

## Education

**Rice University**

09/2012 - Present

Ph.D. Candidate in **Geophysics**, Earth Science Department

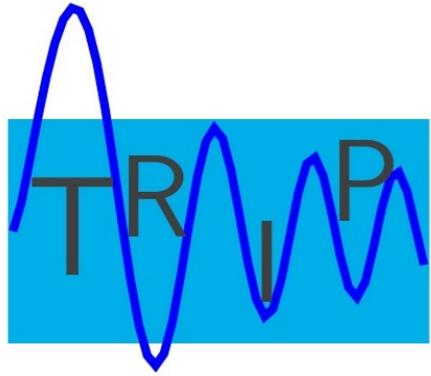
**China University of Petroleum (East China)**

2008 - 2012

B.S. in **Exploration Geophysics**

## Research Interests

- True Amplitude Imaging (RTM)
- Acceleration of linear/nonlinear waveform inversion
- Inversion Velocity Analysis



# Inversion Velocity Analysis

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with approximate Born inversion

Jie Hou

TRIP 2015 Review Meeting

Apr 25, 2016

# FWI

Full Waveform Inversion (FWI) finds velocity model to minimize the data misfit:

$$J_{\text{FWI}}[\mathbf{m}] = \frac{1}{2} \|\mathcal{F}[\mathbf{m}] - \mathbf{d}_{\text{obs}}\|^2$$

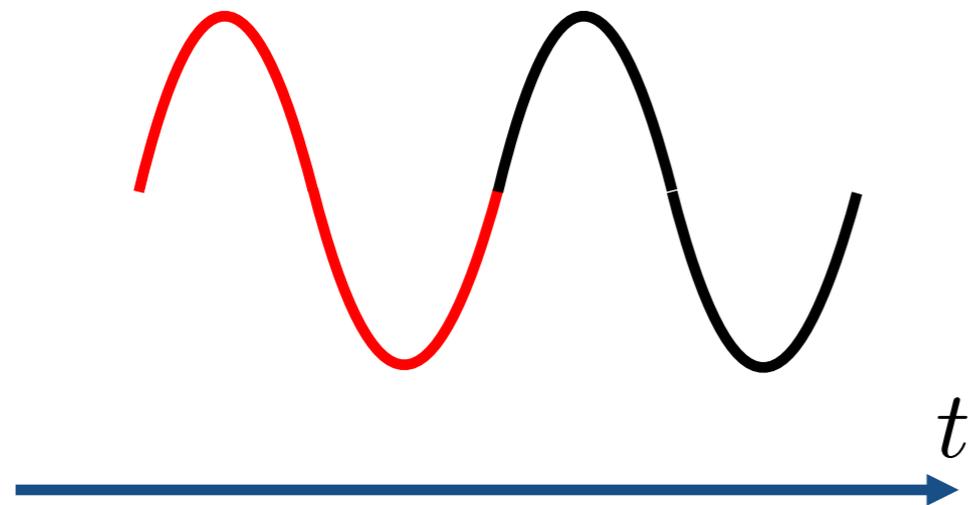
$\mathbf{m}$  velocity model       $\mathbf{d}_{\text{obs}}$  observed data

$\mathcal{F}$  forward modeling operator

- Extends into any modeling physics, data geometry
- Large scale, nonlinear, ill-posed problem

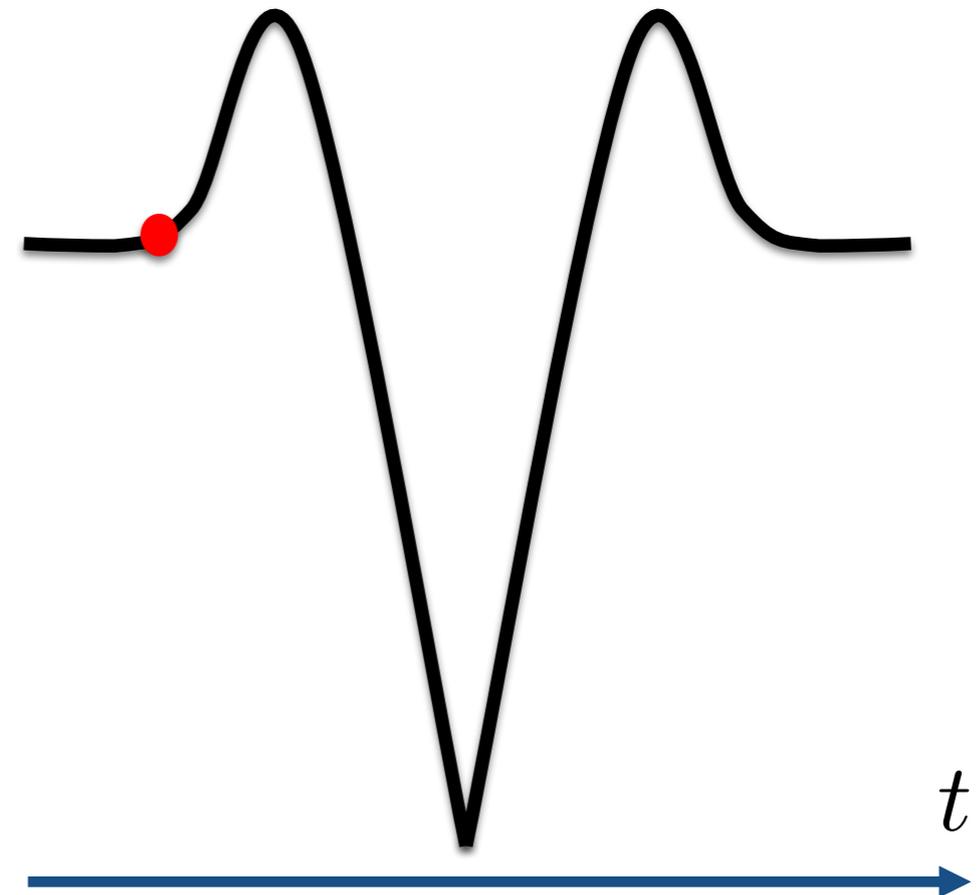
# Behavior of Objective Function

## Simple Example:



— Modeled Data

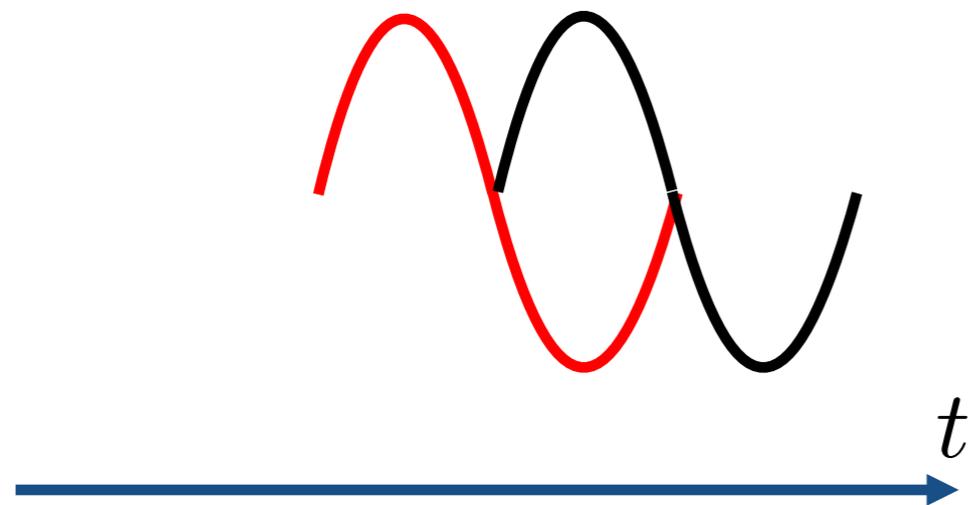
— Observed Data



Objective Function

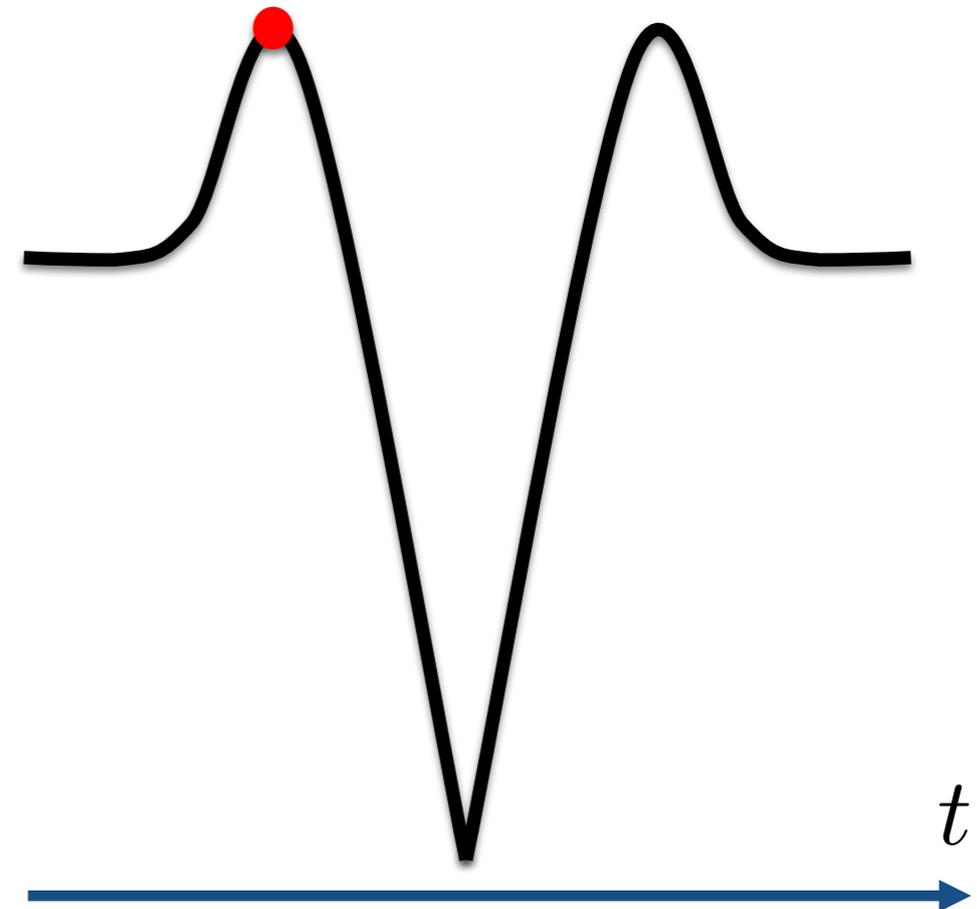
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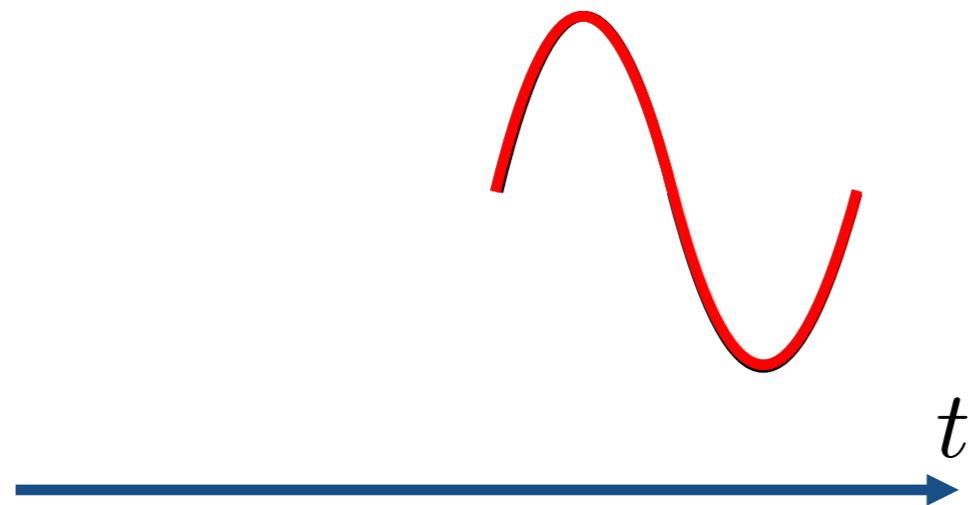
— Observed Data



Objective Function

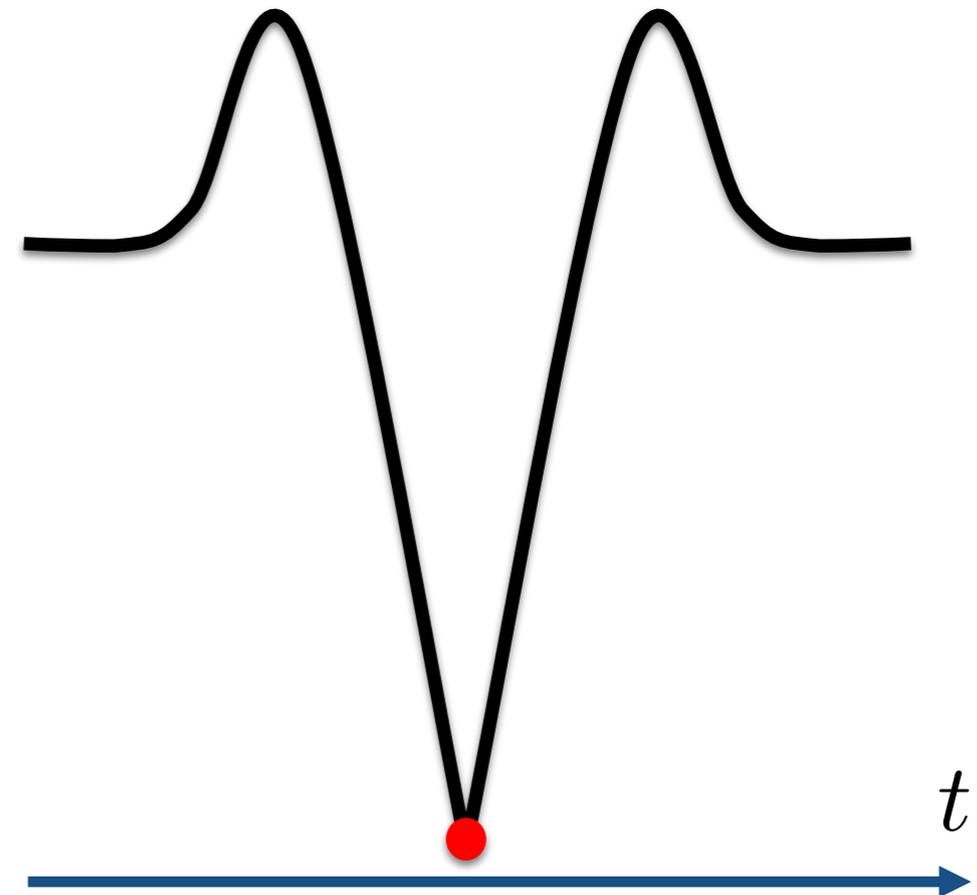
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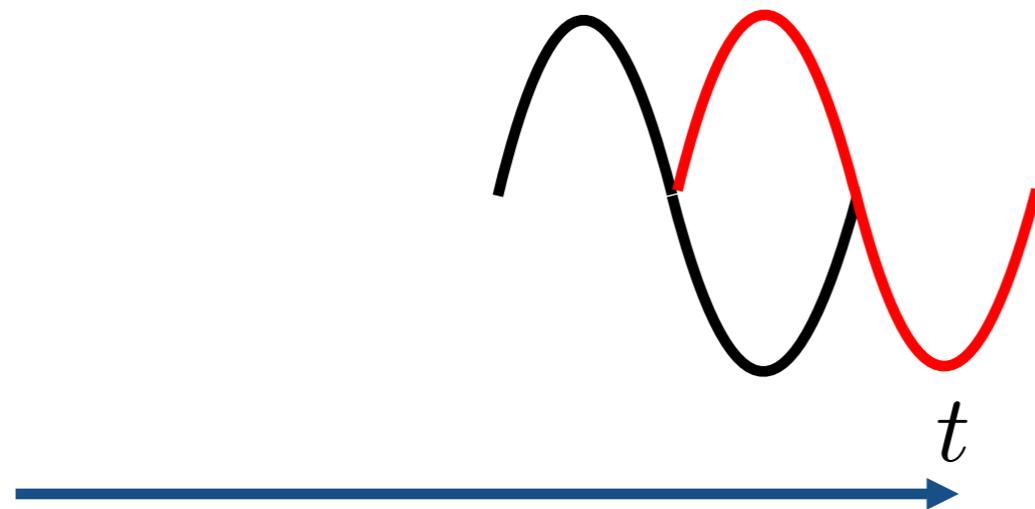
— Observed Data



