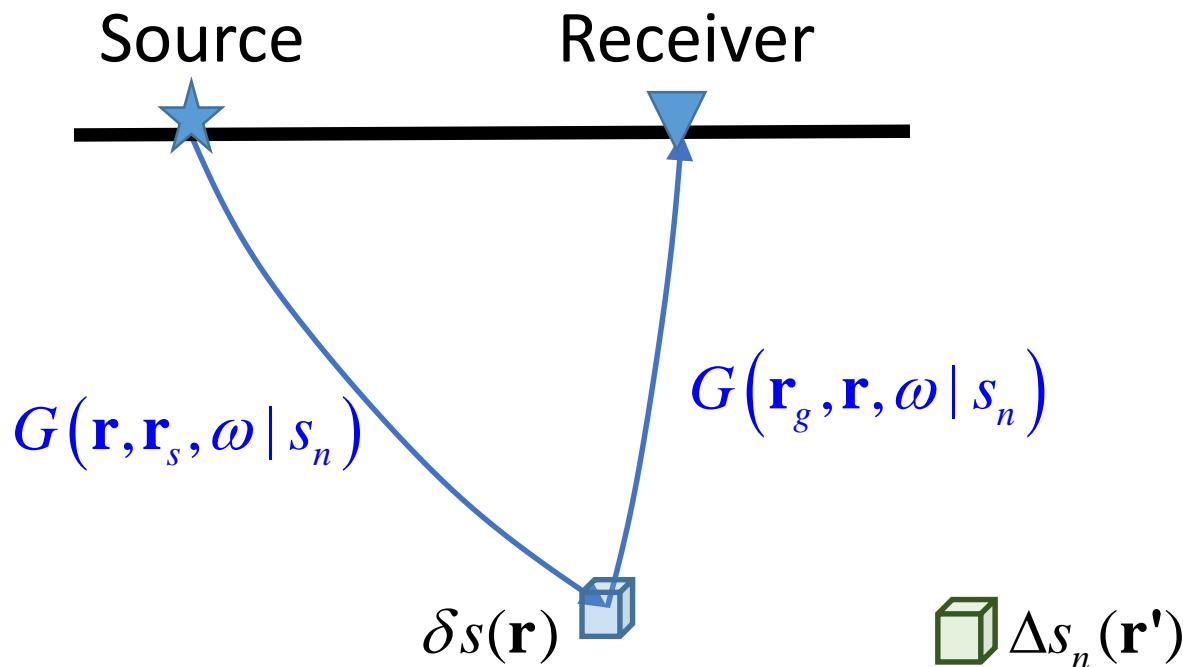
# Sensitivities: from linear to nonlinear



$$\frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)}{\partial s(\mathbf{r})} = \omega^2 G(\mathbf{r}_g, \mathbf{r}, \omega | s_n) G(\mathbf{r}, \mathbf{r}_s, \omega | s_n)$$



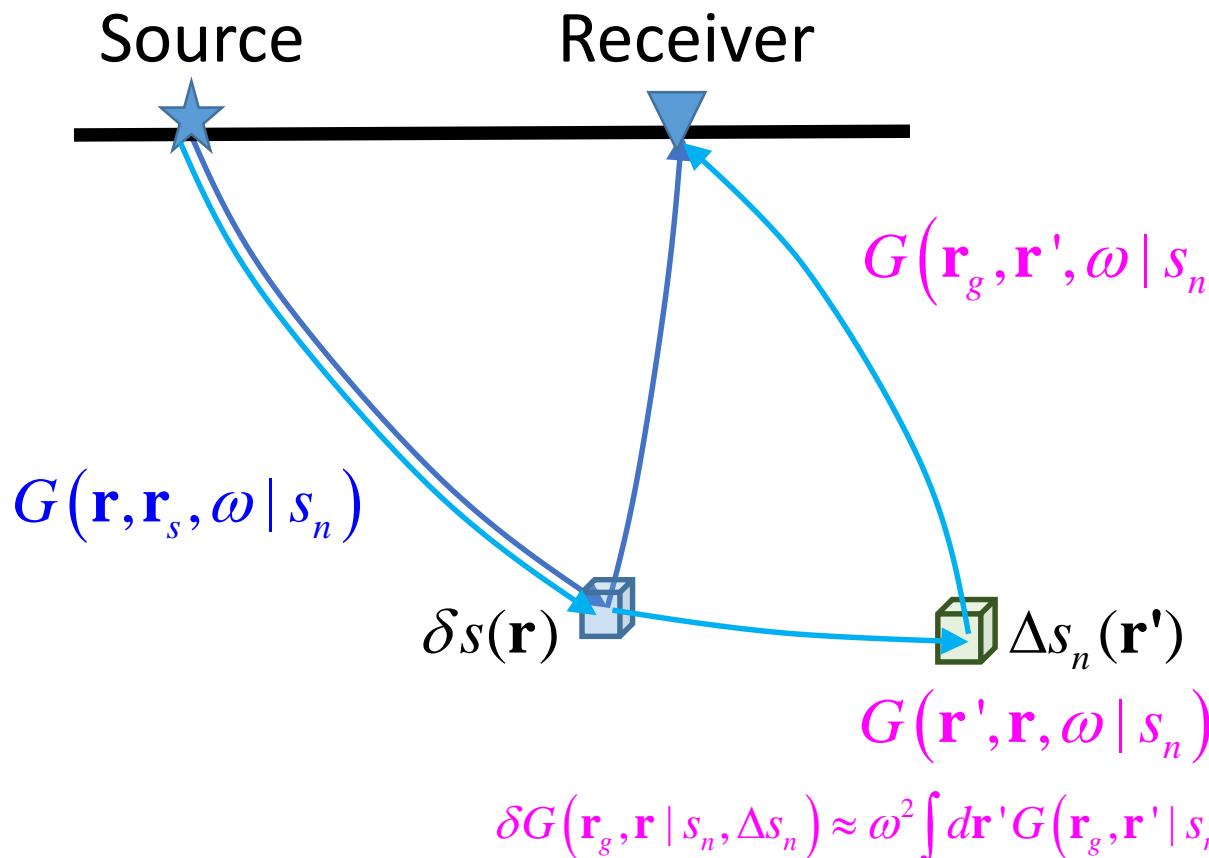
# Sensitivities: from linear to nonlinear



## Zero-order sensitivity

$$\left( \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_{n+1})}{\partial s(\mathbf{r})} \right)_0 = \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)}{\partial s(\mathbf{r})}$$
$$= \omega^2 G(\mathbf{r}_g, \mathbf{r}, \omega | s_n) G(\mathbf{r}, \mathbf{r}_s, \omega | s_n)$$

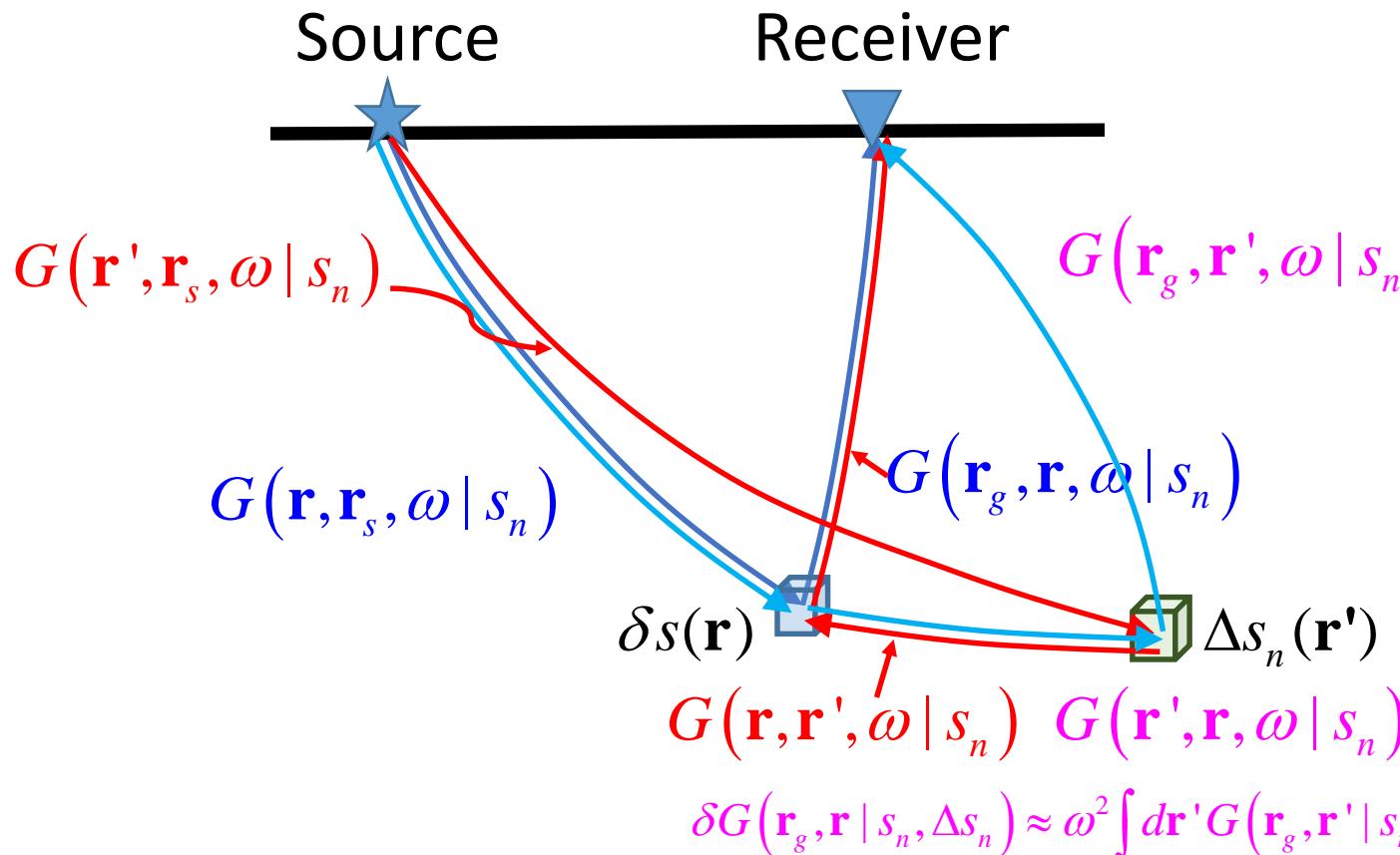
# Sensitivities: from linear to nonlinear



Zero-order sensitivity

$$\left( \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_{n+1})}{\partial s(\mathbf{r})} \right)_0 = \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)}{\partial s(\mathbf{r})}$$
$$= \omega^2 G(\mathbf{r}_g, \mathbf{r}, \omega | s_n) G(\mathbf{r}, \mathbf{r}_s, \omega | s_n)$$