Objective Function

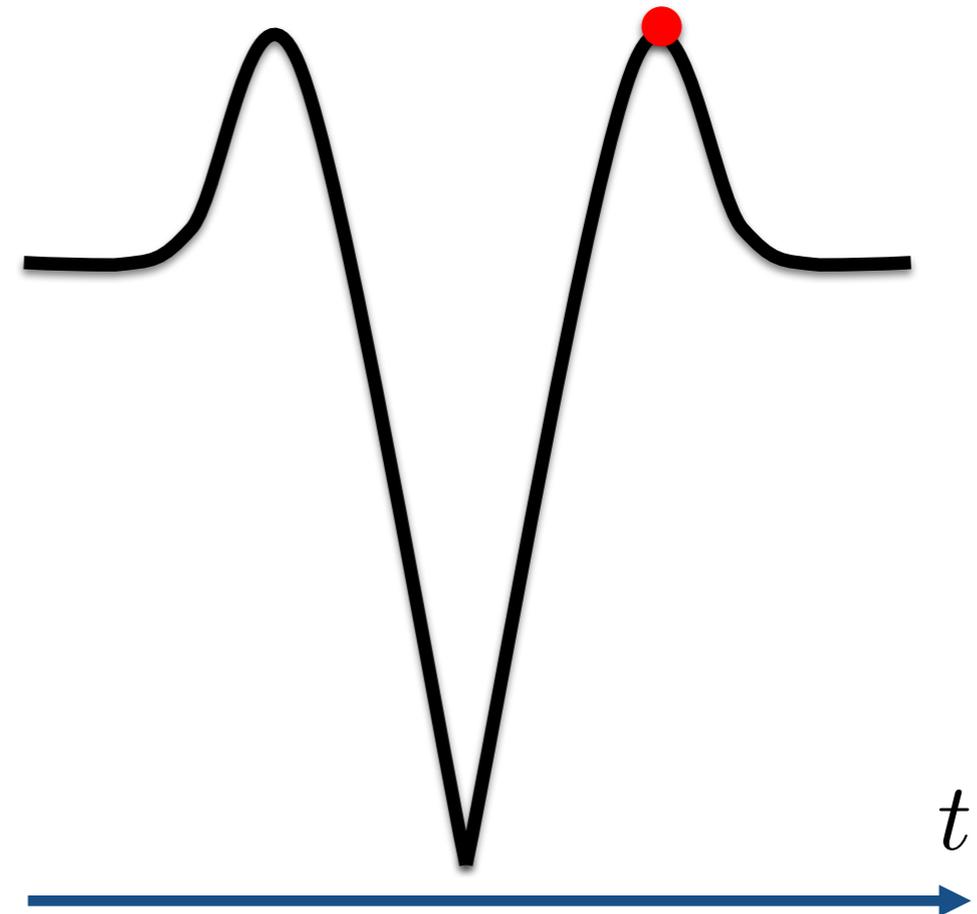
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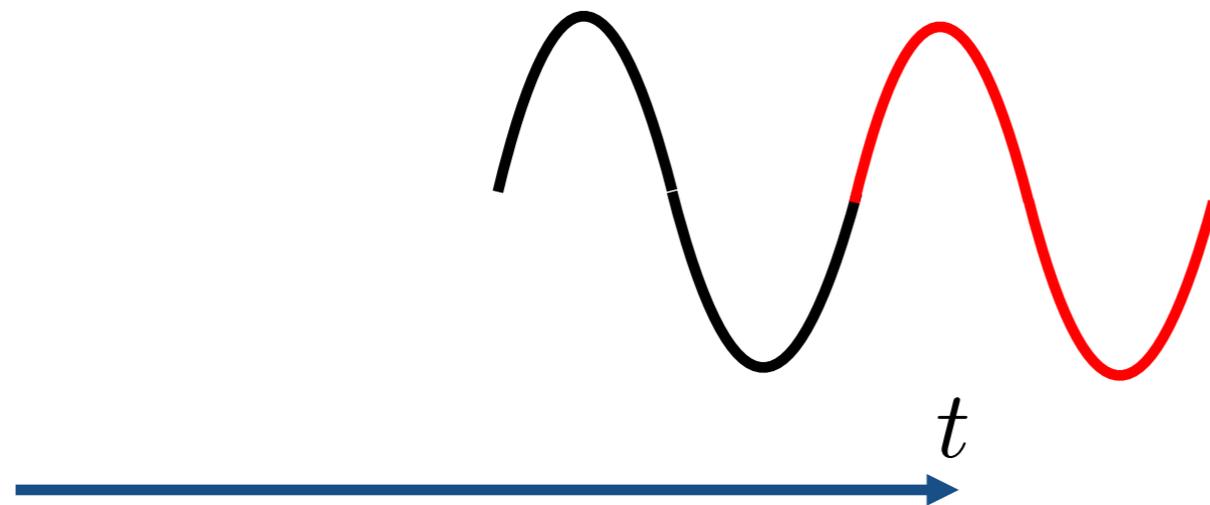
— Observed Data



Objective Function

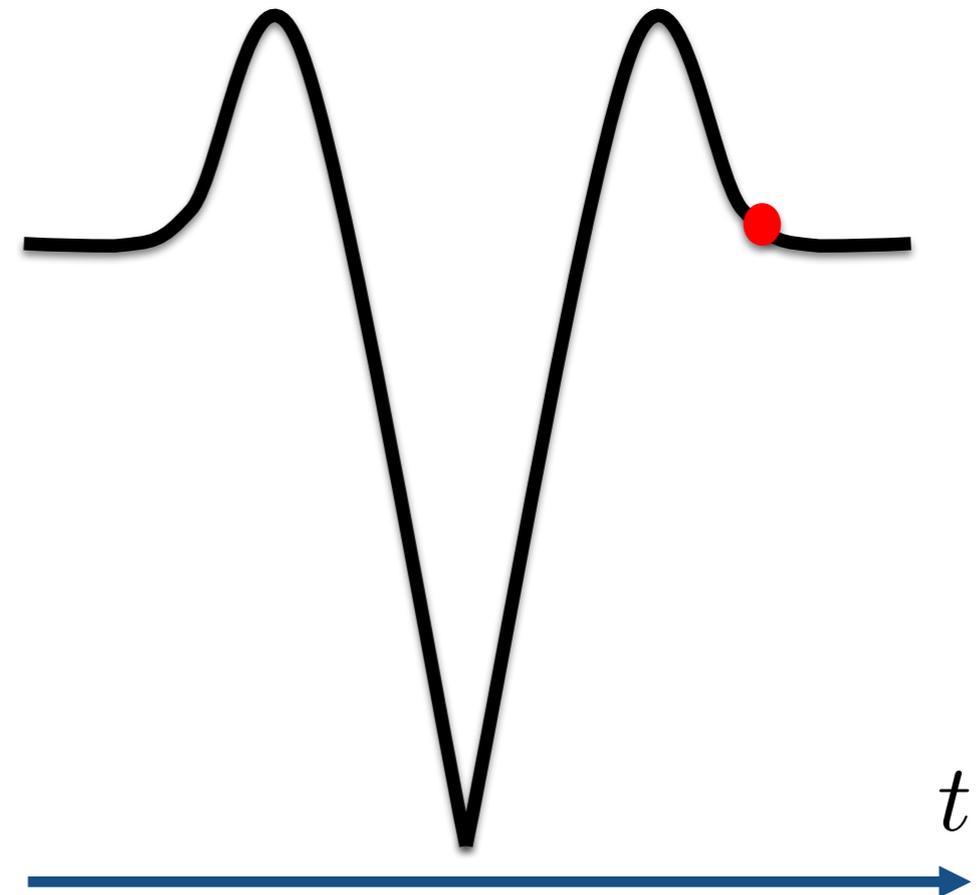
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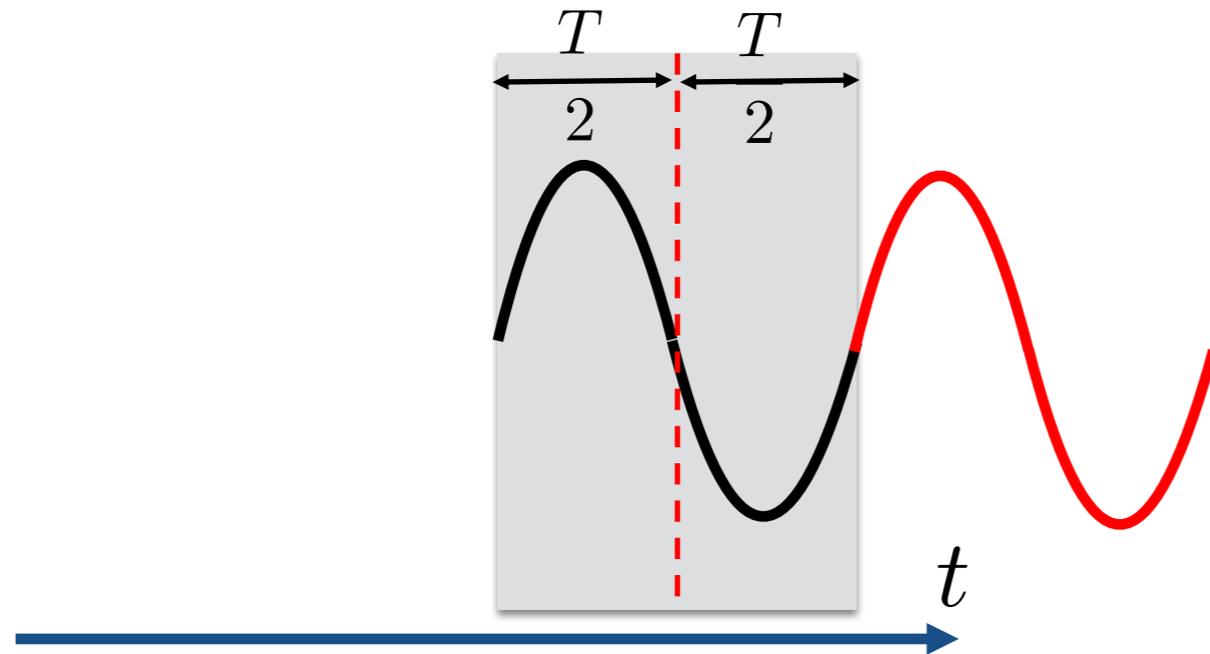
— Observed Data



Objective Function

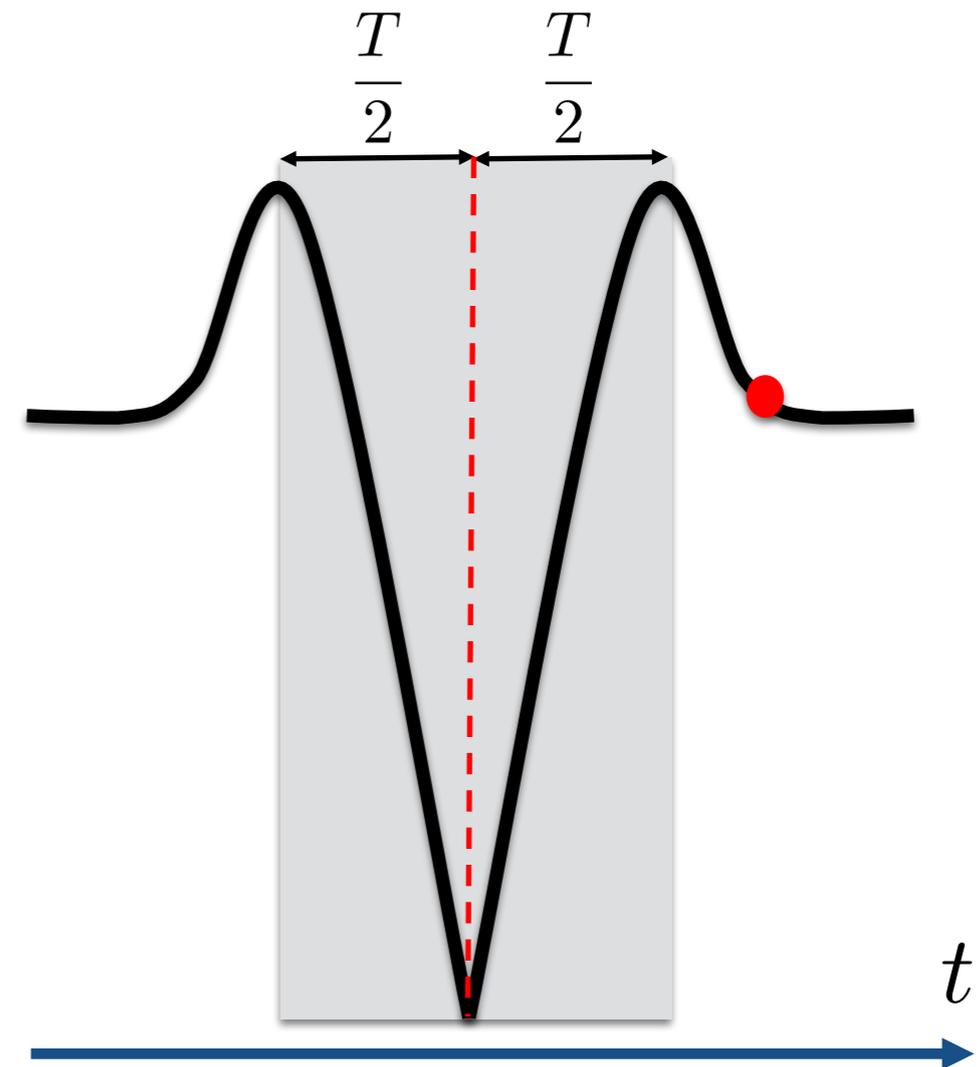
# Behavior of Objective Function

## Simple Example:



— Modeled Data

— Observed Data

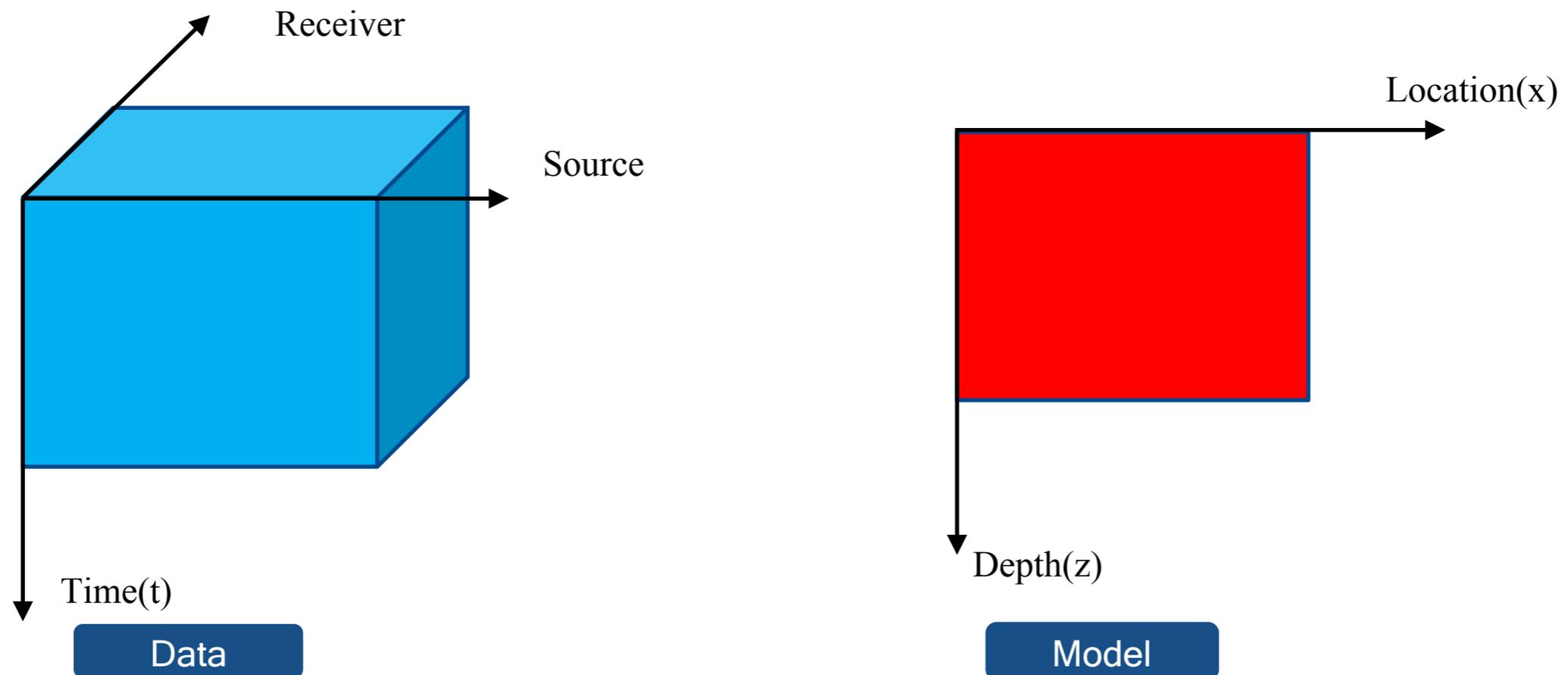


Objective Function

- “Cycle Skipping” problem (local minima)
- start within half wavelength:  
good initial, low frequency, far offset

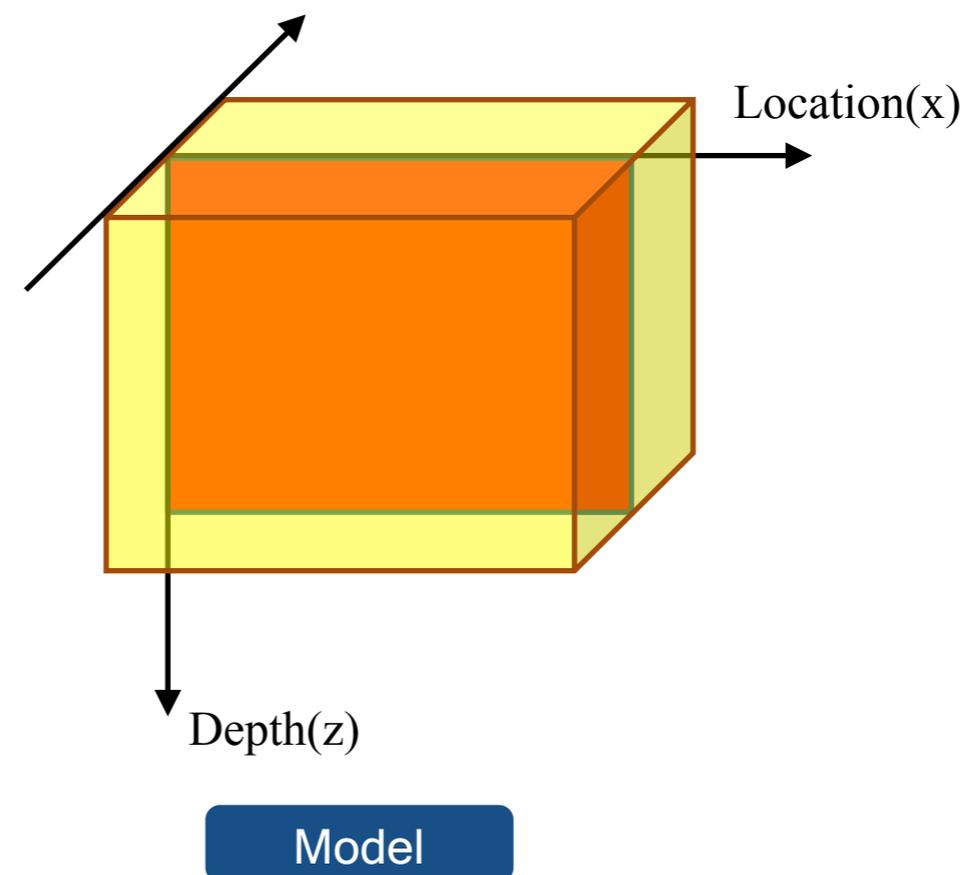
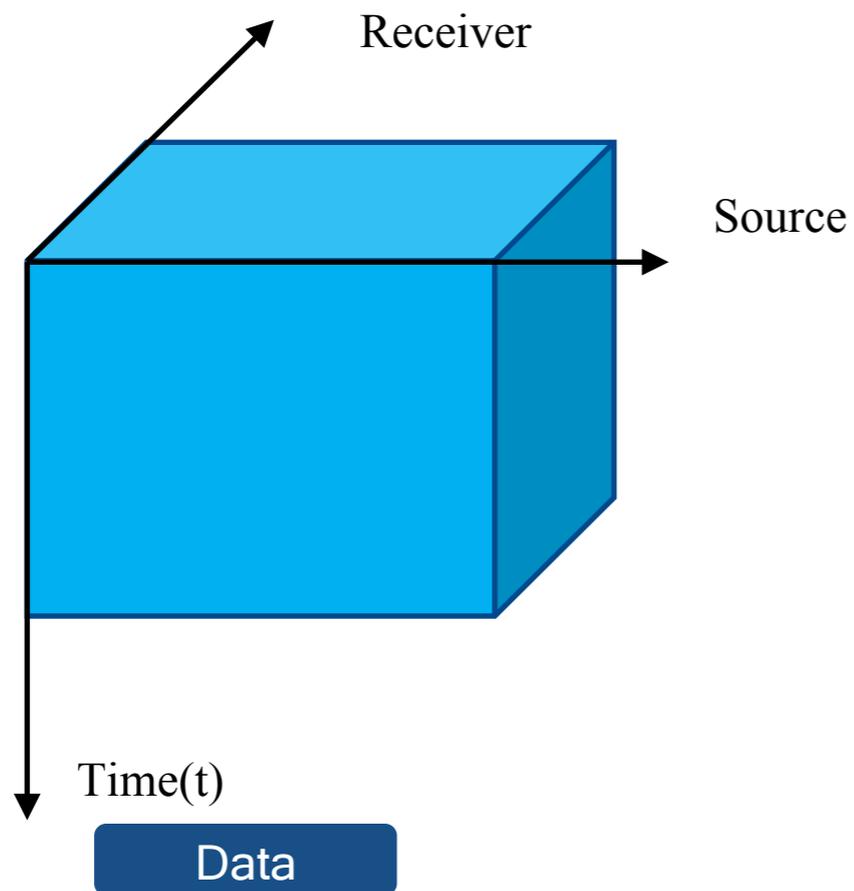
# Model Extension

- Our **goal**: immune to “cycle skipping”
- Solution: Hug the data!!!



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- Solution: Hug the data!!!



# Model Extension

➤ Extended Model:  $\mathcal{M} \rightarrow \bar{\mathcal{M}}$

➤ Extended Modeling

$$\bar{\mathcal{F}} : \bar{\mathcal{M}} \rightarrow \mathcal{D} \quad \mathcal{F}[\mathbf{m}] = \bar{\mathcal{F}}[\bar{\mathbf{m}}]$$

➤ Extension Parameter:

**Many choices:** surface / subsurface offset, source, reflection angle...

➤ Extended model are not **physical**

**Model Extension + Physical Constraint**

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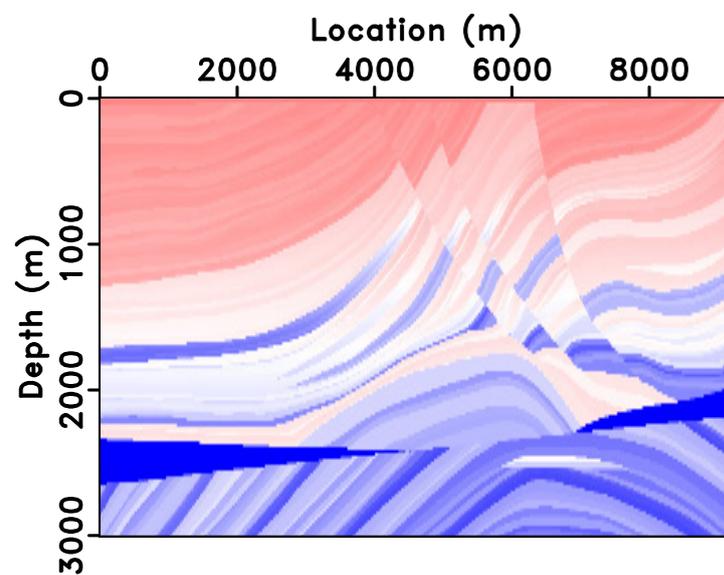
➤ Extended model are not **physical**

**Model Extension + Physical Constraint**

# Model Separation

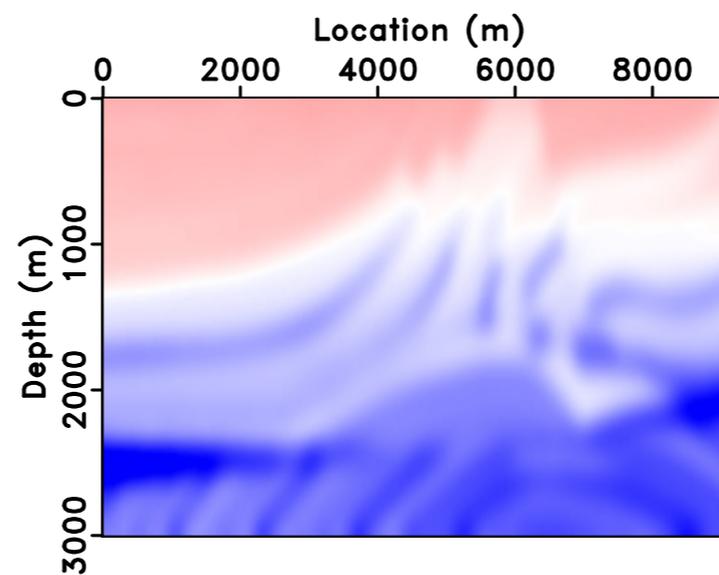
- Separate model into two parts

$$\mathbf{m} = \mathbf{m}_0 + \delta\mathbf{m}$$



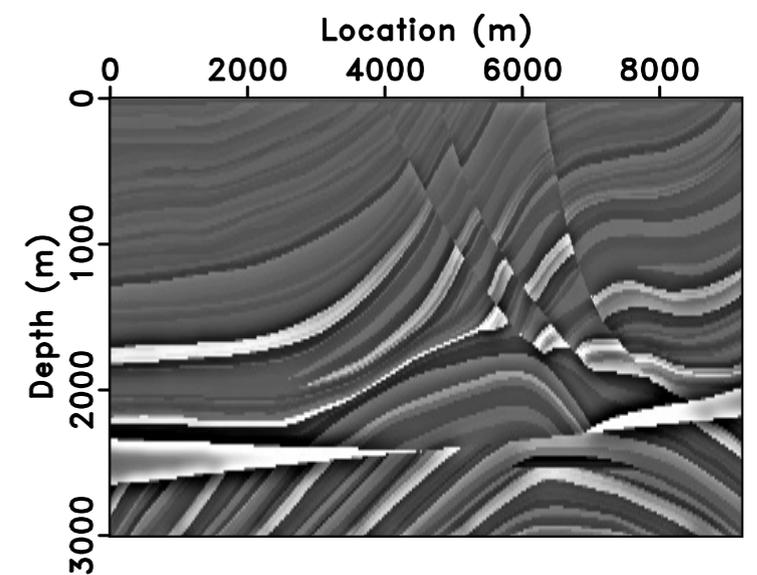
Velocity Model  $\mathbf{m}$

=



Background Model  $\mathbf{m}_0$

+



Reflectivity Model  $\delta\mathbf{m}$

- Two-step Problem

$$\mathcal{F}[\mathbf{m}] \approx \mathcal{F}[\mathbf{m}_0] + F[\mathbf{m}_0]\delta\mathbf{m}$$

Born Modeling Operator

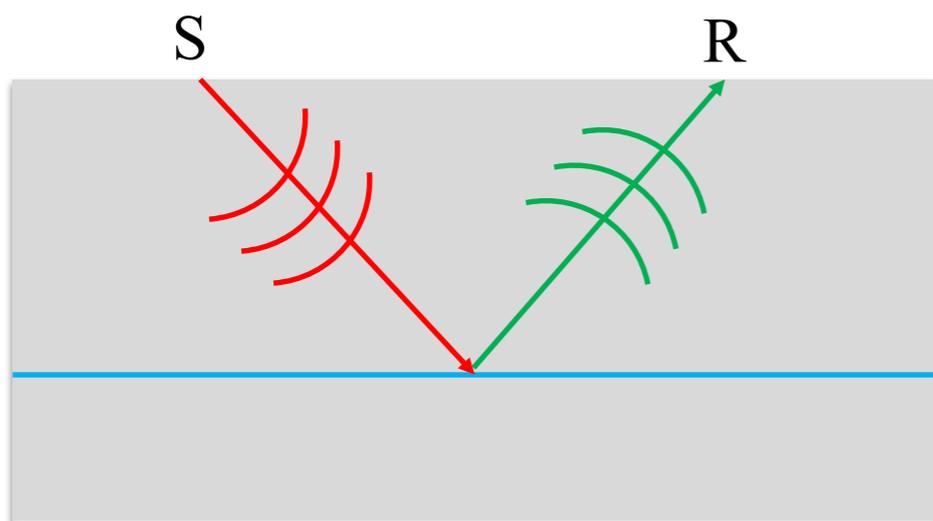
# Born modeling and its adjoint

## ➤ Born modeling (single scattering) operator

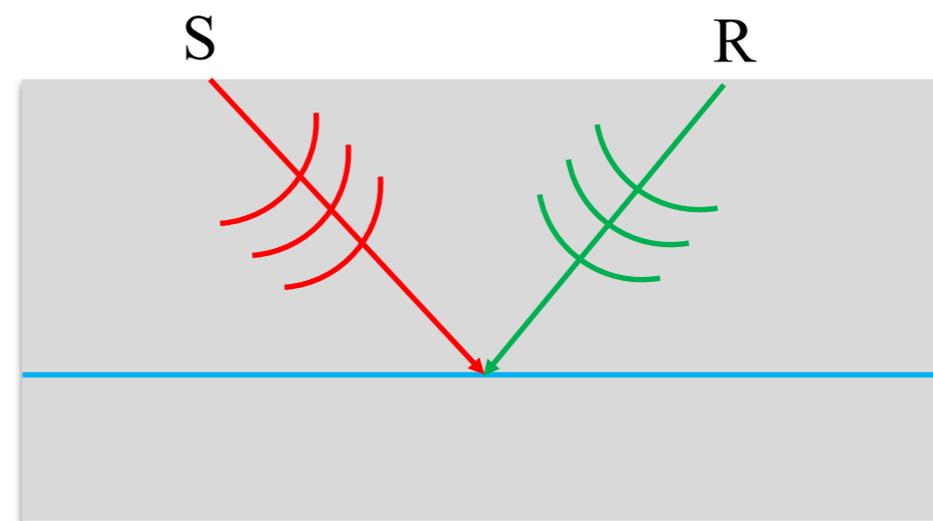
$$(F[v_0]\delta v)(\mathbf{x}_s, \mathbf{x}_r, t) = \frac{\partial^2}{\partial t^2} \int d\mathbf{x}d\tau G(\mathbf{x}_s, \mathbf{x}, \tau) \frac{2\delta v(\mathbf{x})}{v_0(\mathbf{x})^3} G(\mathbf{x}, \mathbf{x}_r, t - \tau)$$

## ➤ Adjoint Operator

$$(F^*[v_0]\delta d)(\mathbf{x}) = \frac{2}{v_0(\mathbf{x})^3} \int d\mathbf{x}_s d\mathbf{x}_r dt d\tau G(\mathbf{x}_s, \mathbf{x}, \tau) G(\mathbf{x}, \mathbf{x}_r, t - \tau) \frac{\partial^2}{\partial t^2} \delta d(\mathbf{x}_s, \mathbf{x}_r, t)$$



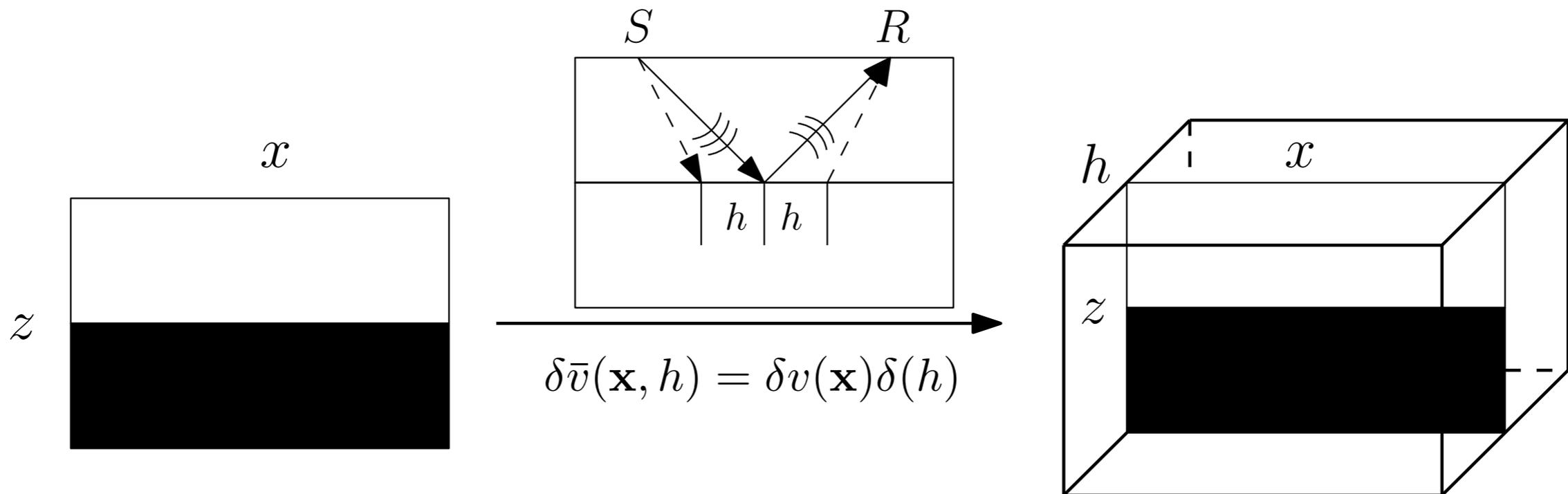
Born Modeling



Adjoint Operator

# Model Extension + Model Separation

## ➤ Subsurface offset extension



Physical Model

Extended Model

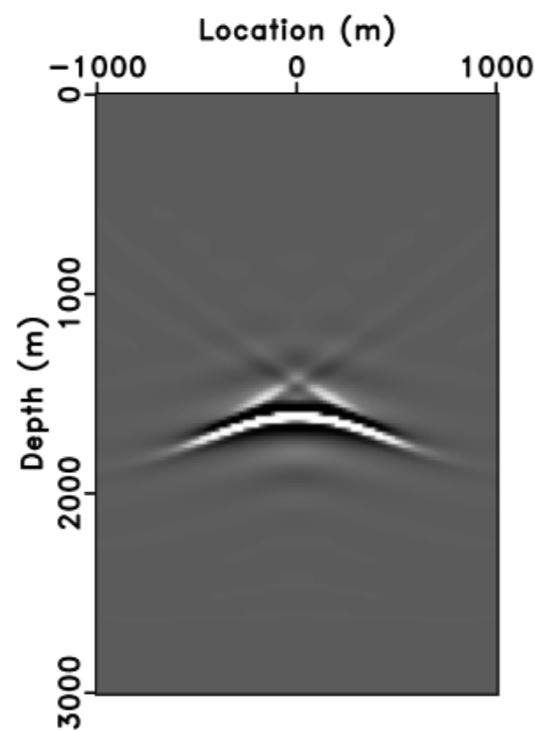
## ➤ Physical Meaning

Action at a distance: stress leads to strain at a distance

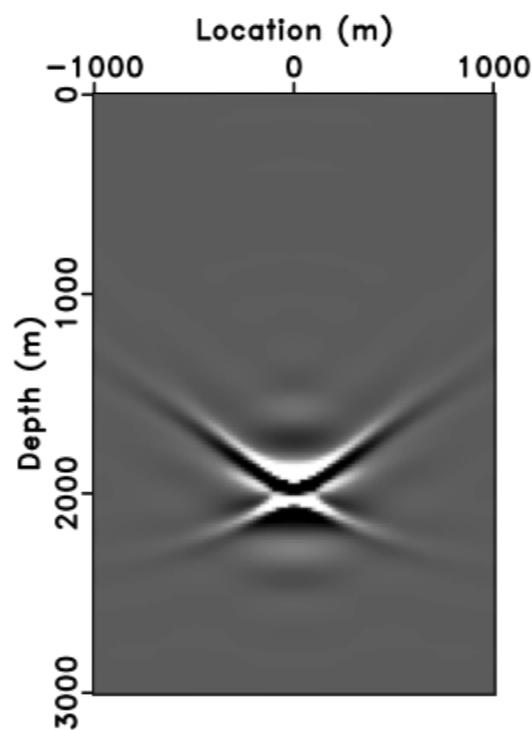
# Extended Born modeling Operator

## ➤ Extended Born modeling operator and its adjoint

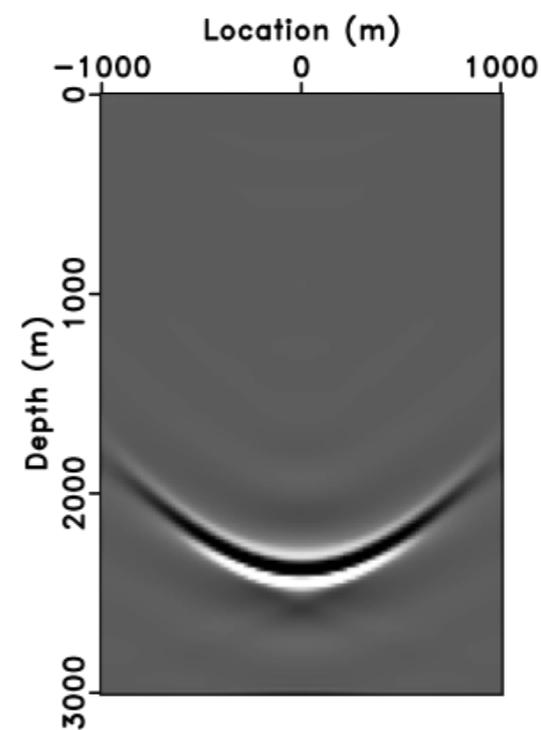
$$(\bar{F}[v_0]\delta\bar{v})(\mathbf{x}_s, \mathbf{x}_r, t) = \frac{\partial^2}{\partial t^2} \int d\mathbf{x}d\mathbf{h}d\tau G(\mathbf{x}_s, \mathbf{x} - \mathbf{h}, \tau) \frac{2\delta\bar{v}(\mathbf{x}, \mathbf{h})}{v_0(\mathbf{x})^3} G(\mathbf{x} + \mathbf{h}, \mathbf{x}_r, t - \tau)$$
$$(\bar{F}^*[v_0]\delta d)(\mathbf{x}, \mathbf{h}) = \frac{2}{v_0(\mathbf{x})^3} \int d\mathbf{x}_s d\mathbf{x}_r dt d\tau G(\mathbf{x}_s, \mathbf{x} - \mathbf{h}, \tau) G(\mathbf{x} + \mathbf{h}, \mathbf{x}_r, t - \tau) \frac{\partial^2}{\partial t^2} \delta d(\mathbf{x}_s, \mathbf{x}_r, t)$$