# Sensitivities: from linear to nonlinear

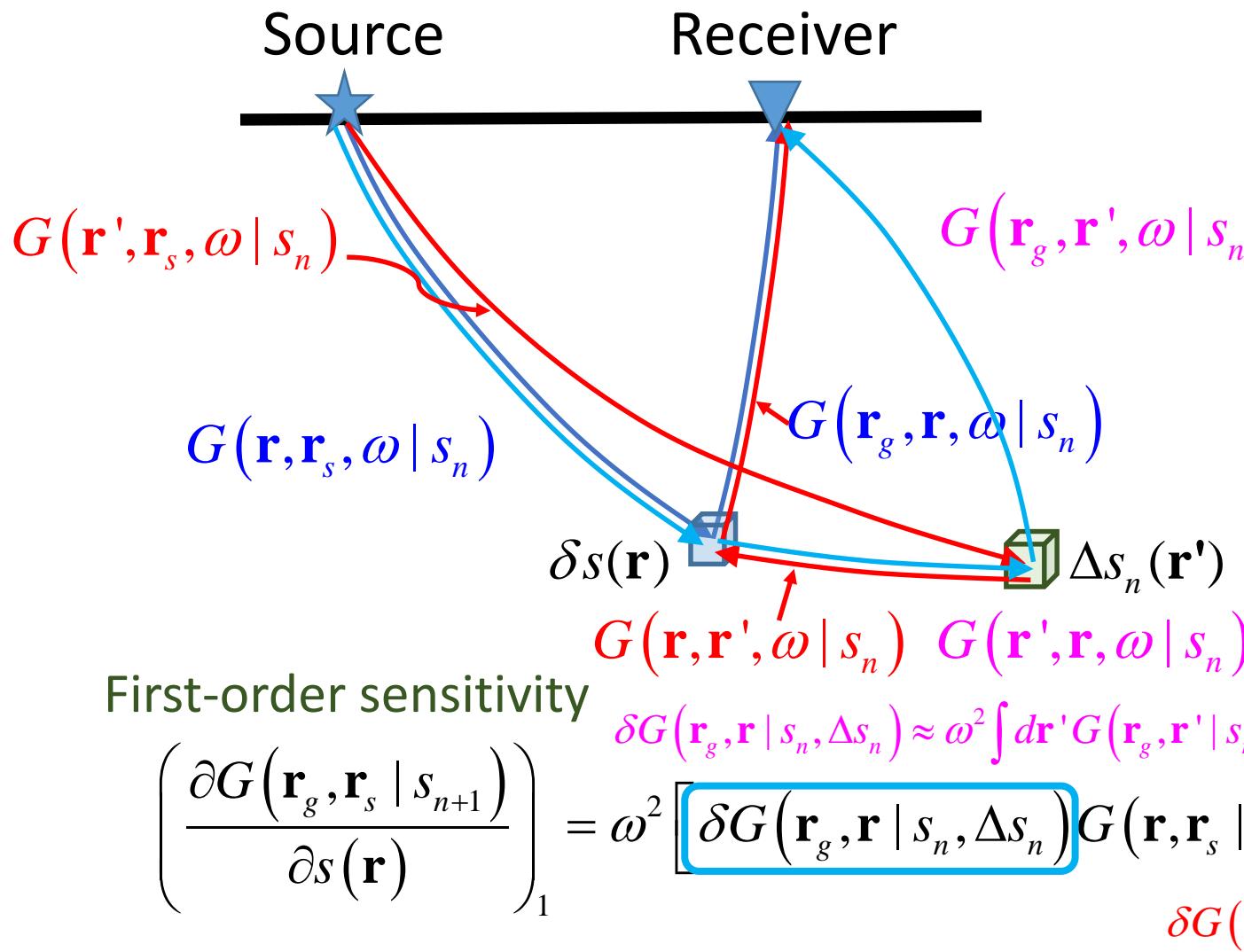


## Zero-order sensitivity

$$\left( \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_{n+1})}{\partial s(\mathbf{r})} \right)_0 = \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)}{\partial s(\mathbf{r})}$$
$$= \omega^2 G(\mathbf{r}_g, \mathbf{r}, \omega | s_n) G(\mathbf{r}, \mathbf{r}_s, \omega | s_n)$$

$$\delta G(\mathbf{r}, \mathbf{r}_s | s_n, \Delta s_n) \approx \omega^2 \int d\mathbf{r}' G(\mathbf{r}, \mathbf{r}' | s_n) G(\mathbf{r}', \mathbf{r}_s | s_n) \Delta s_n(\mathbf{r}')$$

# Sensitivities: from linear to nonlinear

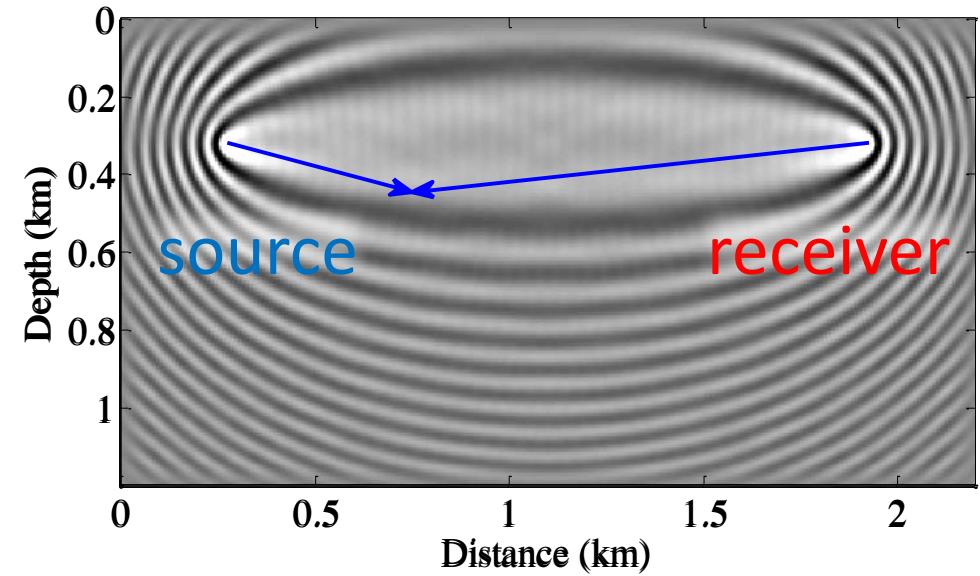
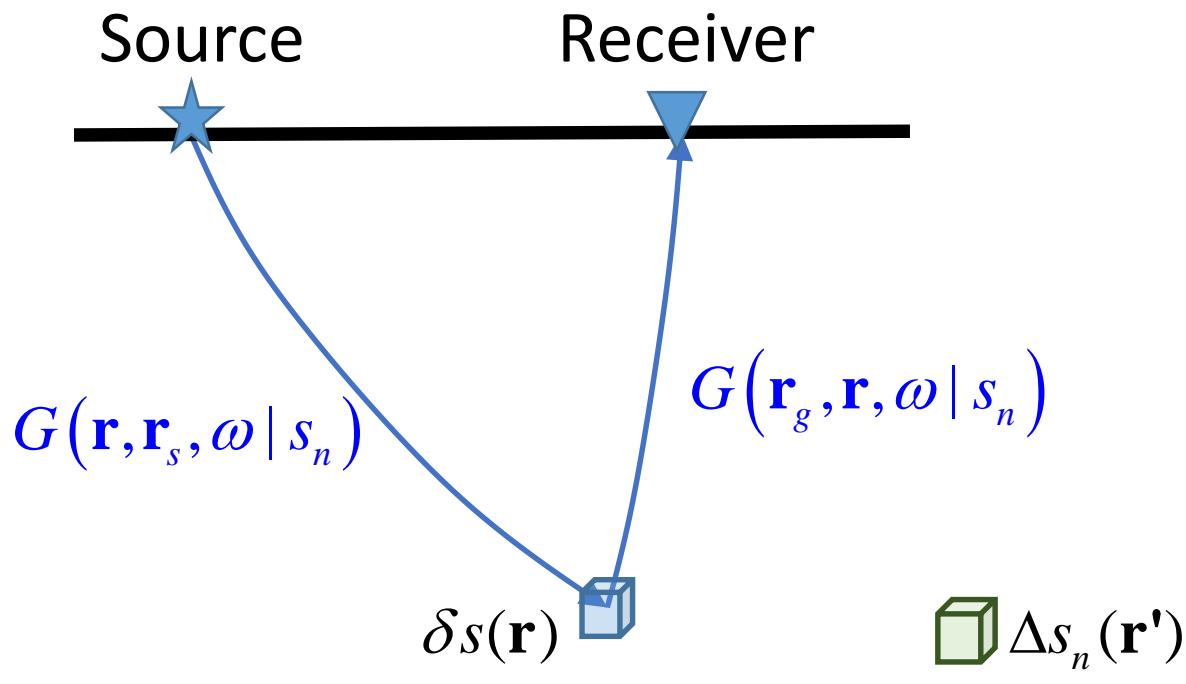