V=2km/s



V=2.5km/s



V=3km/s

Correct Velocity Model

## ➤ Partially Linearized Problem

$$\bar{F}[\mathbf{m}_0]\delta\bar{\mathbf{m}} \simeq \mathbf{d} - \mathcal{F}[\mathbf{m}_0]$$

- Find  $\delta\bar{\mathbf{m}}$  to fit the data
- Find  $\mathbf{m}_0$  to satisfy the **semblance condition**

## ➤ Migration Velocity Analysis (MVA) (Shen and Symes, 2003)

Update **velocity** based on migrated **image volume**

# MVA via DSO

$$J_{\text{MVA}}[\mathbf{m}_0] = \frac{1}{2} \|AI(x, z, h)\|^2$$

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- $A$  : Annihilator,  $A=h$
- $I(x,z,h)$  can be computed via various migration

- $J$  is quadratic in image and data (implicitly), regardless of the frequency components
- Smooth in velocity

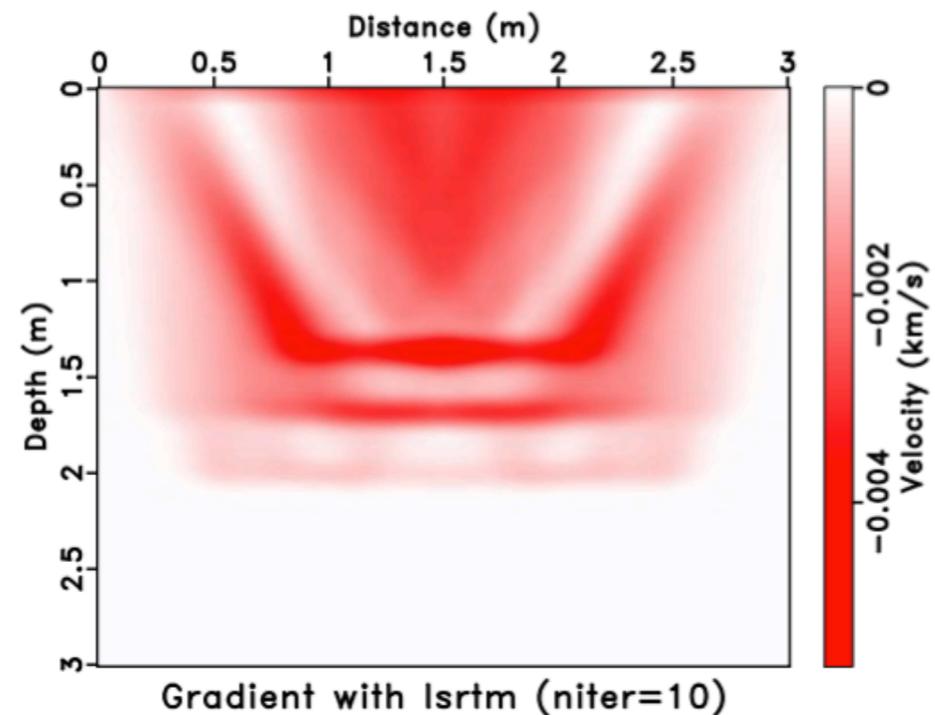
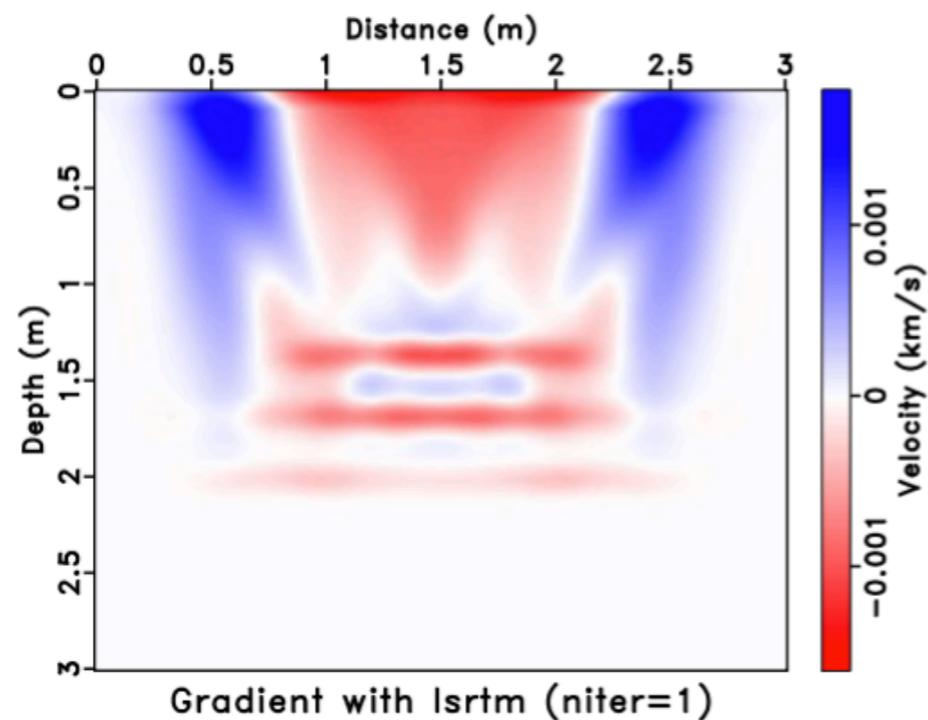
**Only choice** (Stolk & Symes, 2003)

# “Gradient Artifacts”

- Gradient artifacts : updating to the wrong direction

(Fei and Williamson, 2010; Vyas and Tang, 2010)

- Root reason: imperfect image volume



(Liu and Symes, 2014)

# Approximate Inverse Operator

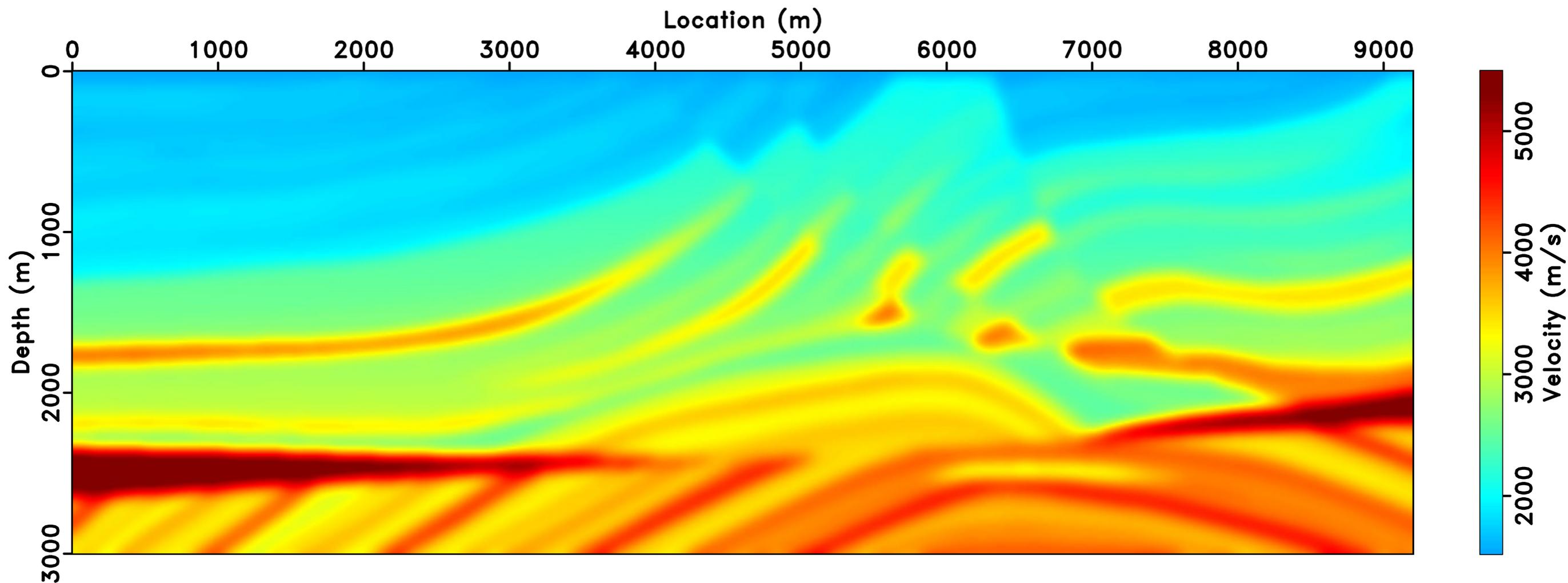
$$\bar{F}^\dagger \simeq W_{model}^{-1} \bar{F}^T W_{data}$$

(Ten Kroode, 2012; Hou and Symes, 2015)

- $W_{model}^{-1} = 4v_0^5 |k_{xz}| |k_{hz}|$   $W_{data} = I_t^4 D_{z_S} D_{z_R}$
- Derivation is based on **High Frequency Approx.**
- Implementation doesn't involve any **ray tracing**
- Invert the data even when **velocity is wrong**

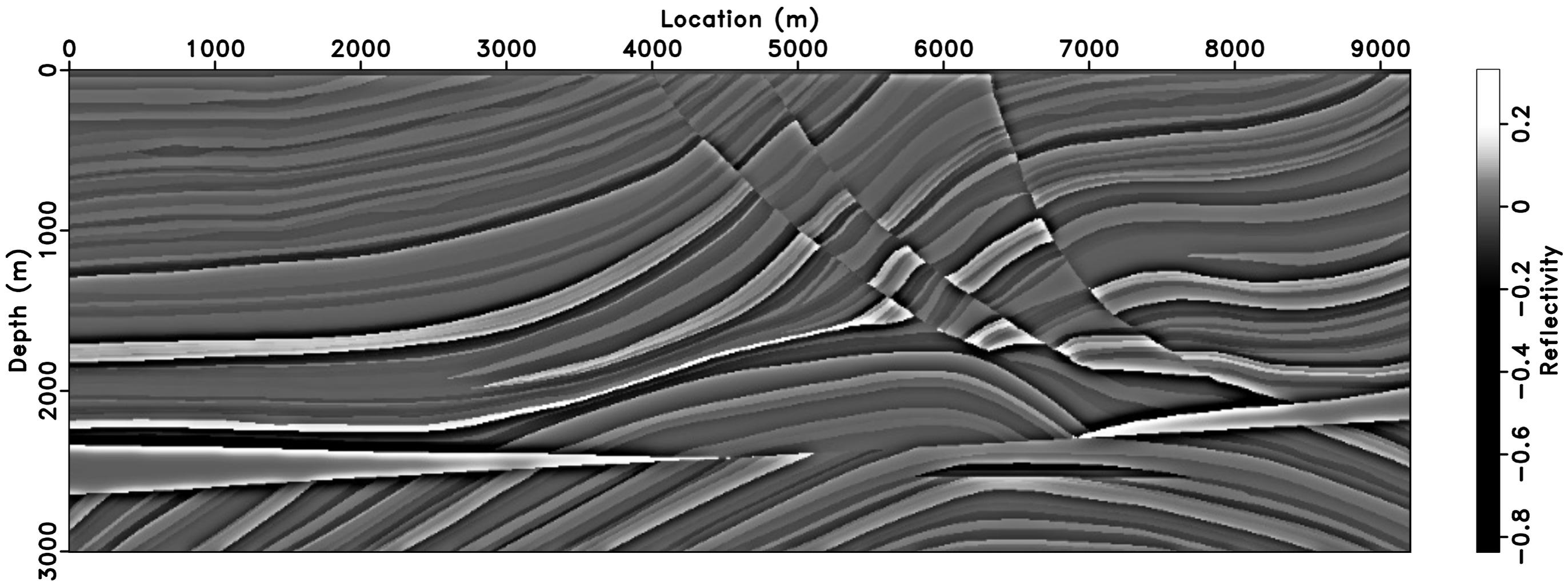
# Approximate Inverse Operator

- 2-8 finite difference, 231 shots & 461 receivers
- 2.5-5-30-35Hz Bandpass wavelet
- 1ms time sample, 10m grid interval



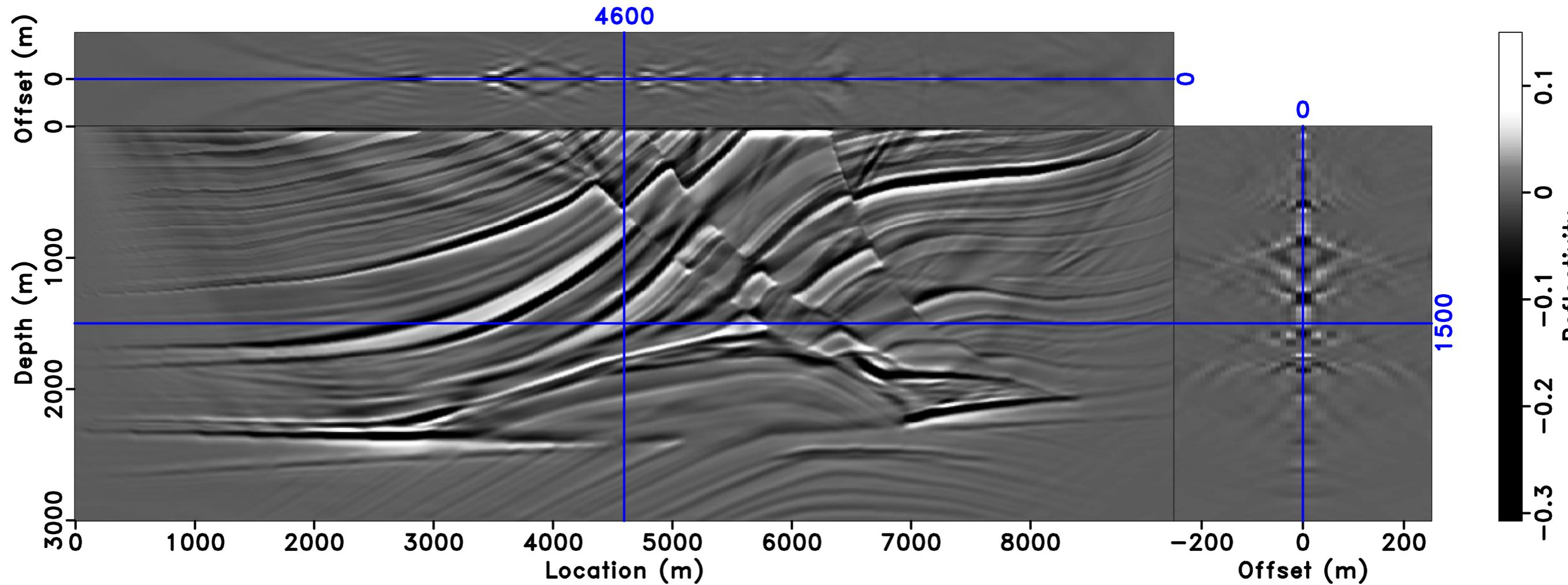
Background Velocity Model

# Approximate Inverse Operator



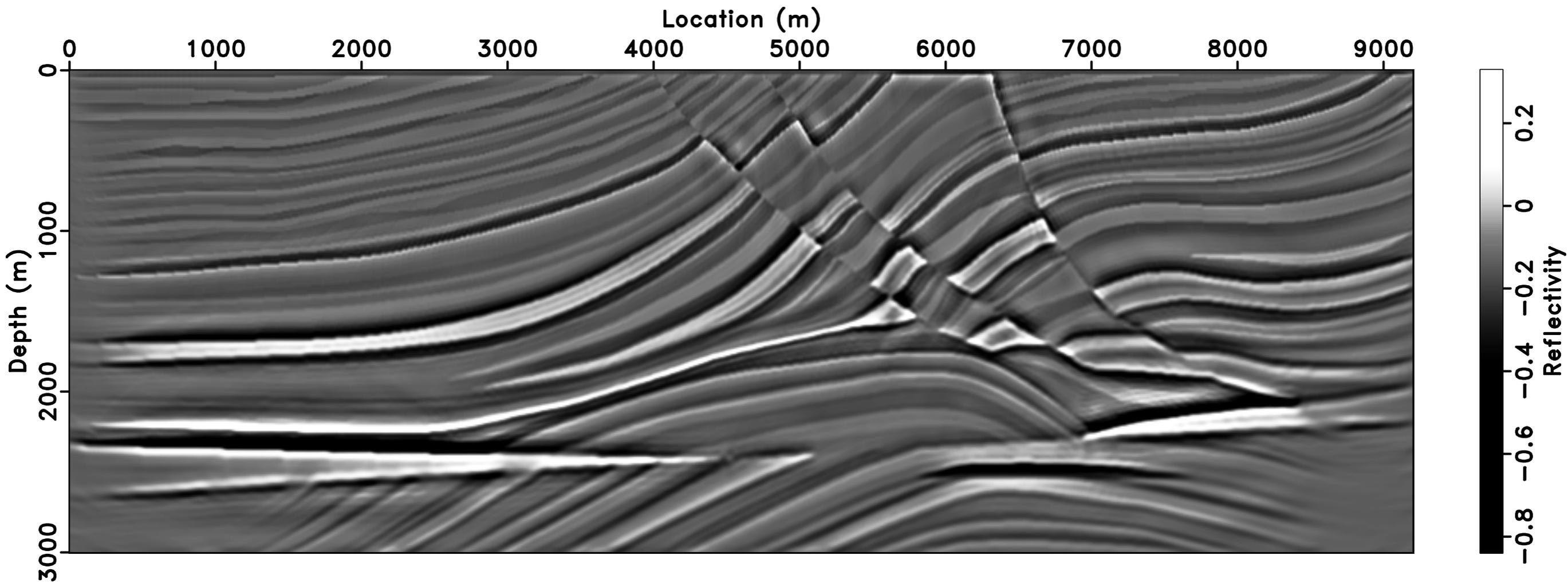
Reflectivity Model

# Approximate Inverse Operator



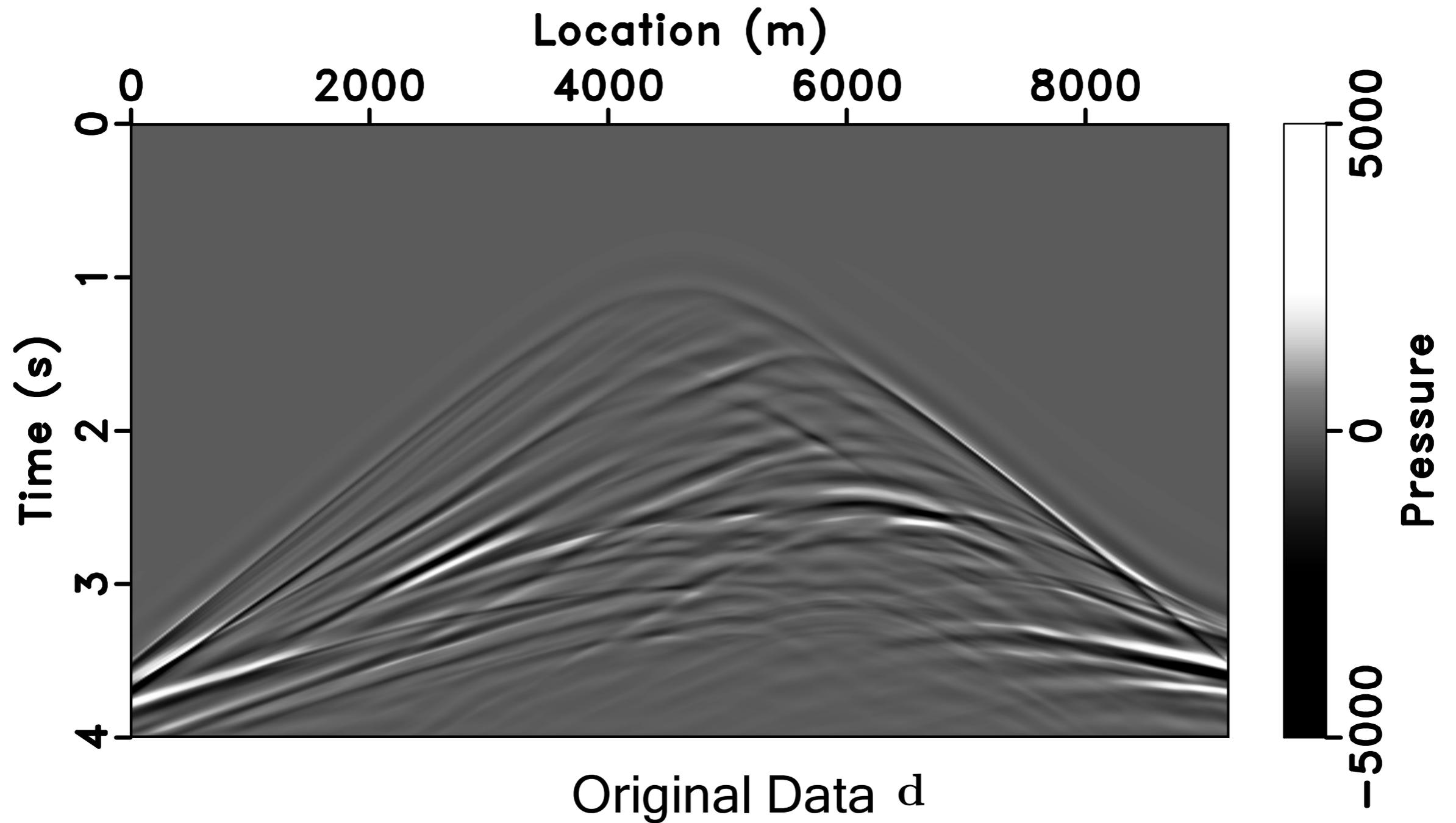
Inverted Model  $\bar{F}^\dagger d$

# Approximate Inverse Operator

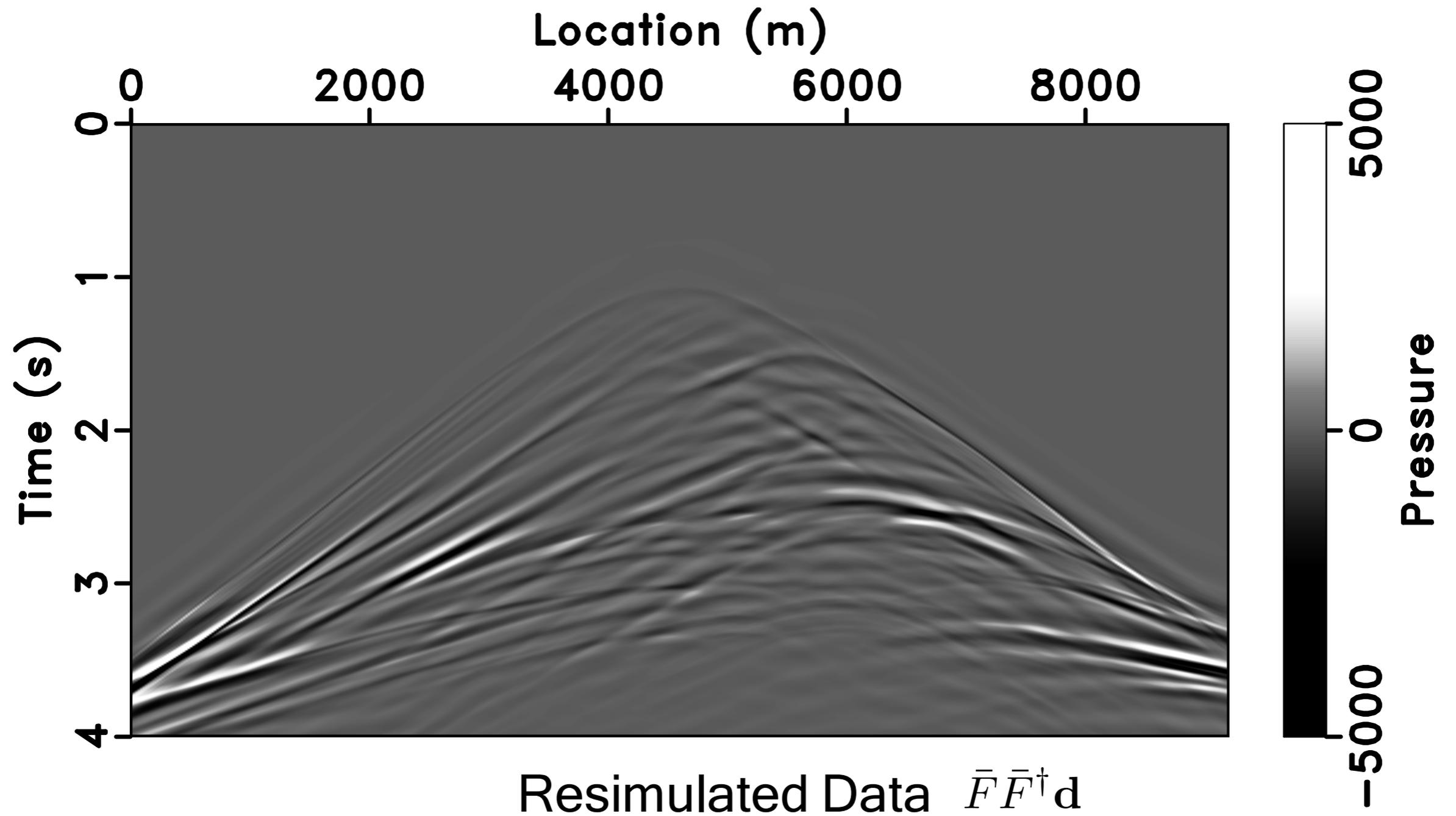


Stacked Model  $\sum_h I(\mathbf{x}, h)$

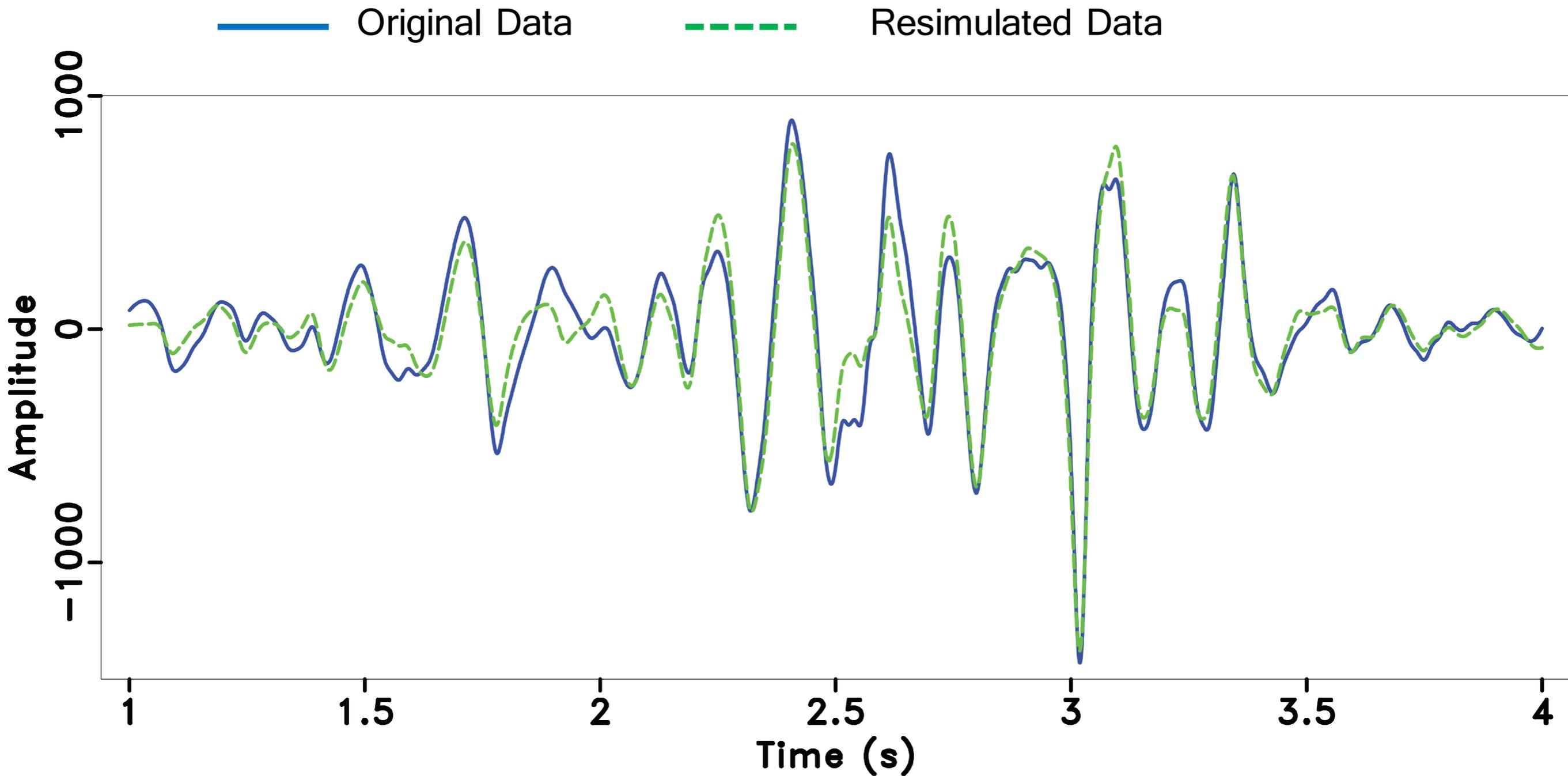
# Approximate Inverse Operator



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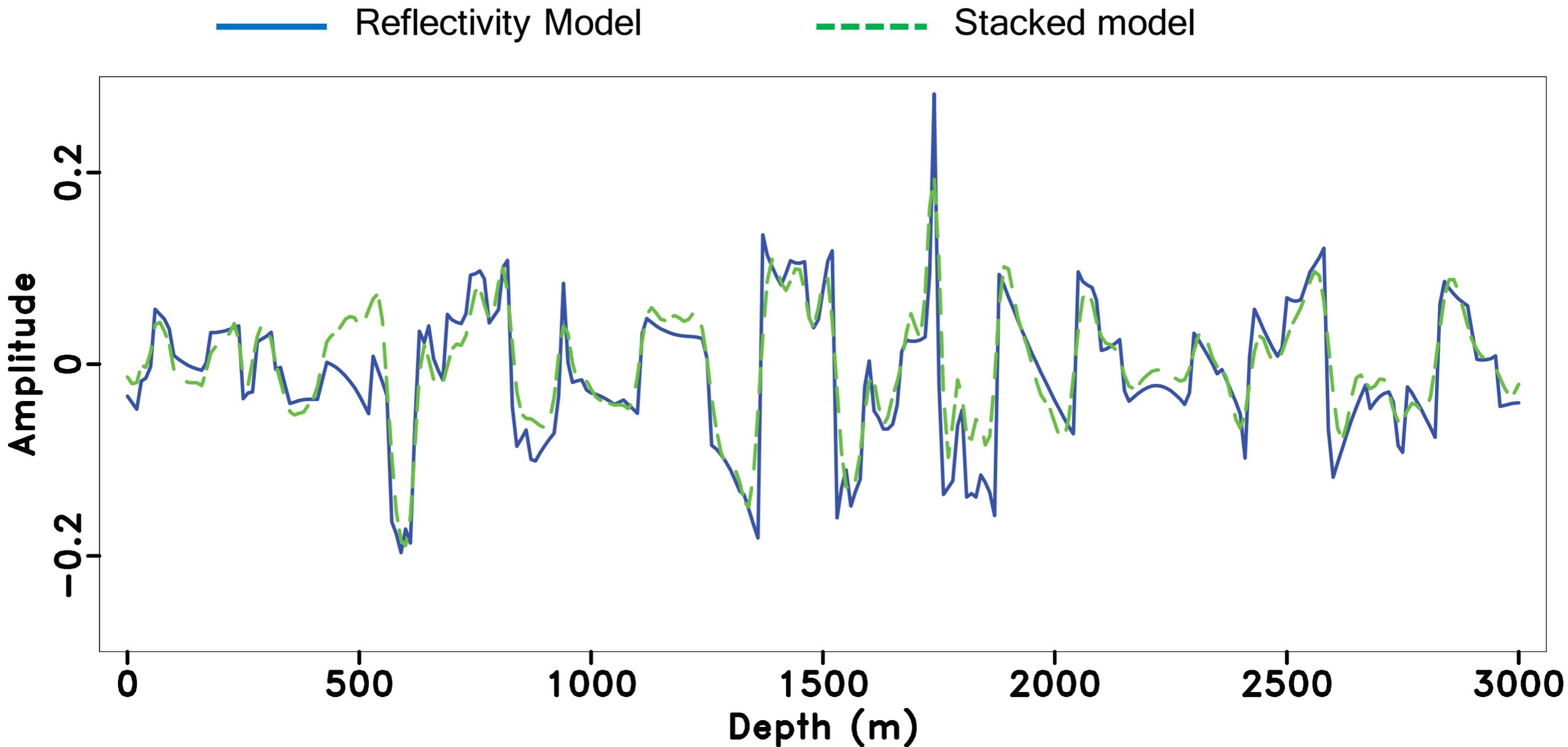


# Approximate Inverse Operator



Single **Data** Trace Comparison

# Approximate Inverse Operator



Single Image Trace Comparison

# Imaging Operators

## ➤ Conventional RTM Operator

$$I(\mathbf{x}, h) = \int d\mathbf{x}_S d\mathbf{x}_R dt d\tau G(\mathbf{x}_S, \mathbf{x} - h, \tau) G(\mathbf{x} + h, \mathbf{x}_R, t - \tau) d(\mathbf{x}_S, \mathbf{x}_R, t)$$

## ➤ Adjoint Operator

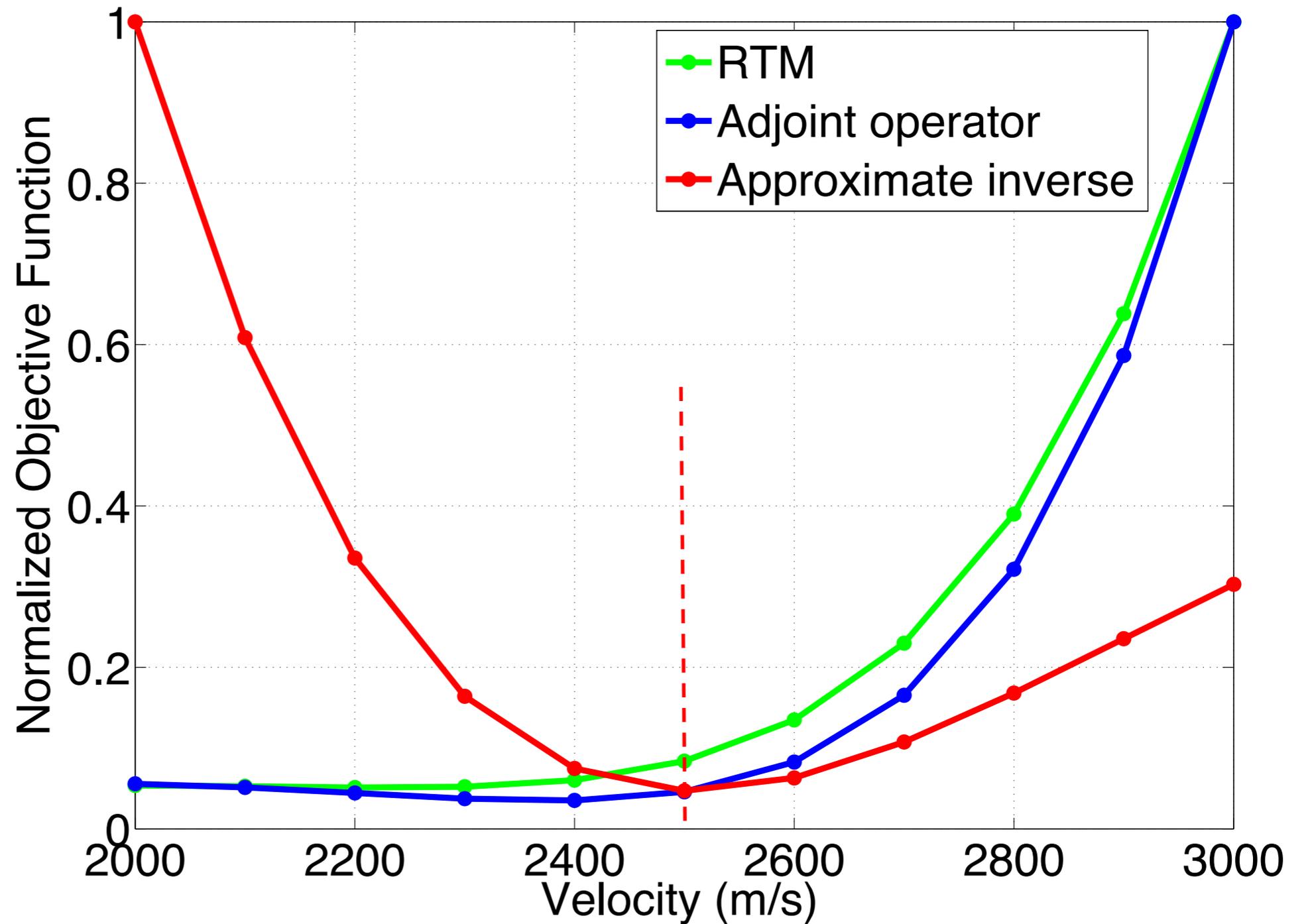
$$I(\mathbf{x}, h) = \frac{2}{v_0(\mathbf{x})^3} \int d\mathbf{x}_S d\mathbf{x}_R dt d\tau G(\mathbf{x}_S, \mathbf{x} - h, \tau) G(\mathbf{x} + h, \mathbf{x}_R, t - \tau) \frac{\partial^2}{\partial t^2} d(\mathbf{x}_S, \mathbf{x}_R, t)$$