## Zero-order sensitivity

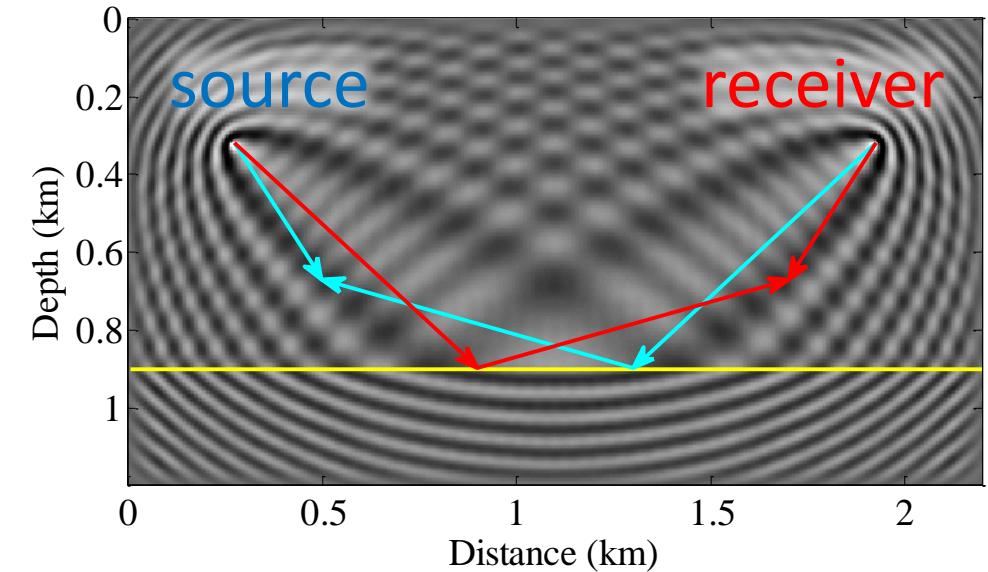
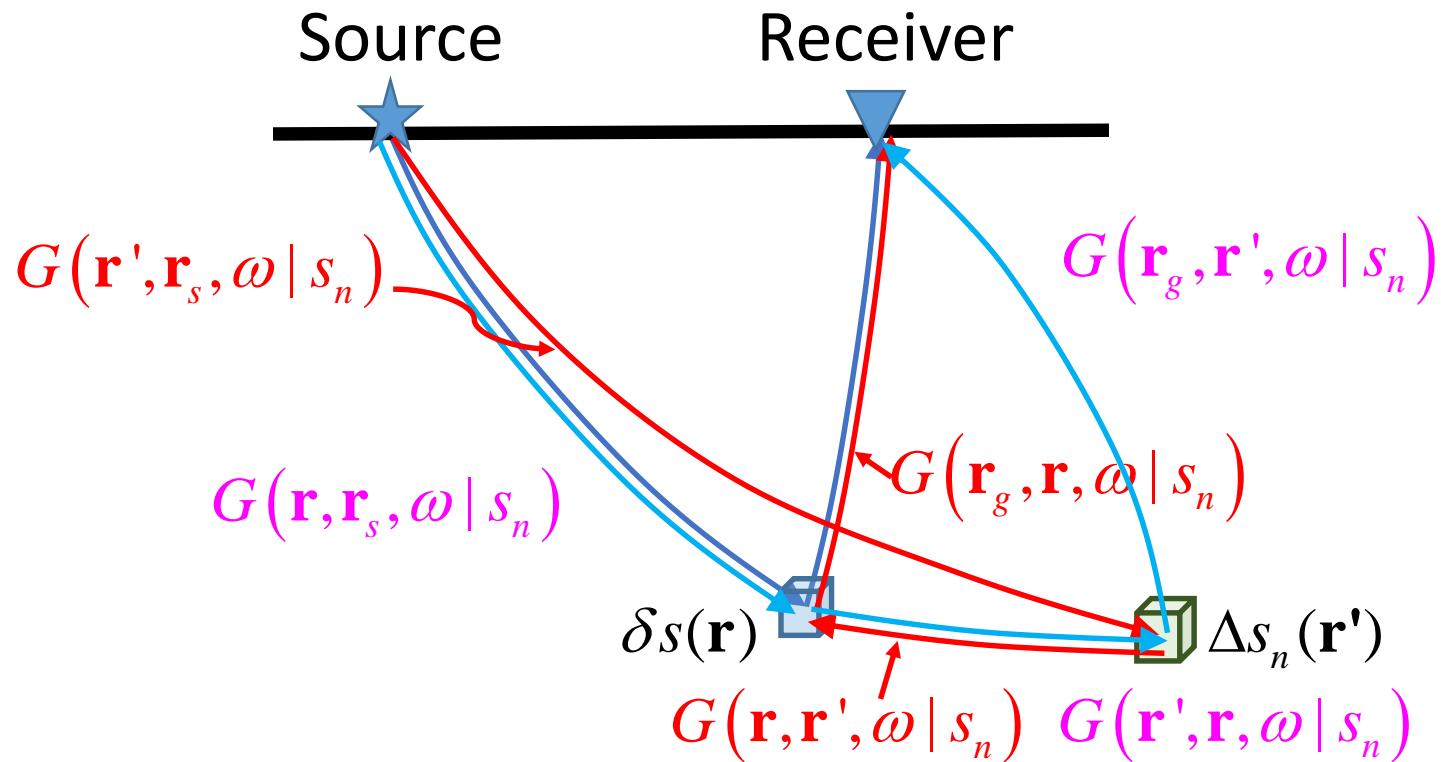
$$\left( \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_{n+1})}{\partial s(\mathbf{r})} \right)_0 = \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)}{\partial s(\mathbf{r})}$$

$$= \omega^2 G(\mathbf{r}_g, \mathbf{r}, \omega | s_n) G(\mathbf{r}, \mathbf{r}_s, \omega | s_n)$$

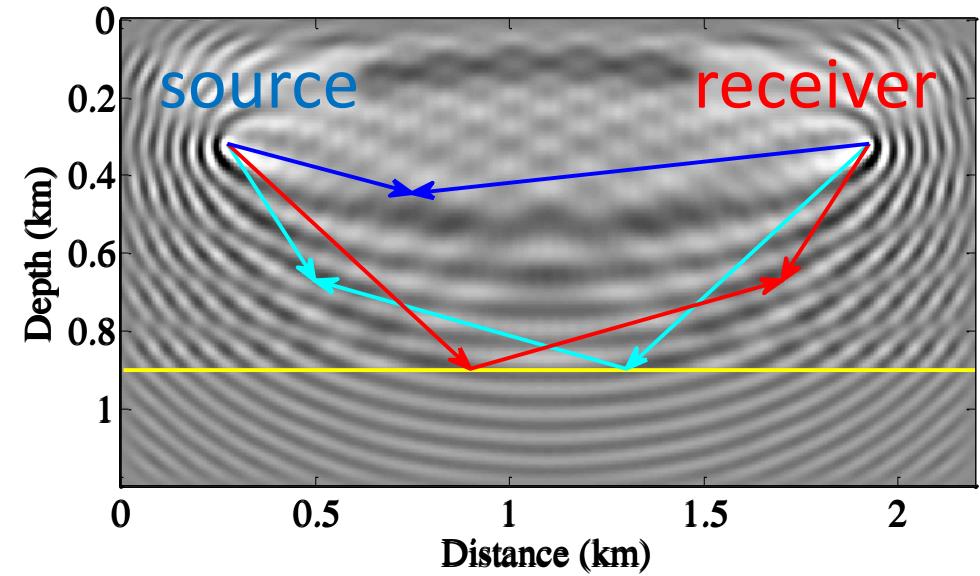
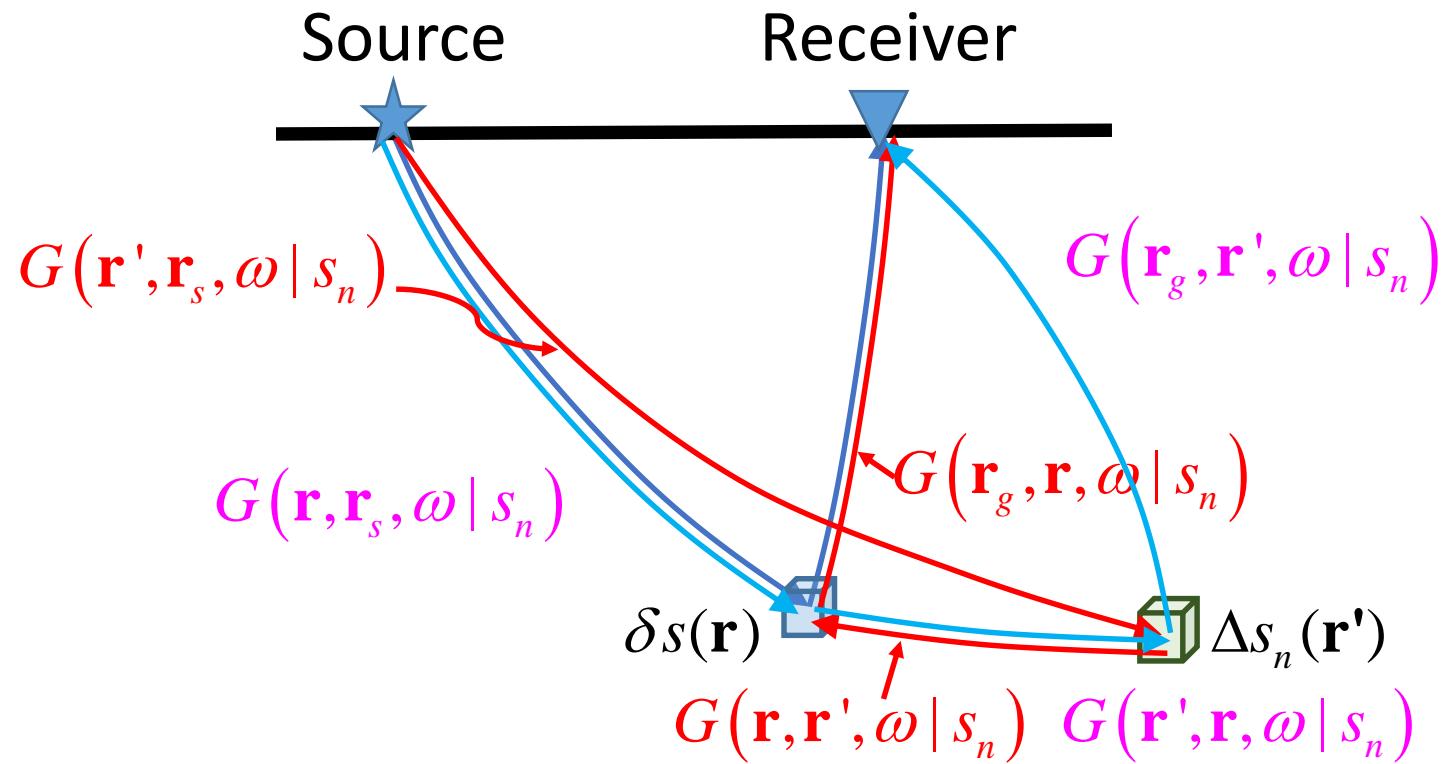
# Sensitivities: from linear to nonlinear



# Sensitivities: from linear to nonlinear



# Sensitivities: from linear to nonlinear



$$\frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_{n+1})}{\partial s(\mathbf{r})} = \left( \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_{n+1})}{\partial s(\mathbf{r})} \right)_0 + \left( \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | s_{n+1})}{\partial s(\mathbf{r})} \right)_1 + \dots$$

# Determining model perturbation before n+1th iteration

- Determining perturbation before n+1th FWI iteration

Linear relationship between  $\delta P$  and  $\Delta s$

$$\delta P(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n) = \omega^2 \int d\mathbf{r}' G(\mathbf{r}_g, \mathbf{r}', \omega | s_n) \Delta s(\mathbf{r}') G(\mathbf{r}', \mathbf{r}_s, \omega | s_n) + \dots$$

Difference between true model and the current updated model  $s_n$

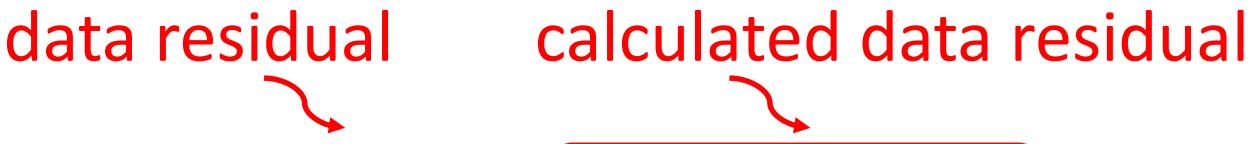
- Direct nonlinear inverse scattering
- Approach by the solution of a linear inverse problem

# Determining model perturbation before n+1th iteration

- Misfit function

$$\phi(\delta\tilde{s}) = \frac{1}{2} \sum_{\mathbf{r}_s} \sum_{\mathbf{r}_g} \sum_{\omega} \left\| \delta P(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n) - \delta P_{cal}(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n) \right\|^2$$

data residual      calculated data residual



- Update  $\delta\tilde{s}$  iteratively

$$\tilde{g}_m(\mathbf{r}) = - \sum_{\mathbf{r}_s} \sum_{\mathbf{r}_g} \sum_{\omega} \operatorname{Re} \left( \omega^2 G(\mathbf{r}_g, \mathbf{r}, \omega | s_n) G(\mathbf{r}, \mathbf{r}_s, \omega | s_n) \delta S_m^*(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n) \right)$$

$$\delta\tilde{s}_{m+1}(\mathbf{r}) = \delta\tilde{s}_m(\mathbf{r}) - \alpha_m \tilde{g}_m(\mathbf{r})$$

# Nonlinear FWI

- Direct update of the model (**DWI**)

$$\tilde{s}_{n+1}(\mathbf{r}) = s_n(\mathbf{r}) + \delta\tilde{s}(\mathbf{r})$$

- Nonlinear FWI with gradient

$$g_n(\mathbf{r}) = - \sum_{\mathbf{r}_s} \sum_{\mathbf{r}_g} \sum_{\omega} \operatorname{Re} \left( \frac{\partial G(\mathbf{r}_g, \mathbf{r}_s, \omega | \tilde{s}_{n+1})}{\partial s(\mathbf{r})} \delta P^*(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n) \right)$$

• **FOFWI:**  $= \omega^2 [G(\mathbf{r}, \mathbf{r}_s, \omega | s_n) G(\mathbf{r}_g, \mathbf{r}, \omega | s_n) + \delta G(\mathbf{r}_g, \mathbf{r}, \omega | s_n, \delta\tilde{s}) G(\mathbf{r}, \mathbf{r}_s, \omega | s_n) + G(\mathbf{r}_g, \mathbf{r}, \omega | s_n) \delta G(\mathbf{r}, \mathbf{r}_s, \omega | s_n, \delta\tilde{s})]$

• **NFWI:**  $= \omega^2 [G(\mathbf{r}, \mathbf{r}_s, \omega | \tilde{s}_{n+1}) G(\mathbf{r}_g, \mathbf{r}, \omega | \tilde{s}_{n+1}) - \delta G(\mathbf{r}_g, \mathbf{r}, \omega | s_n, \delta\tilde{s}) \delta G(\mathbf{r}, \mathbf{r}_s, \omega | s_n, \delta\tilde{s})]$

# Two-iteration nonlinear FWI

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**Algorithm 2** Algorithm for frequency domain nonlinear FWI

---

**Input:** Recorded seismic data  $P(\mathbf{r}_g, \mathbf{r}_s, \omega)$ , initial model  $s_0$

**Output:** Inverted model  $s_n$

```
1: for  $\omega = \omega_{min}, \dots, \omega_{max}$  do                                ▷ Frequency loop
2:   for  $n = 0, \dots, n_{max}$  do                                ▷ Outer loop for FWI
3:     Get data residual  $\delta P(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)$ 
4:
5:
6:
7:   No inner iteration=> FWI
8:
9:
10:
11:
12:   Calculate nonlinear gradient (equation (33) or (39))
13:   Calculate the step length  $\mu_n$  using the line search method
14:   Update to  $s_{n+1}$  (equation (8))
15: end for
16: end for
```

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# Two-iteration nonlinear FWI

---

**Algorithm 2** Algorithm for frequency domain nonlinear FWI

---

**Input:** Recorded seismic data  $P(\mathbf{r}_g, \mathbf{r}_s, \omega)$ , initial model  $s_0$

**Output:** Inverted model  $s_n$

```
1: for  $\omega = \omega_{min}, \dots, \omega_{max}$  do                                ▷ Frequency loop
2:   for  $n = 0, \dots, n_{max}$  do                                     ▷ Outer loop for FWI
3:     Get data residual  $\delta P(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)$ 
4:     Initial perturbation  $\tilde{s}_0 \leftarrow 0$ 
5:     for  $m = 0, \dots, m_{max}$  do           ▷ Inner loop for perturbation
6:       Calculate  $\delta P_{cal}(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)$ 
7:       Get data residual  $\delta S(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)$ 
8:       Calculate the gradient  $\tilde{g}_m(\mathbf{r})$  (equation (24))
9:       Calculate the step length  $\alpha_m$  using the line search method
10:      Update  $\tilde{s}_m$  (equation (23))
11:    end for
12:    Calculate nonlinear gradient (equation (33) or (39))
13:    Calculate the step length  $\mu_n$  using the line search method
14:    Update to  $s_{n+1}$  (equation (8))
15:  end for
16: end for
```

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# Two-iteration nonlinear FWI

---

**Algorithm 2** Algorithm for frequency domain nonlinear FWI

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**Input:** Recorded seismic data  $P(\mathbf{r}_g, \mathbf{r}_s, \omega)$ , initial model  $s_0$

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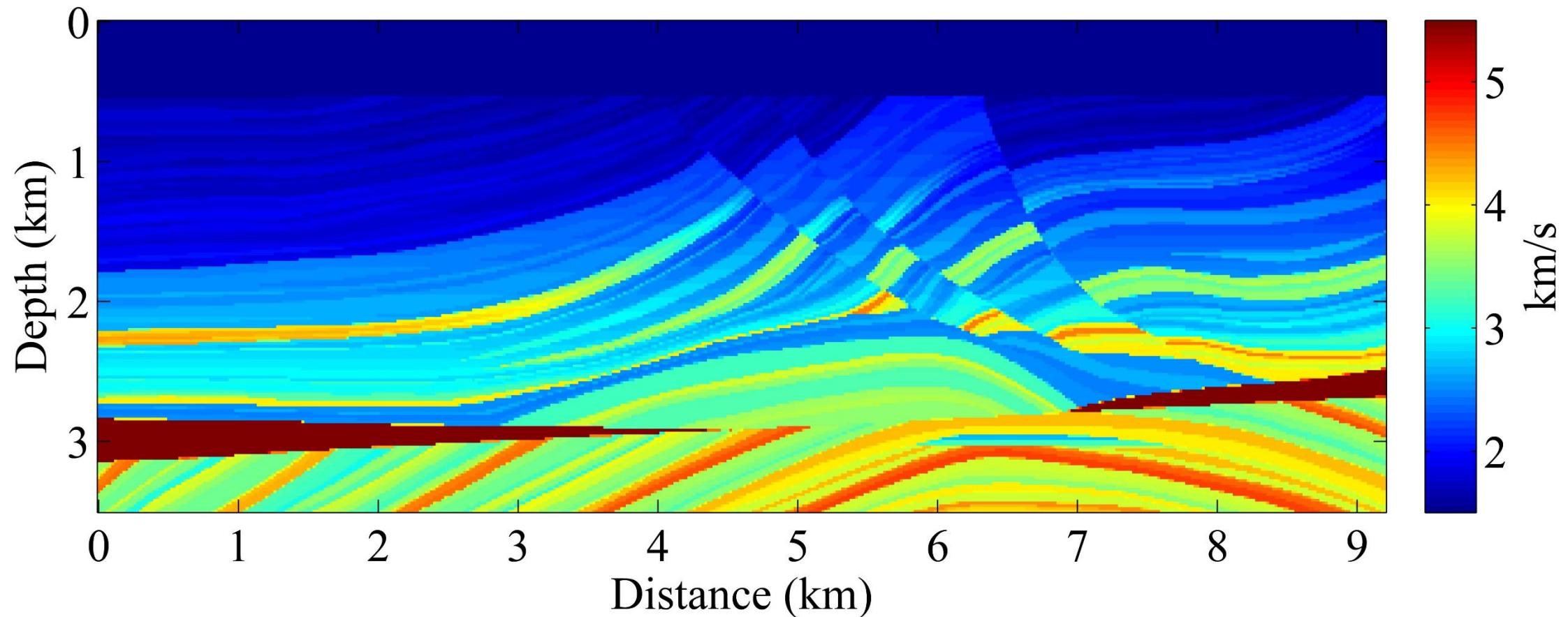
```
1: for  $\omega = \omega_{min}, \dots, \omega_{max}$  do                                ▷ Frequency loop
2:   for  $n = 0, \dots, n_{max}$  do                                     ▷ Outer loop for FWI
3:     Get data residual  $\delta P(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)$ 
4:     Initial perturbation  $\tilde{s}_0 \leftarrow 0$ 
5:     for  $m = 0, \dots, m_{max}$  do           ▷ Inner loop for perturbation
6:       Calculate  $\delta P_{cal}(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)$ 
7:       Get data residual  $\delta S(\mathbf{r}_g, \mathbf{r}_s, \omega | s_n)$ 
8:       Calculate the gradient  $\tilde{g}_m(\mathbf{r})$  (equation (24))
9:       Calculate the step length  $\alpha_m$  using the line search method
10:      Update  $\tilde{s}_m$  (equation (23))
11:    end for
12:
13:    Direct update
14:    Update to  $s_{n+1}$  (equation (8))
15:  end for
16: end for
```

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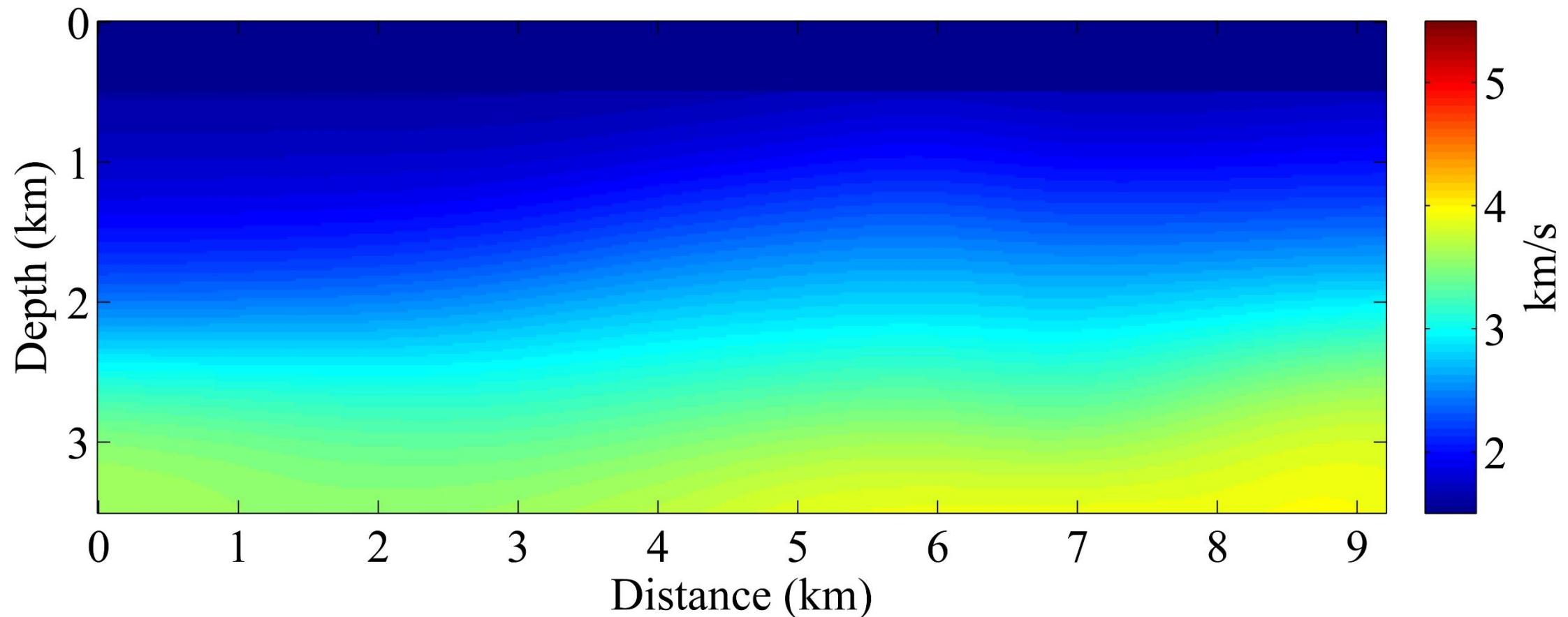
# Numerical examples