## ➤ Approximate Inverse Operator

$$\bar{F}^\dagger \simeq W_{model}^{-1} \bar{F}^T W_{data}$$

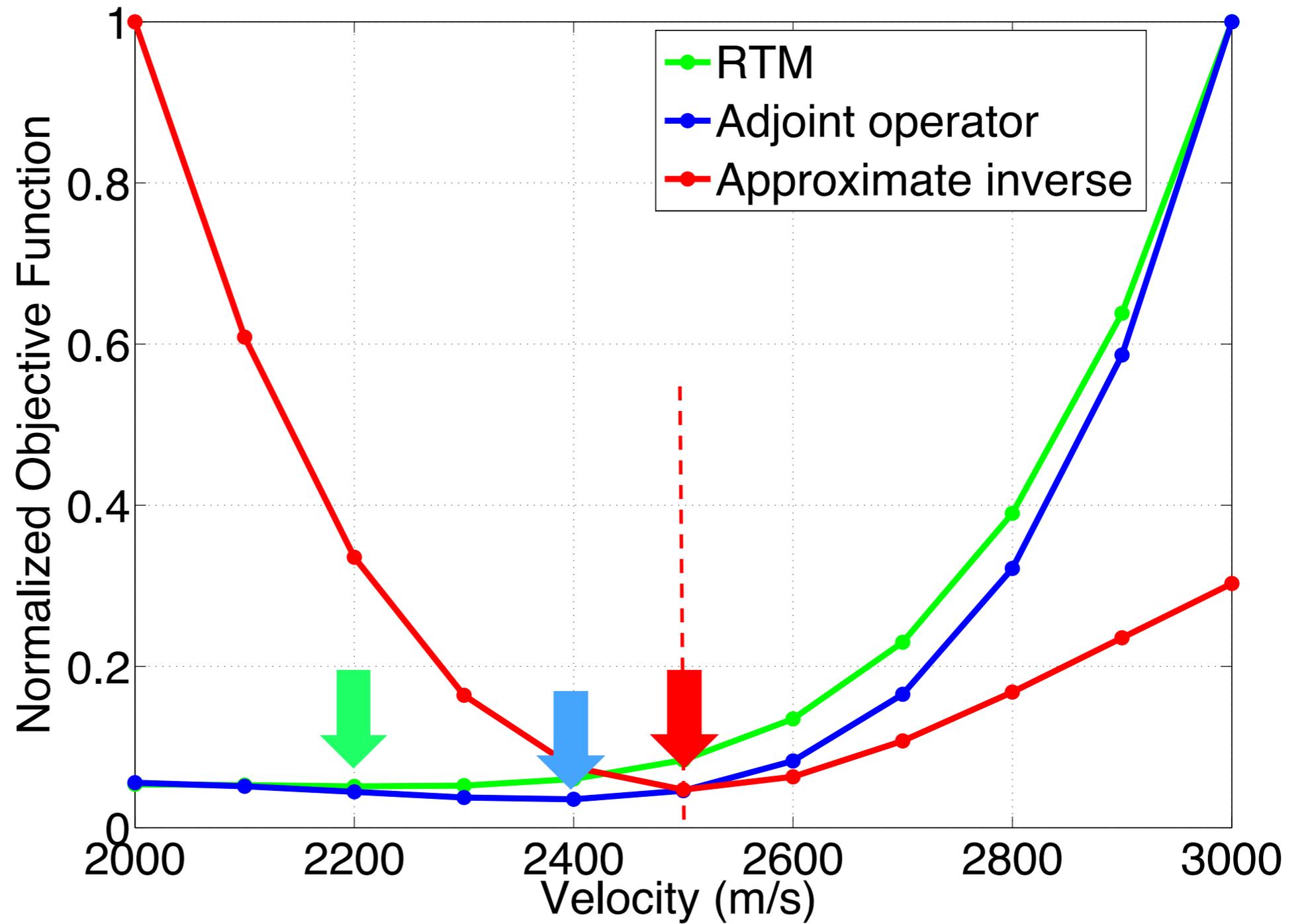
# Objective Function Behavior

2500m/s



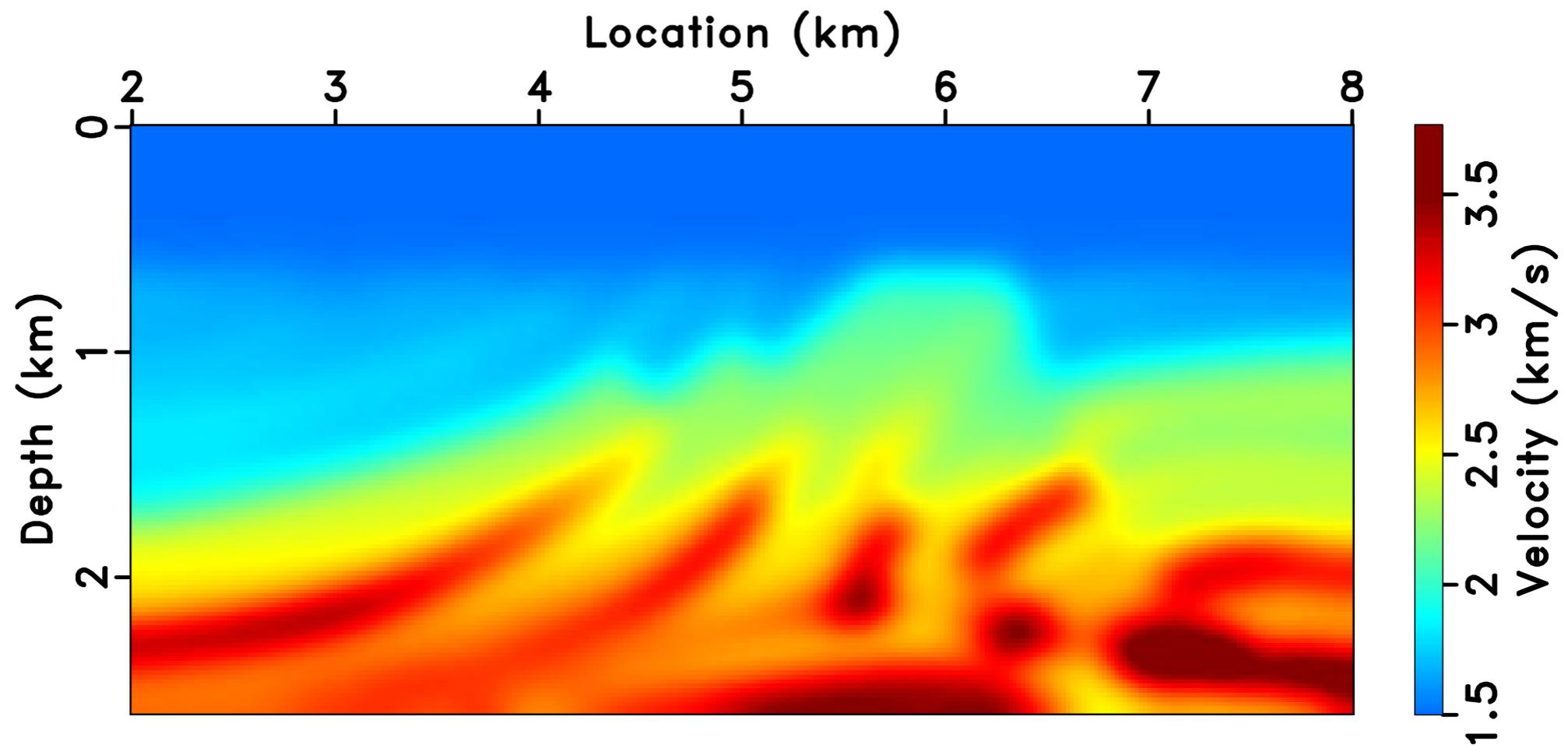
# Objective Function Behavior

2500m/s



# Numerical Experiment

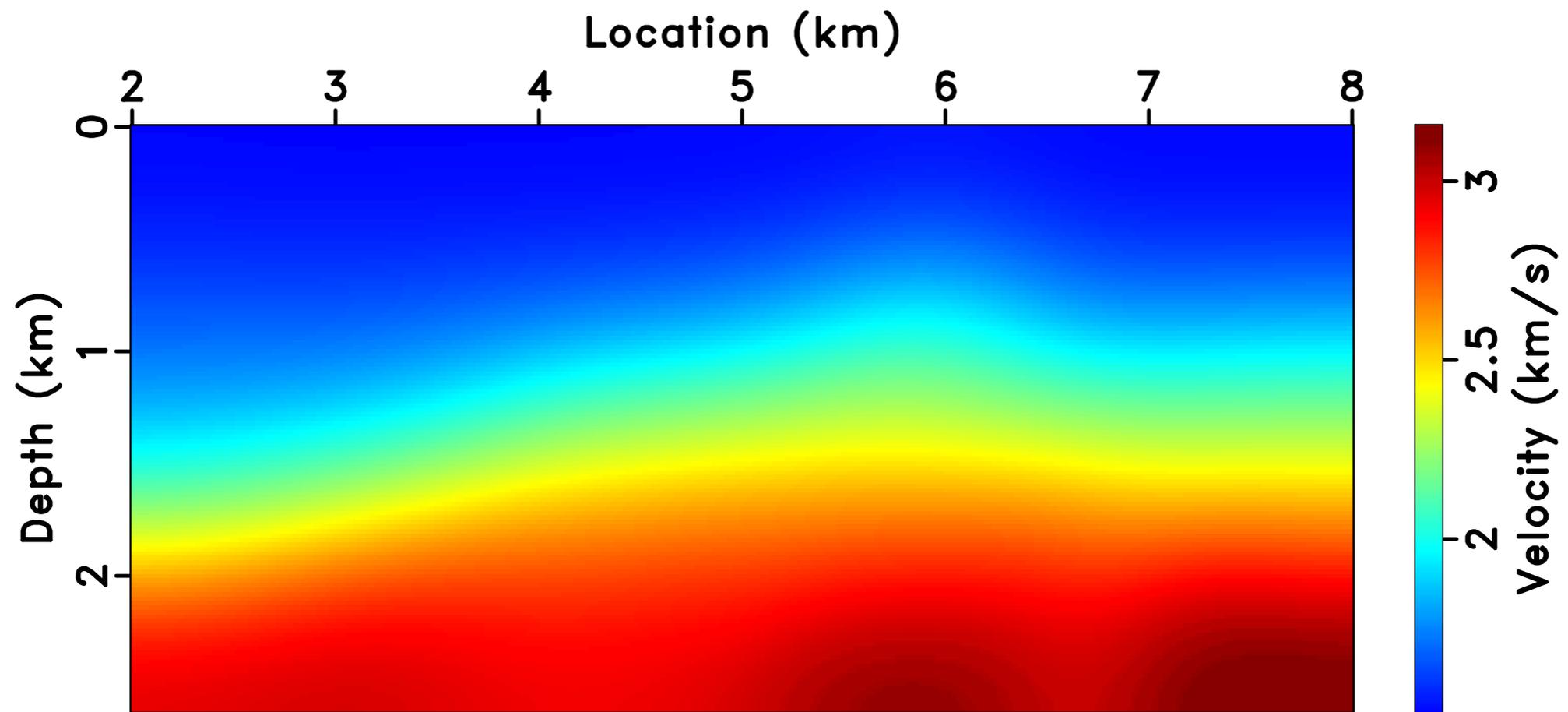
- 2-8 finite difference, 151 shots & 301 receivers
- 2.5-5-20-25Hz Bandpass wavelet
- 2ms time sample, 20m grid interval



True Model

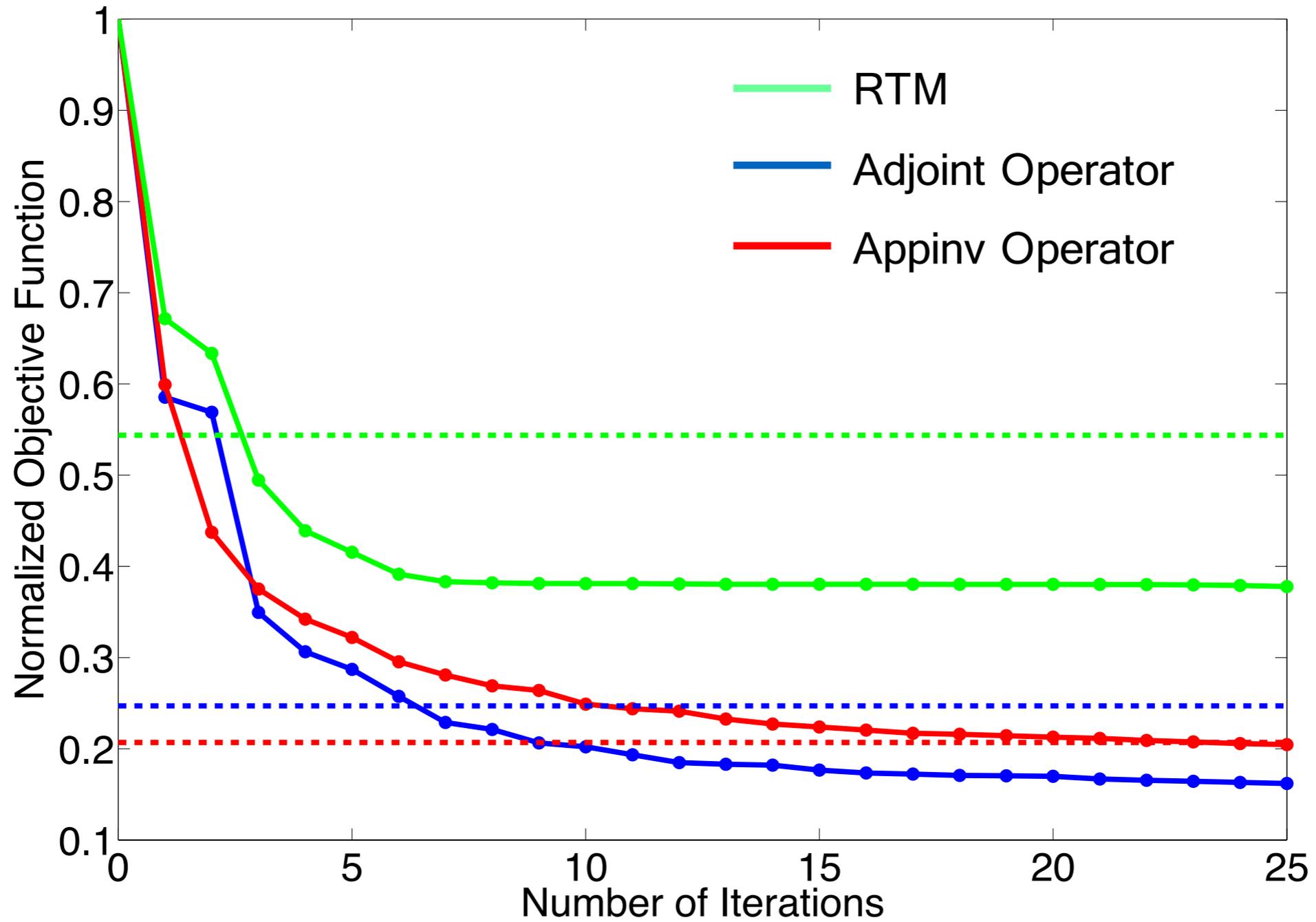
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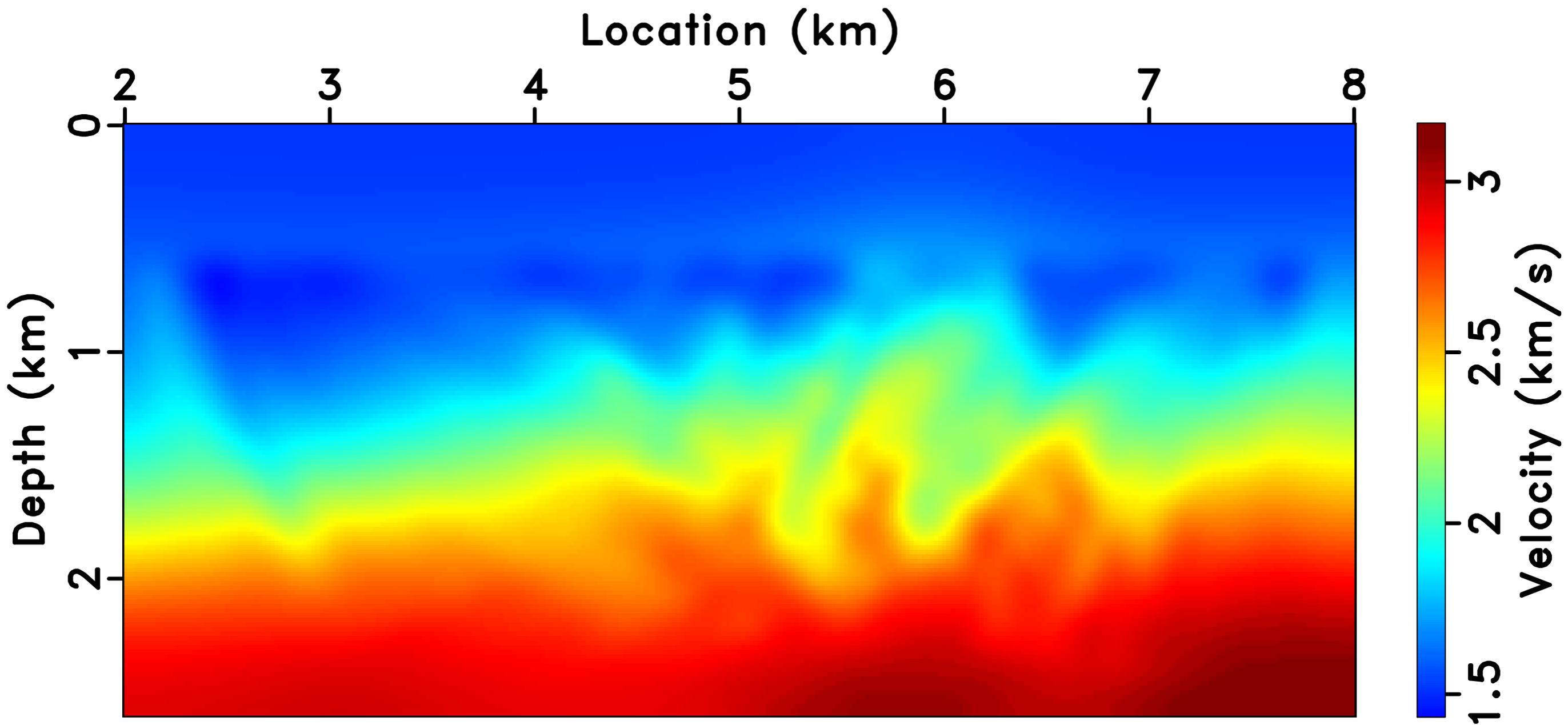


Initial Model

# Numerical Experiment

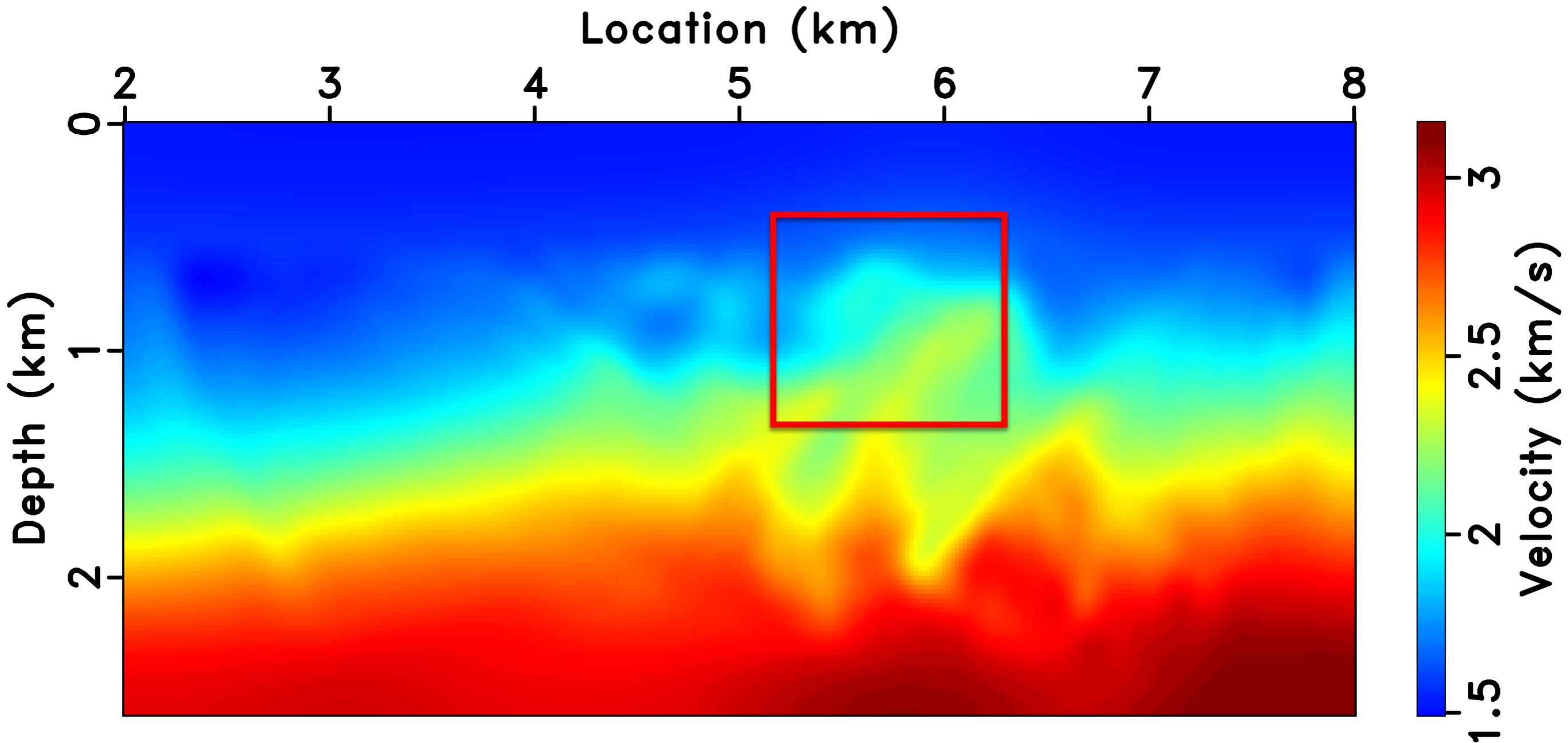


# Numerical Experiment



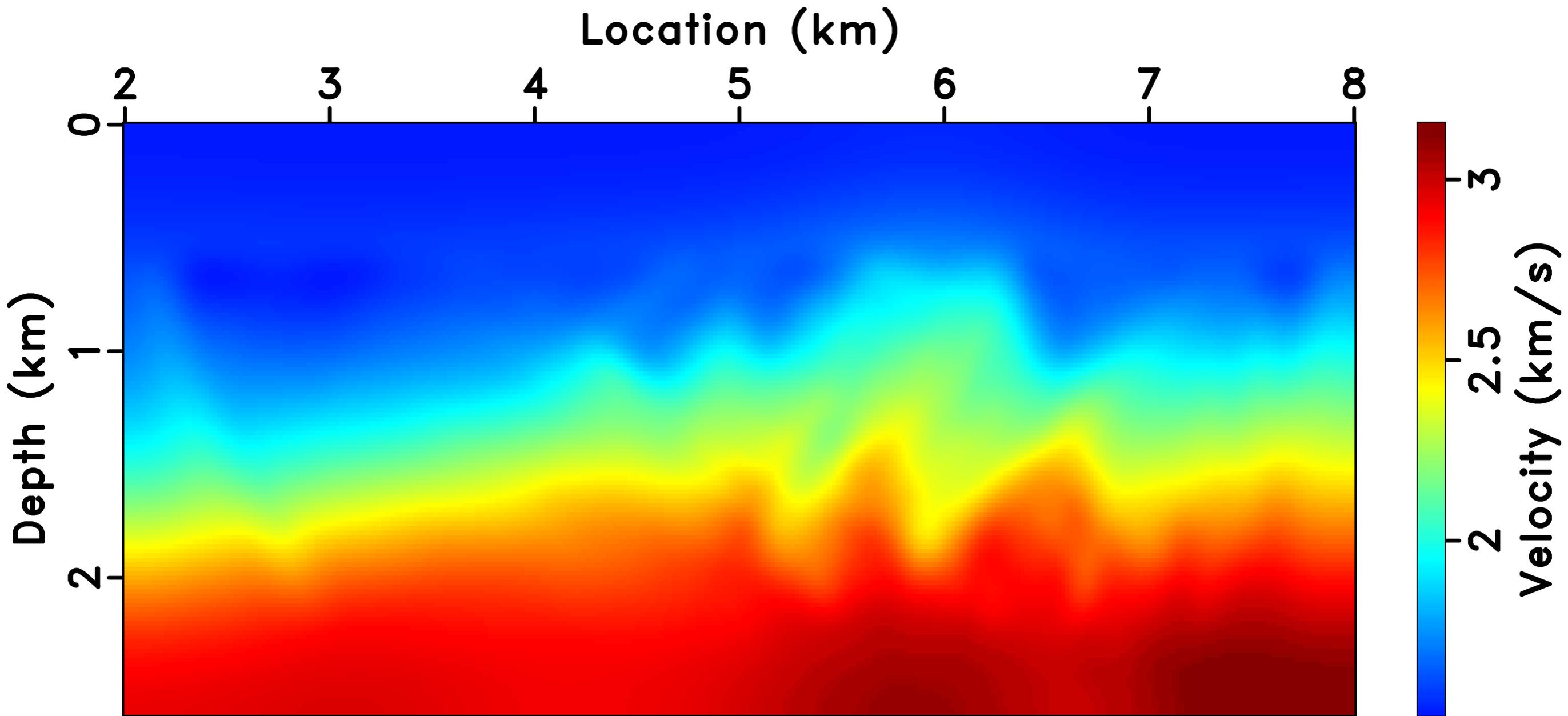
VA with Conventional RTM Operator

# Numerical Experiment



VA with Adjoint Operator

# Numerical Experiment



VA with Appinv. Operator

# Numerical Experiment

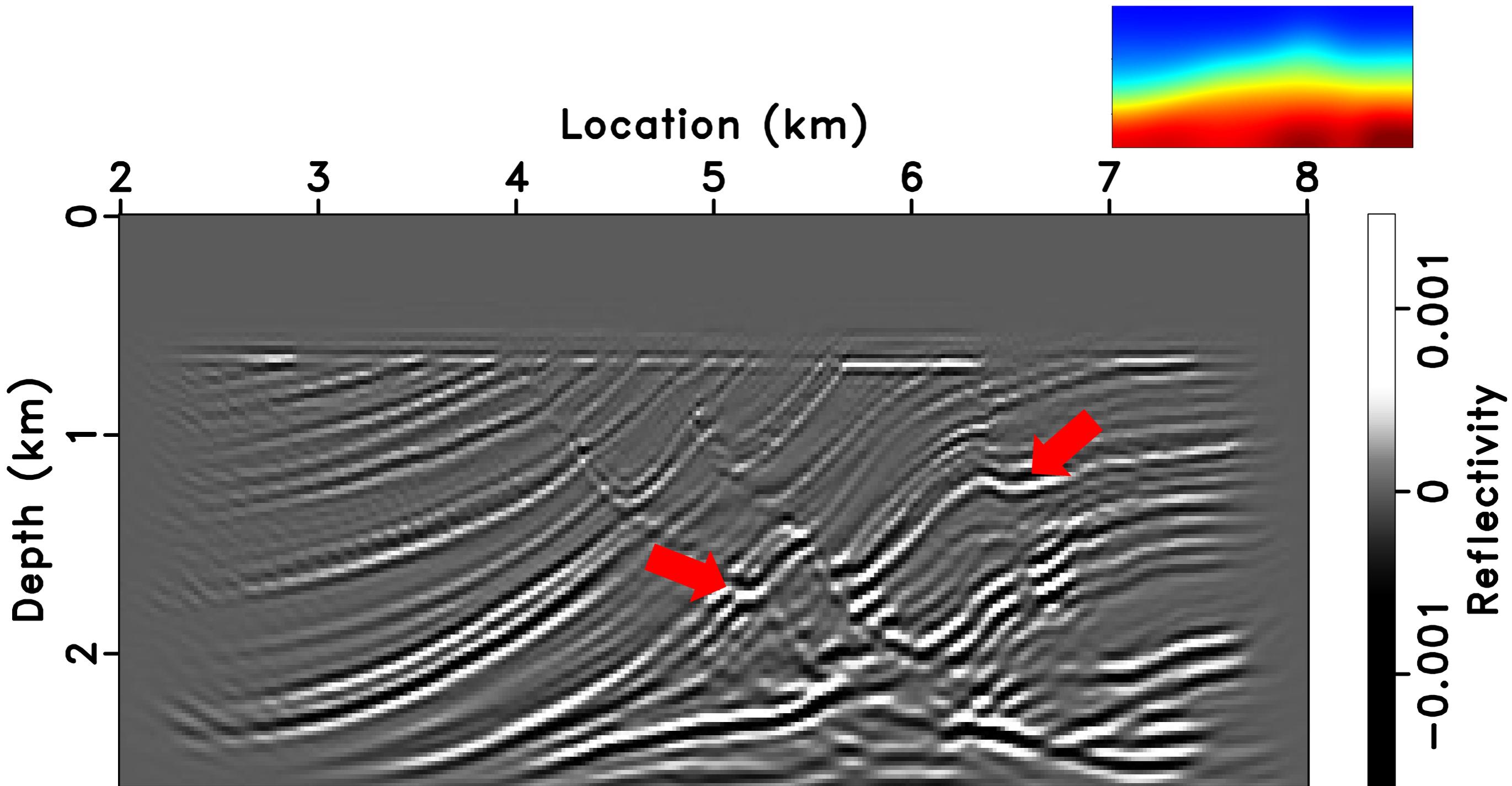


Image with Initial Model

# Numerical Experiment

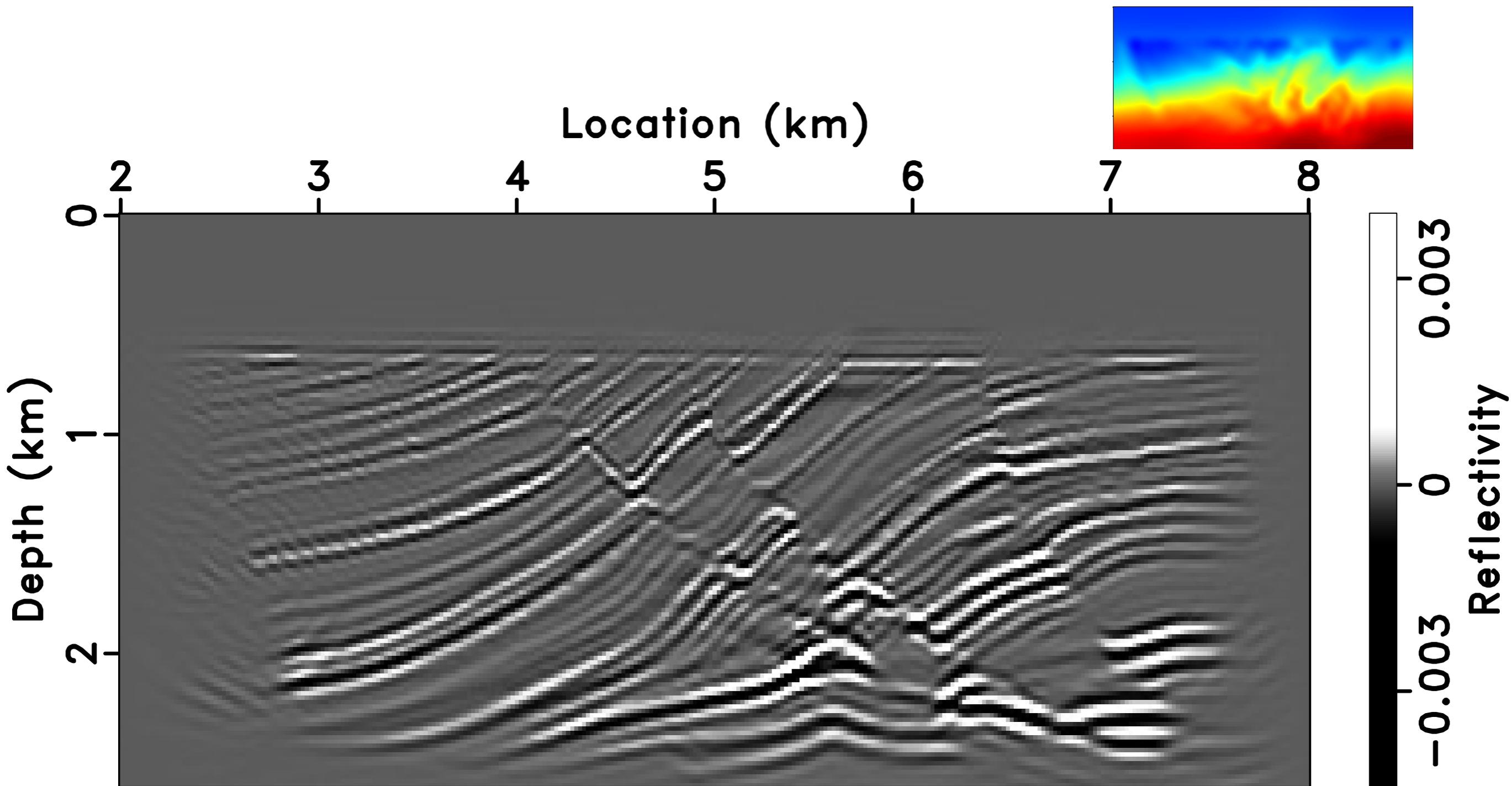


Image with Recovered Model (Conventional RTM)