- 2D Marmousi model with added water layer
  - Model size  $3.5\text{km} \times 9.2\text{Km}$  with  $20\text{m}$  interval
  - 46 sources with  $200\text{m}$  spacing, starting from  $100\text{m}$
  - 461 receivers with  $20\text{m}$  spacing, starting from  $0\text{m}$
- Test with different initial model
  - True model smoothed with Gaussian window
  - Vertically linearly increasing model ( $1500\text{-}4000\text{m/s}$ )
- Test with different frequency range
  - Starting from  $2\text{Hz}$  to  $15\text{Hz}$  ( $2\text{Hz}, 3.3\text{Hz}, 5.5\text{Hz}, 9\text{Hz}, 14.9\text{Hz}$ )
  - Starting from  $4\text{Hz}$  to  $15\text{Hz}$ 
    - $4\text{Hz}, 6.6\text{Hz}, 14.9\text{Hz}$
    - 12 frequencies with  $1\text{Hz}$  interval

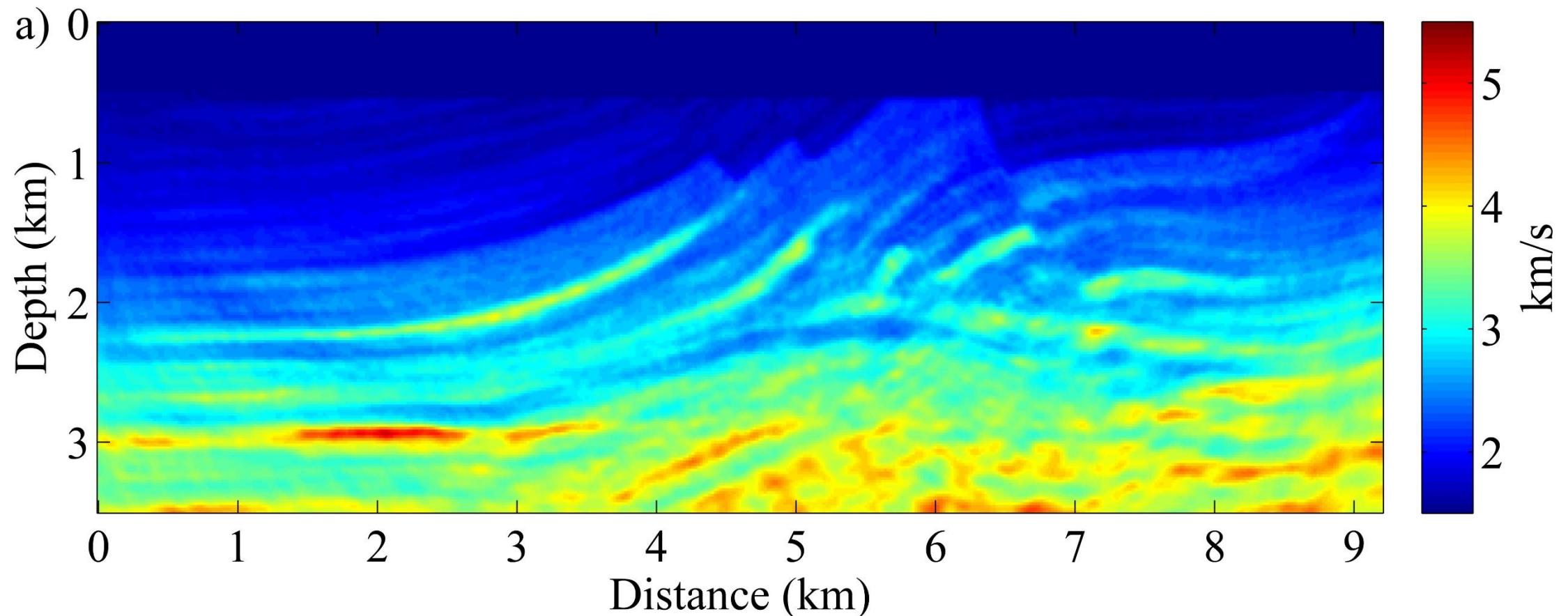
# True Marmousi model



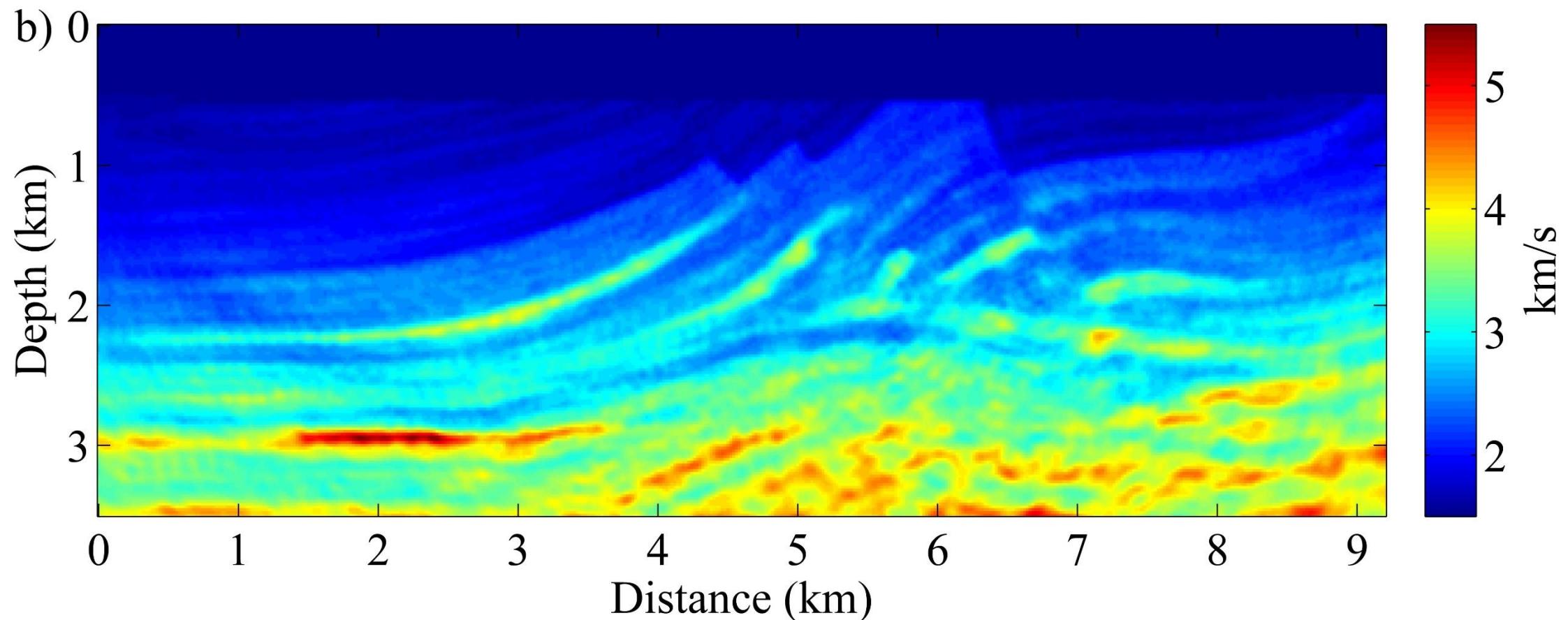
# Initial model obtained after smooth the true model



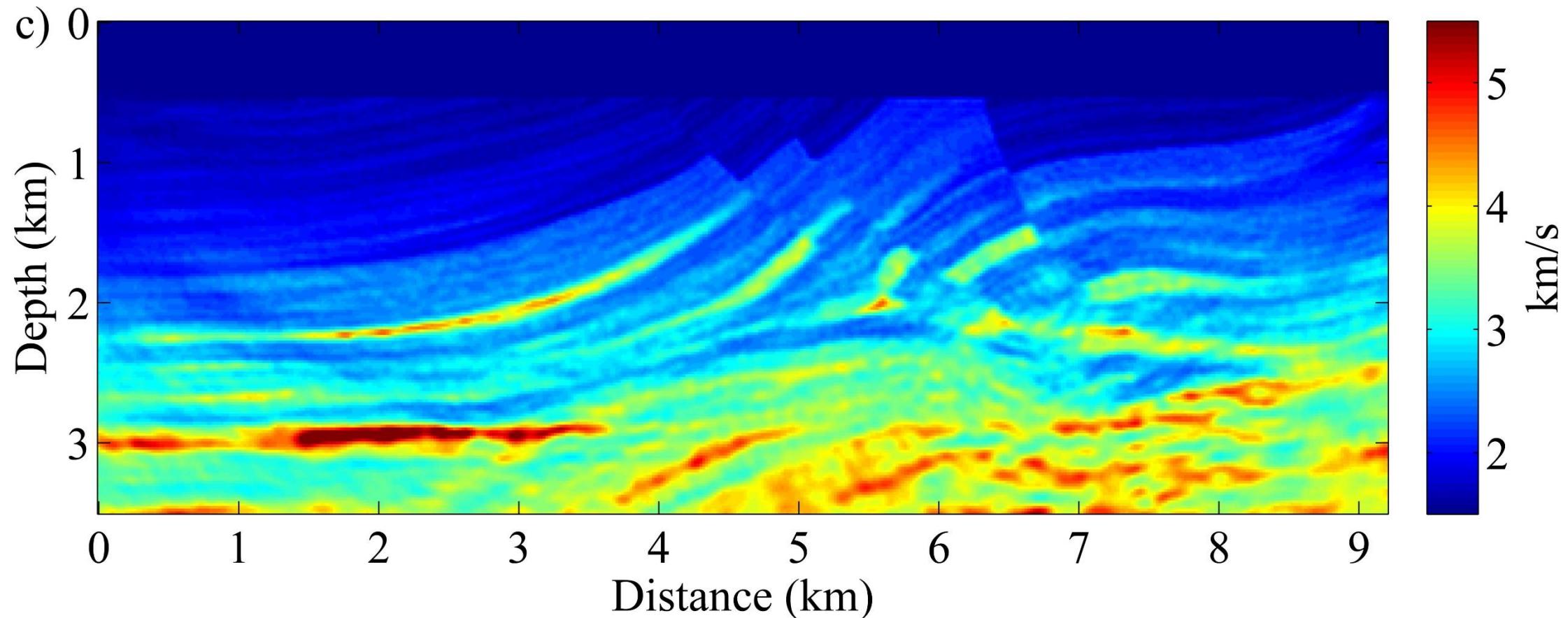
# FWI result with 10 iterations for 3 frequencies (4-15Hz)



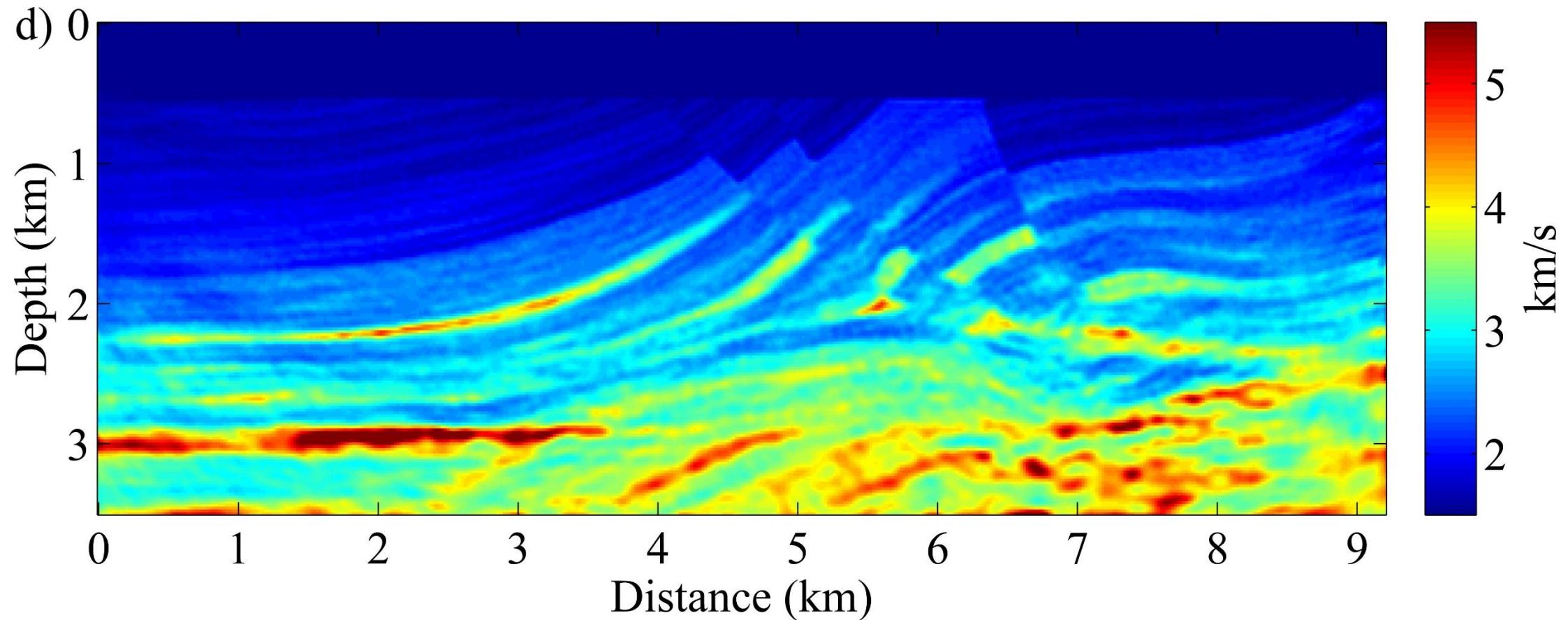
# FOFWI result with 10 iterations for 3 frequencies (4-15Hz)



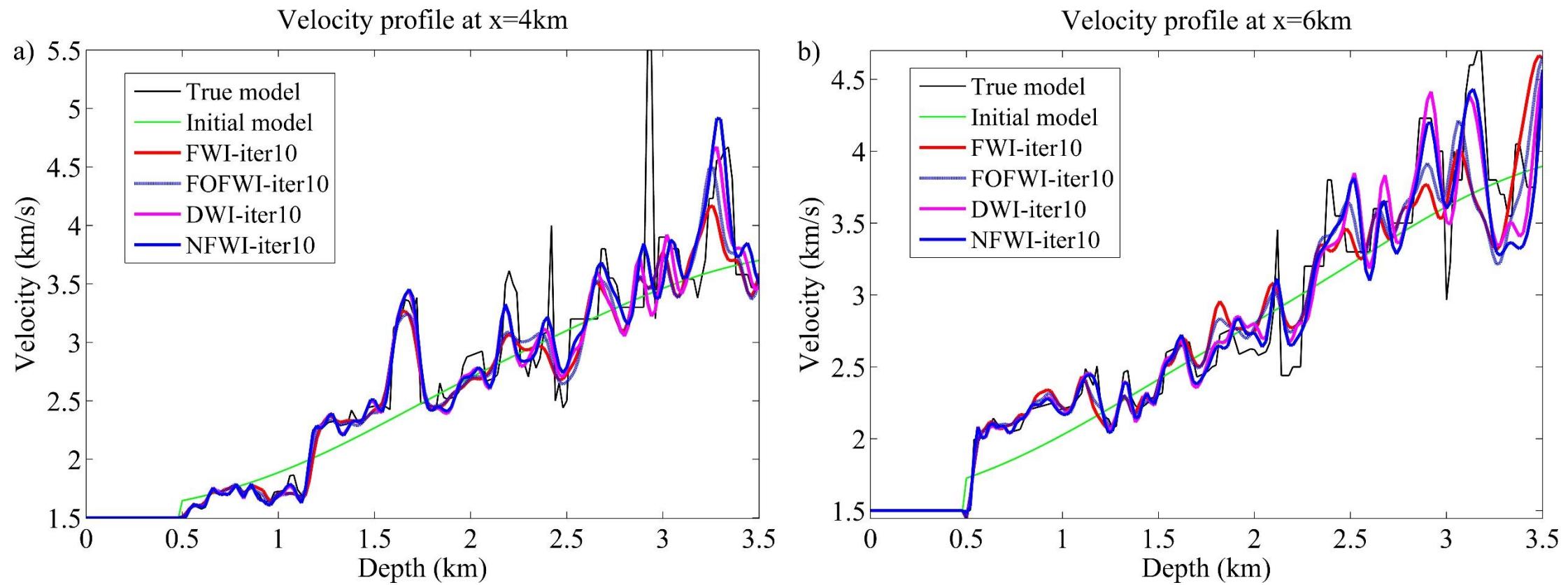
# DWI result with 10 iterations for 3 frequencies (4-15Hz)



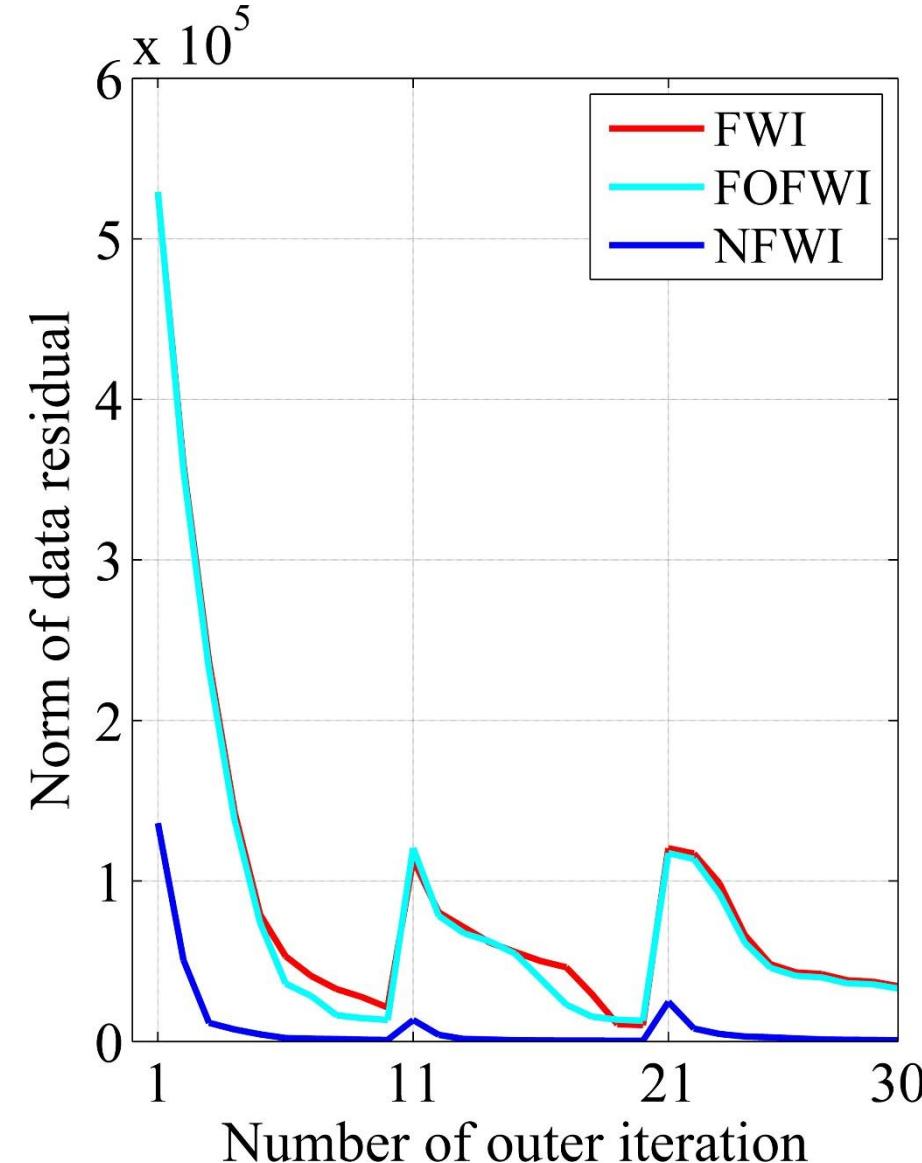
# NFWI result with 10 iterations for 3 frequencies (4-15Hz)



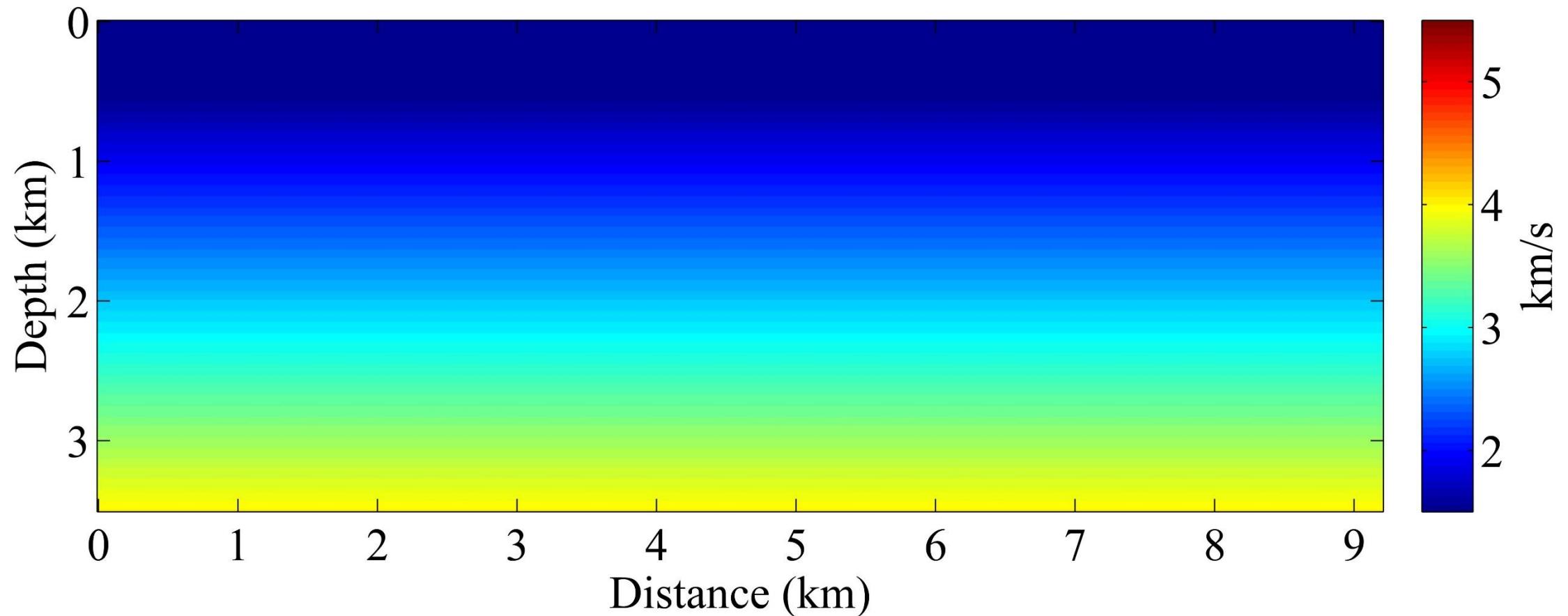
# Velocity profiles



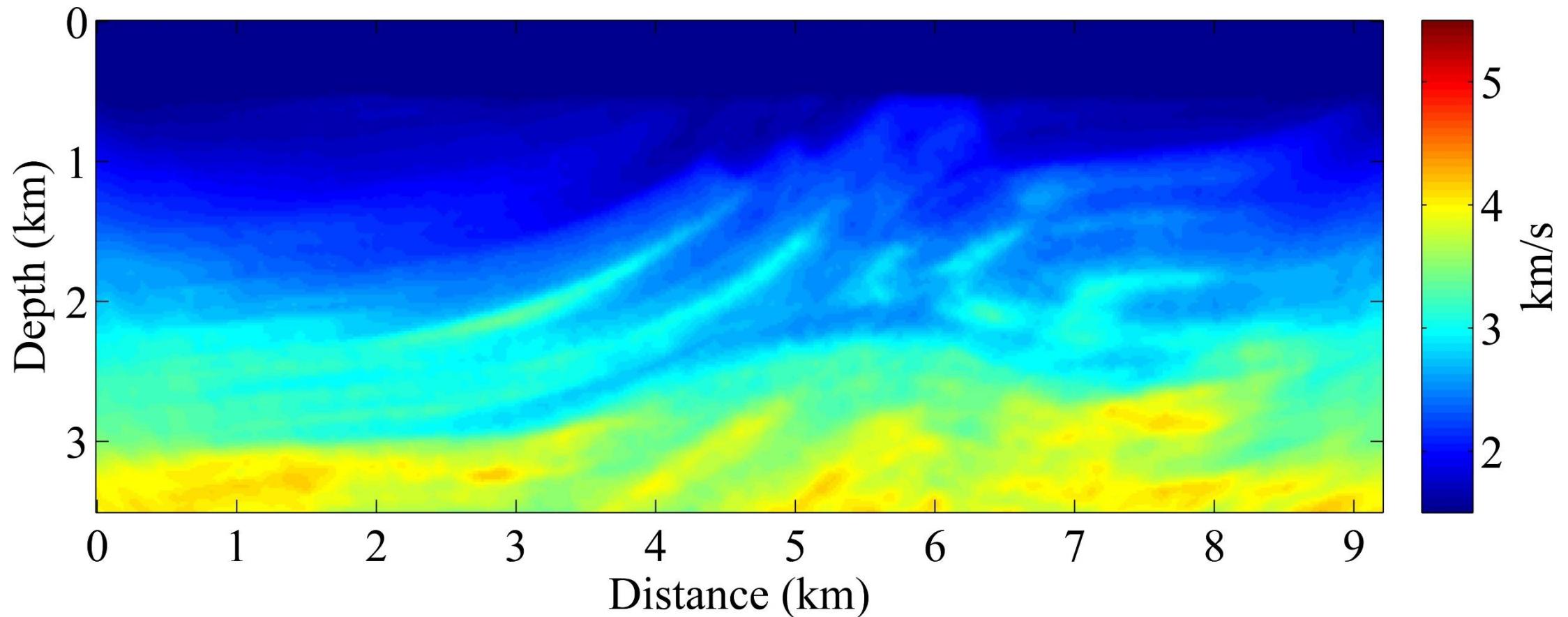
# Data residual vs iteration number



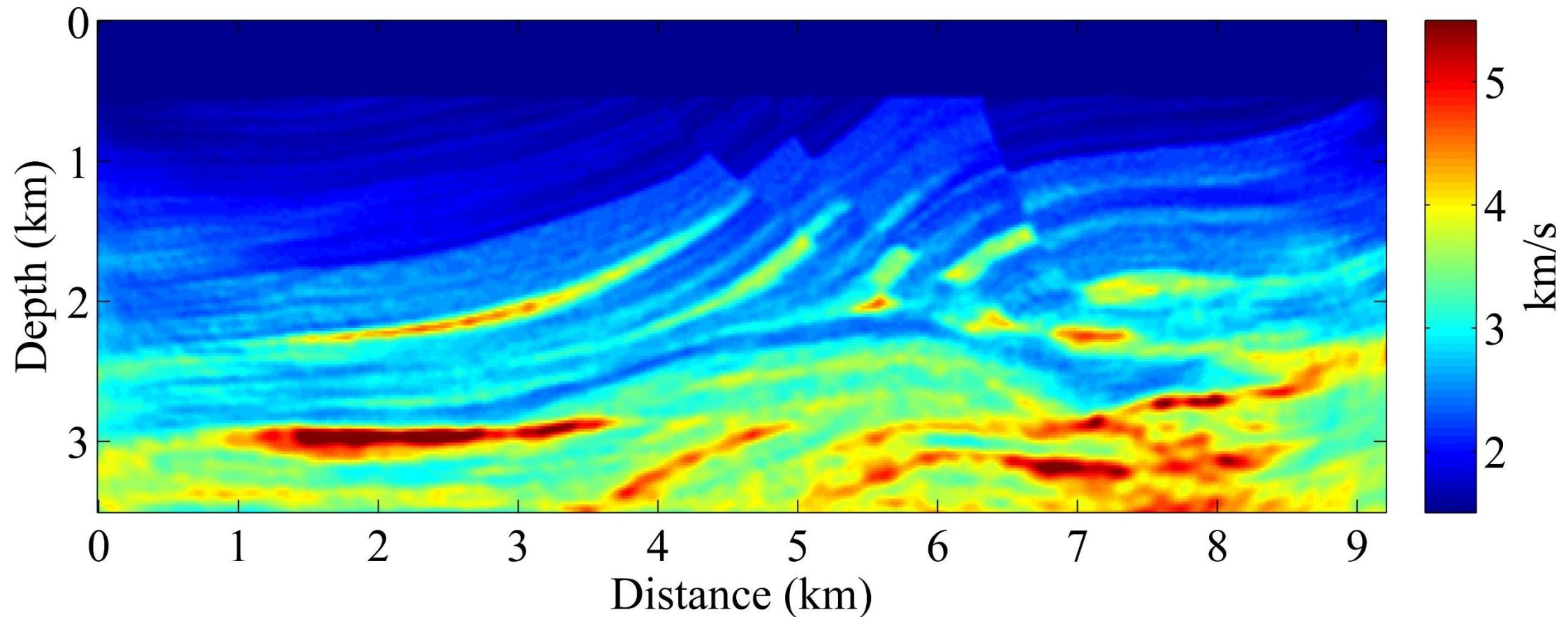
# Initial model (linearly changed from 1.5km/s to 4km/s along z)



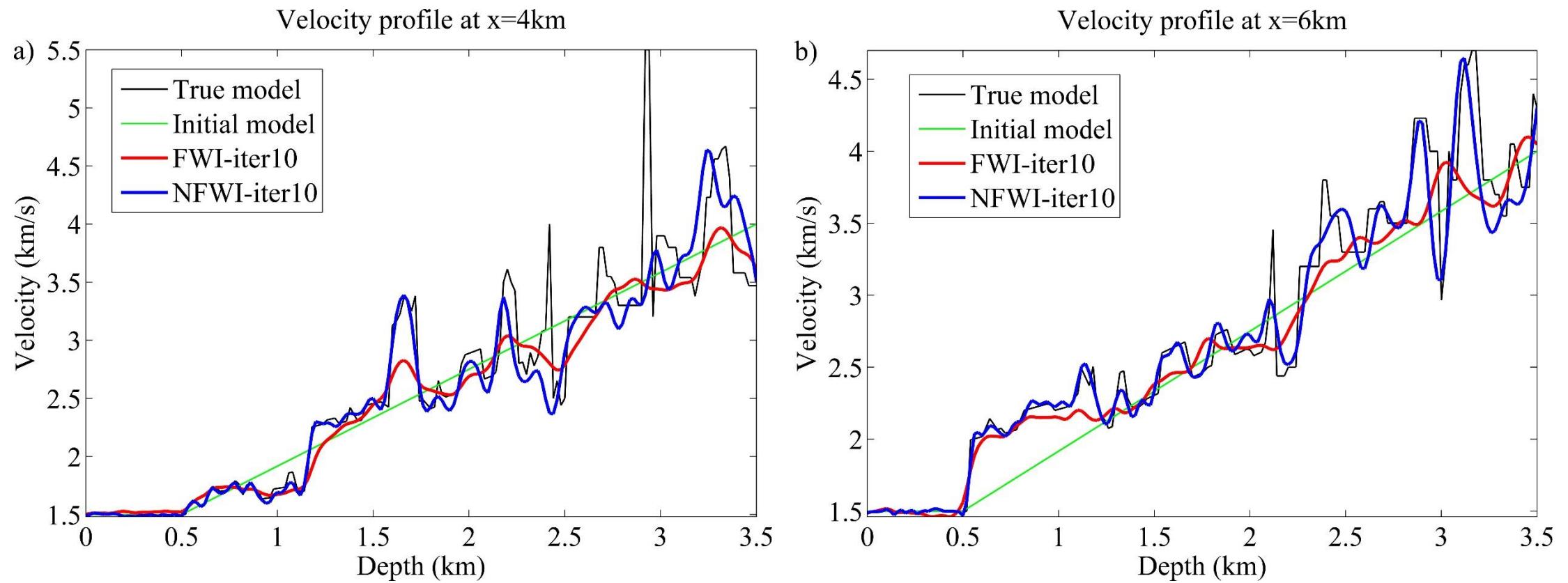
# FWI result with 10 iterations for 5 frequencies (2-15Hz)



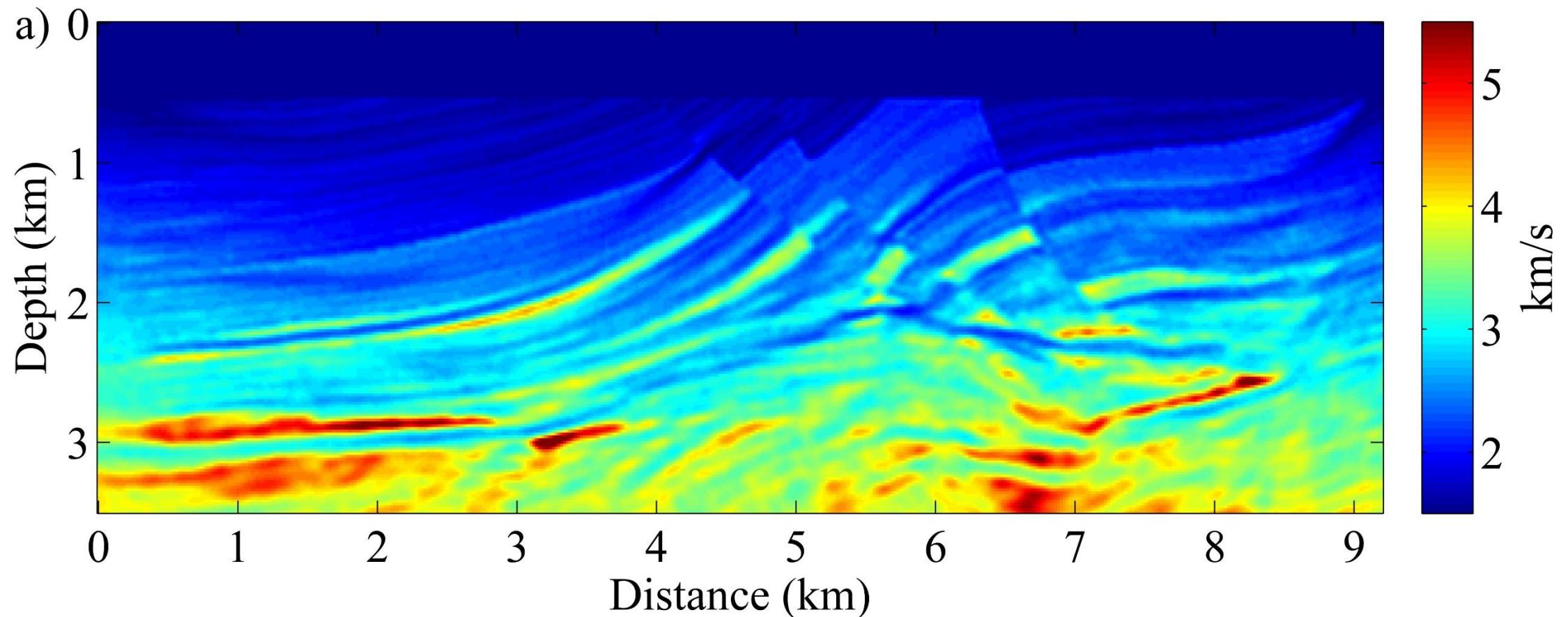
# NFWI result with 10 iterations for 5 frequencies (2-15Hz)



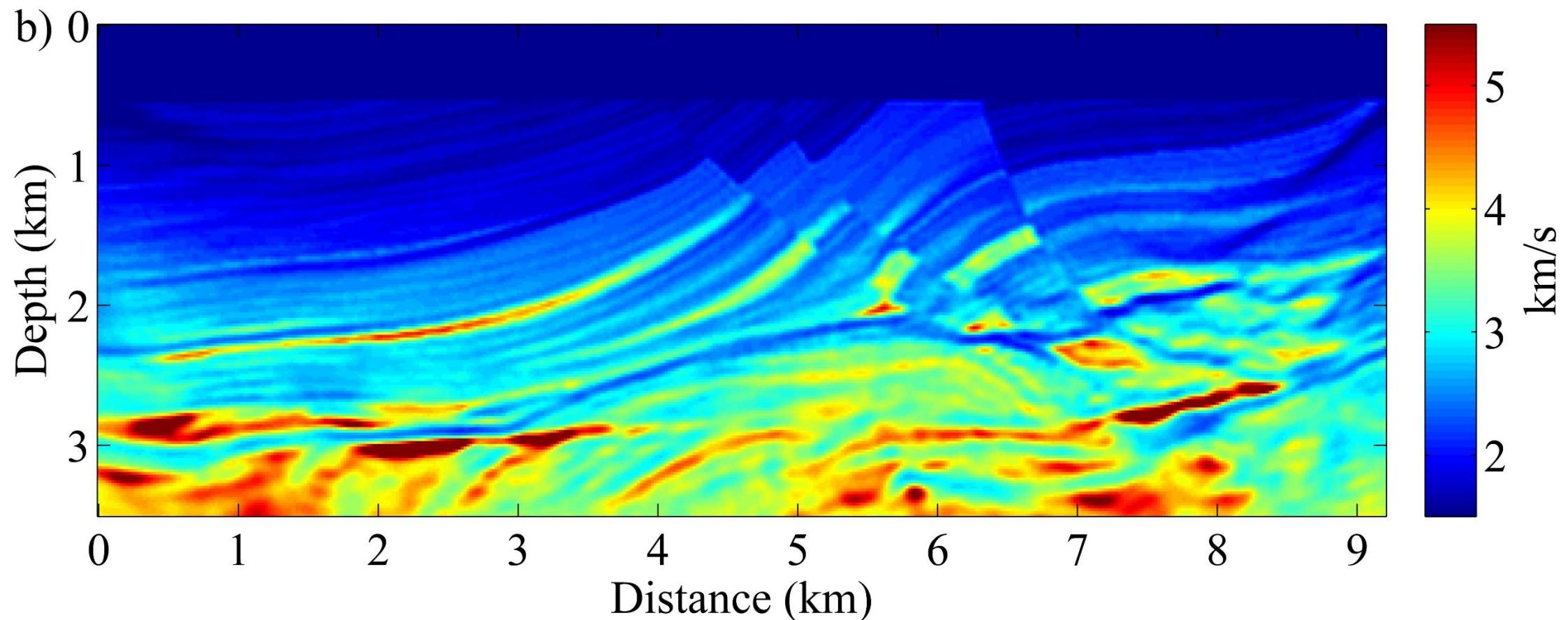
# Velocity profiles



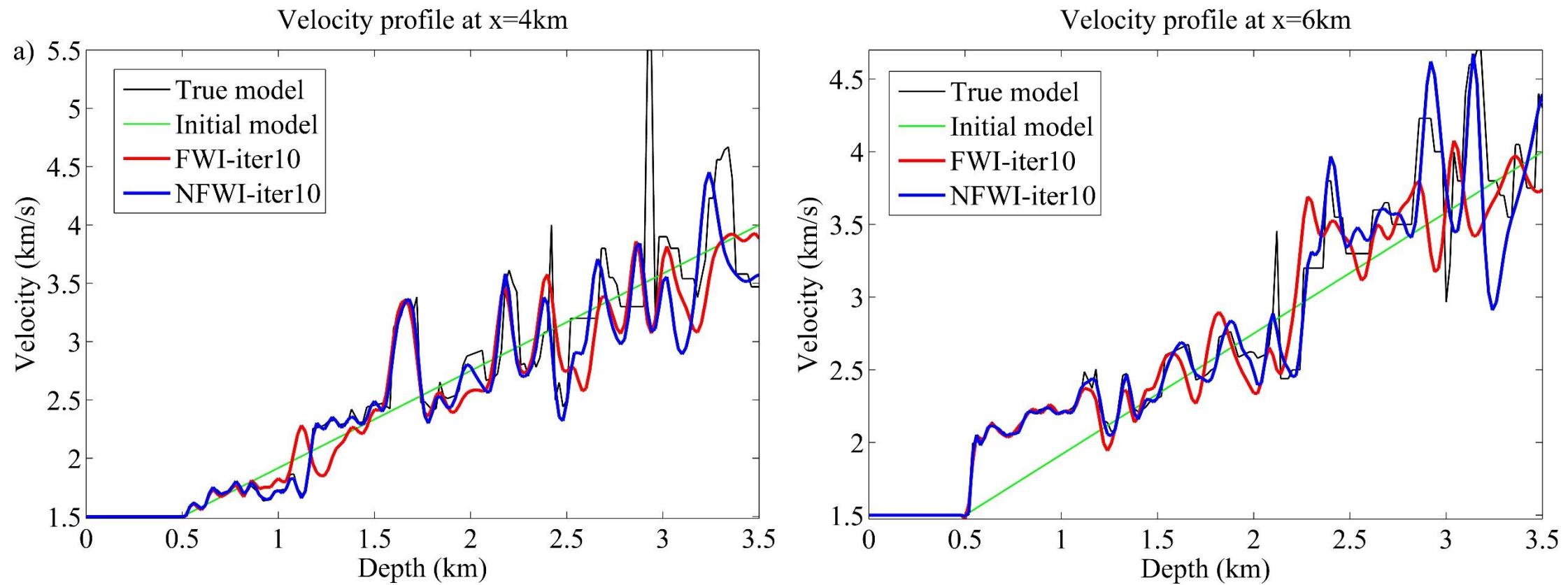
# FWI result with 10 iterations for 12 frequencies (4-15Hz)



# NFWI result with 10 iterations for 12 frequencies (4-15Hz)



# Velocity profiles



# Conclusions

- Nonlinear sensitivities provide the possibility to better handle the nonlinearity in FWI
  - Update the long wavelength components in the deeper regions as well as the shallow regions
  - Help to converge faster
- A two-iteration nonlinear FWI approach
  - Using data residual from the current iteration to perform a linear inversion to update perturbation
  - Using perturbation to construct gradient

# Acknowledgements

- All CREWES Sponsors
- NSERC
- Wenyong Pan, Junxiao Li, Jian Sun, Huaizhen Chen, Khalid Almutteri
- All the other researchers in CREWES

# Thank You