# Numerical Experiment

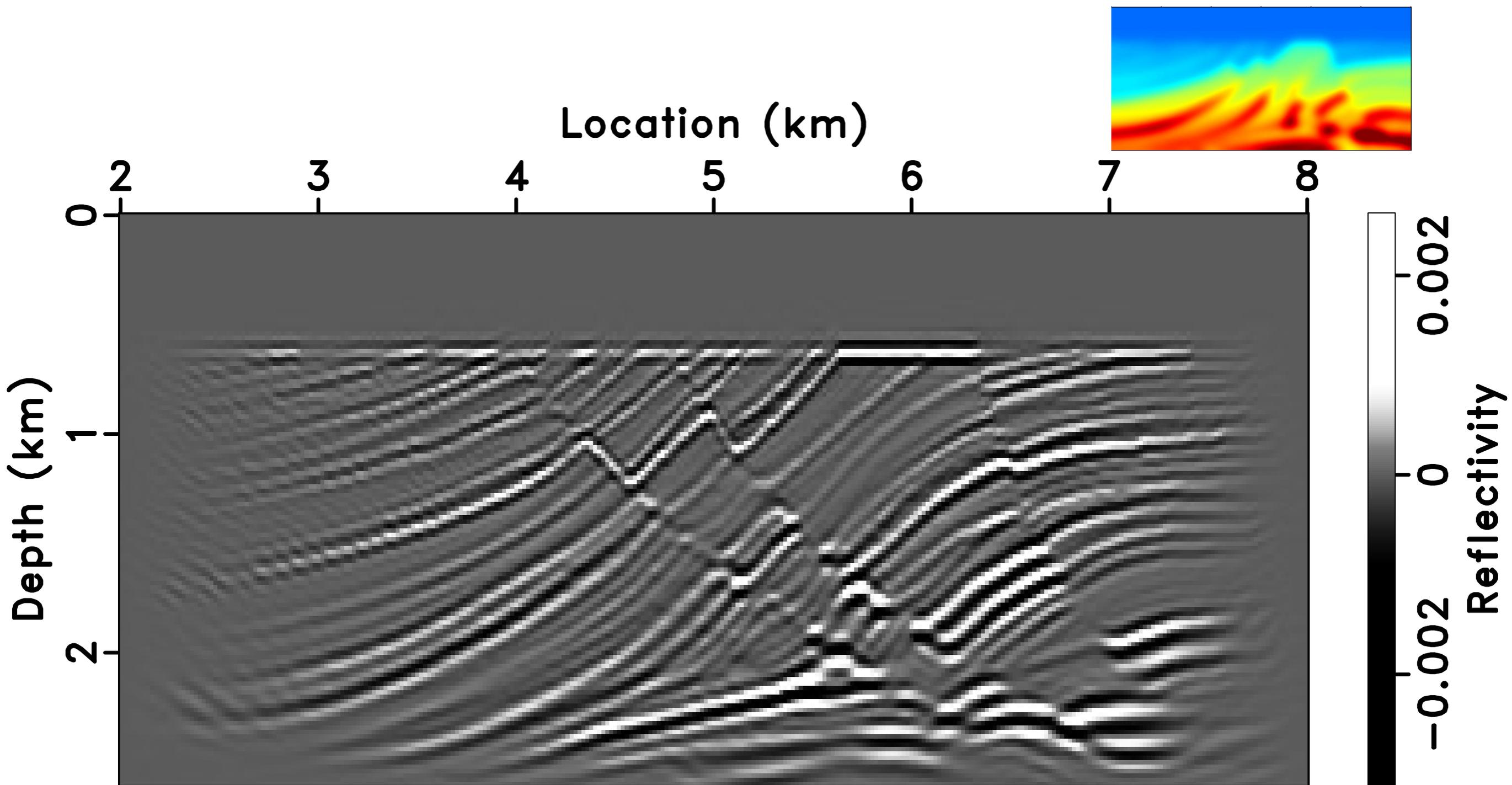


Image with **True Model**

# Numerical Experiment

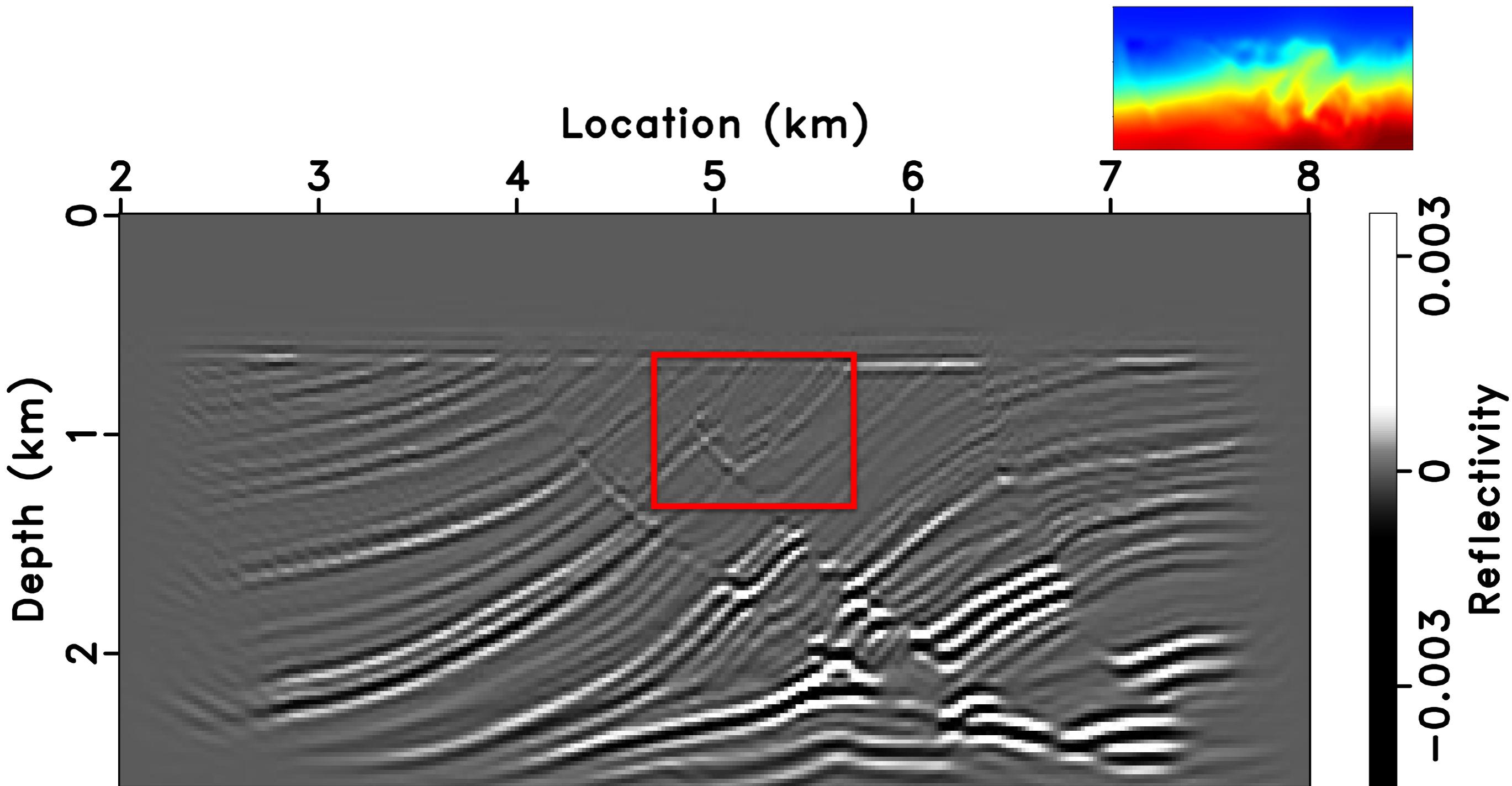


Image with Recovered Model ([Adjoint Operator](#))

# Numerical Experiment

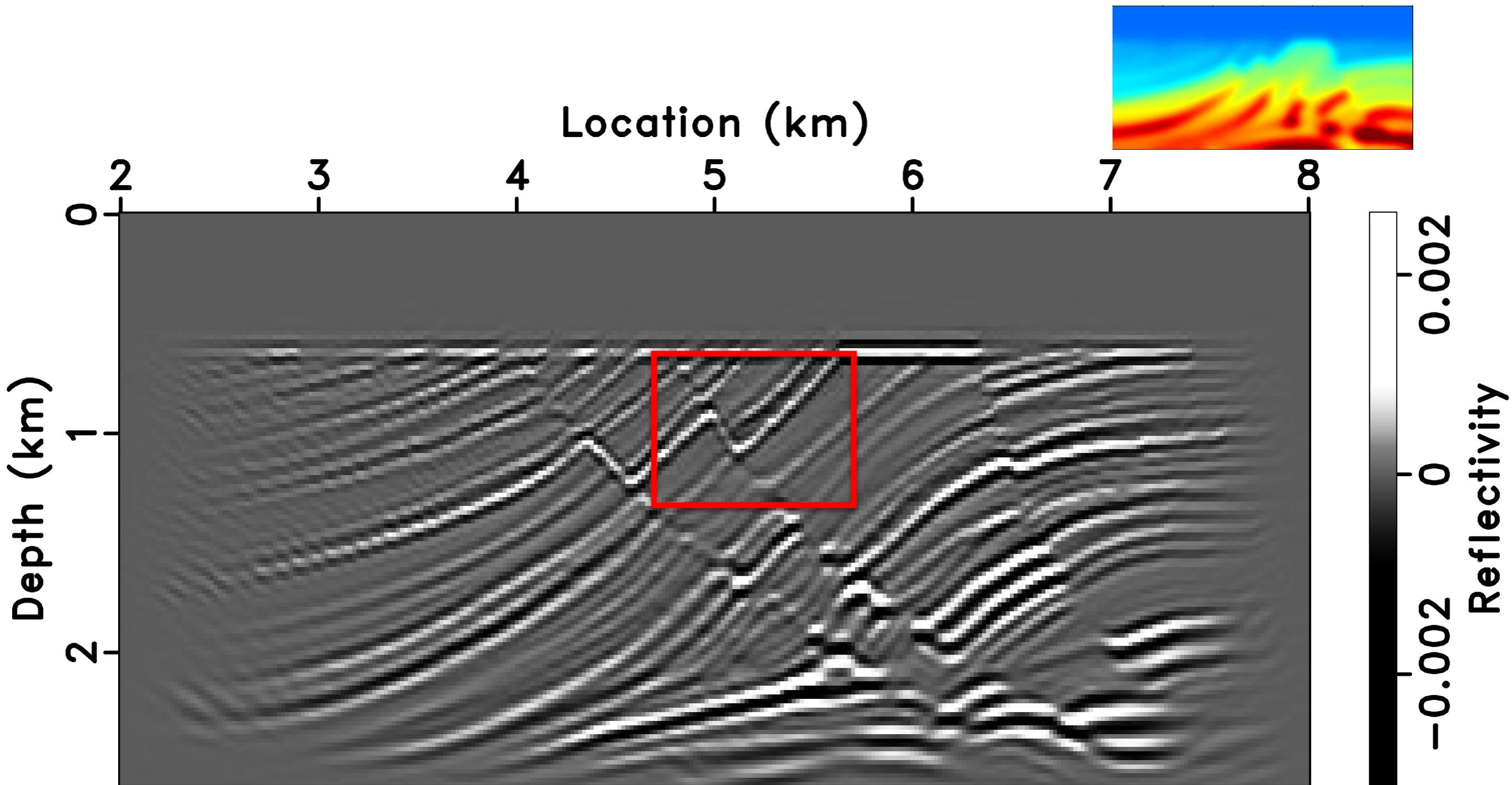


Image with **True Model**

# Numerical Experiment

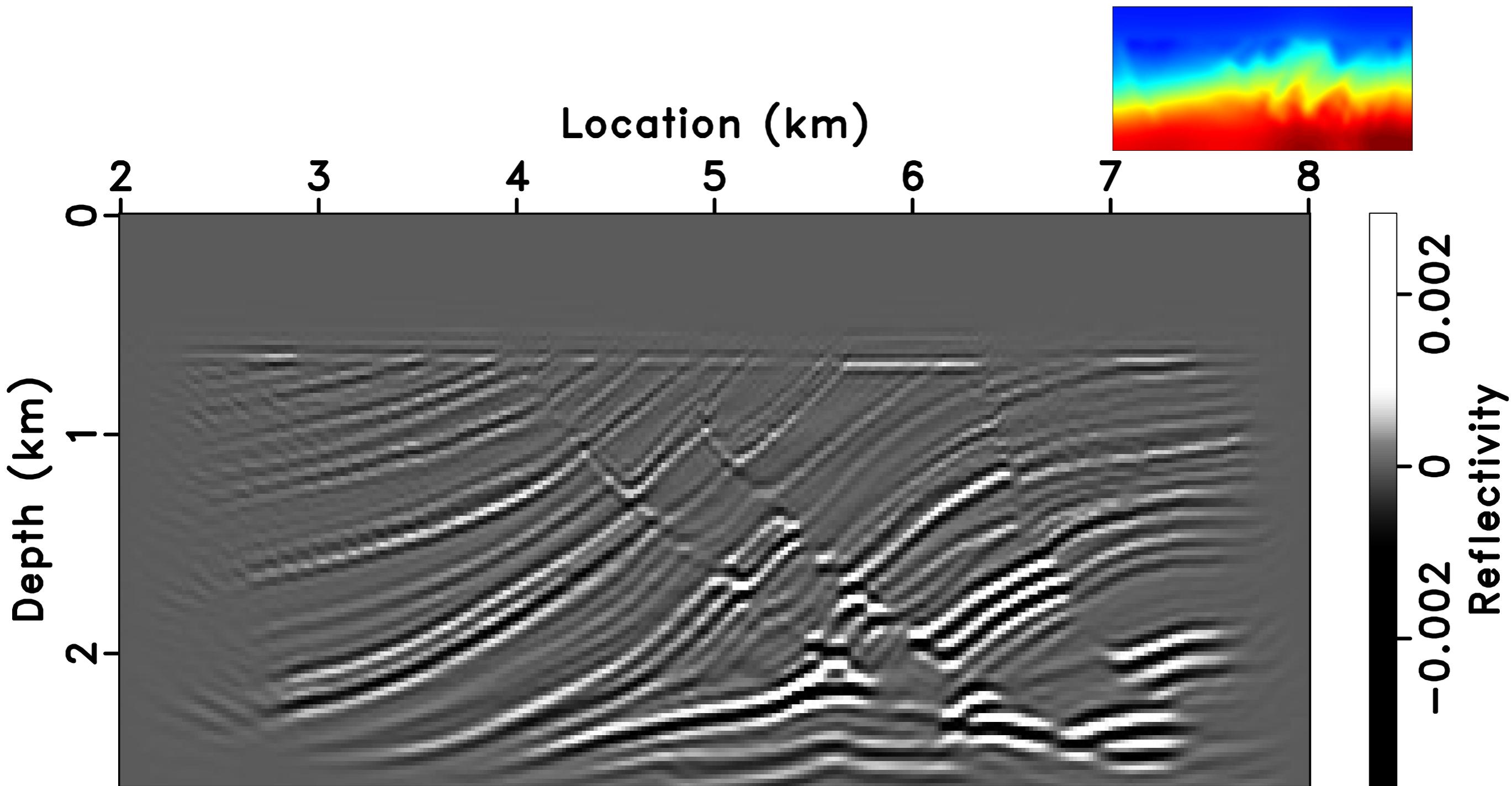


Image with Recovered Model ([Appinv](#))

# Numerical Experiment

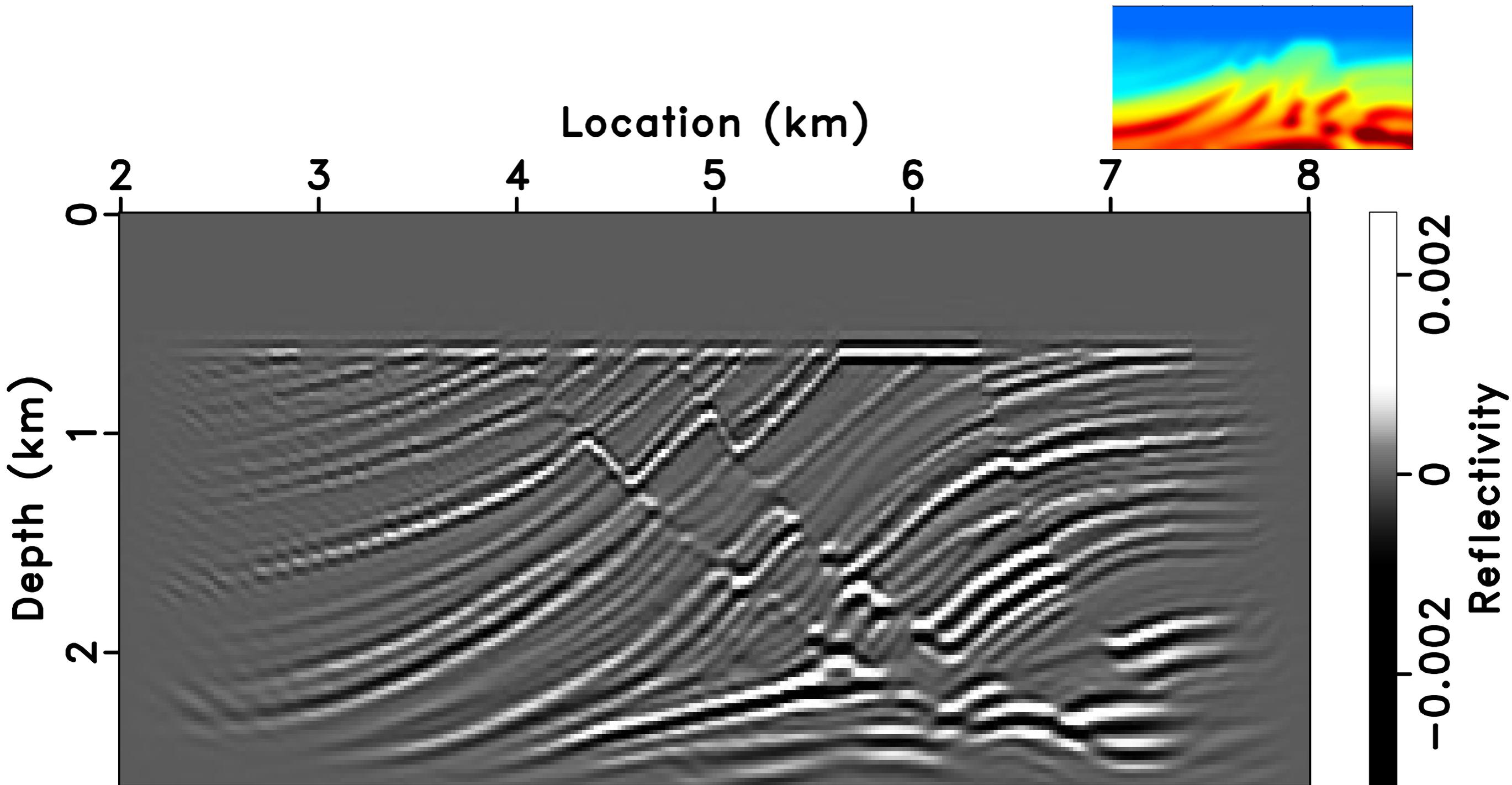
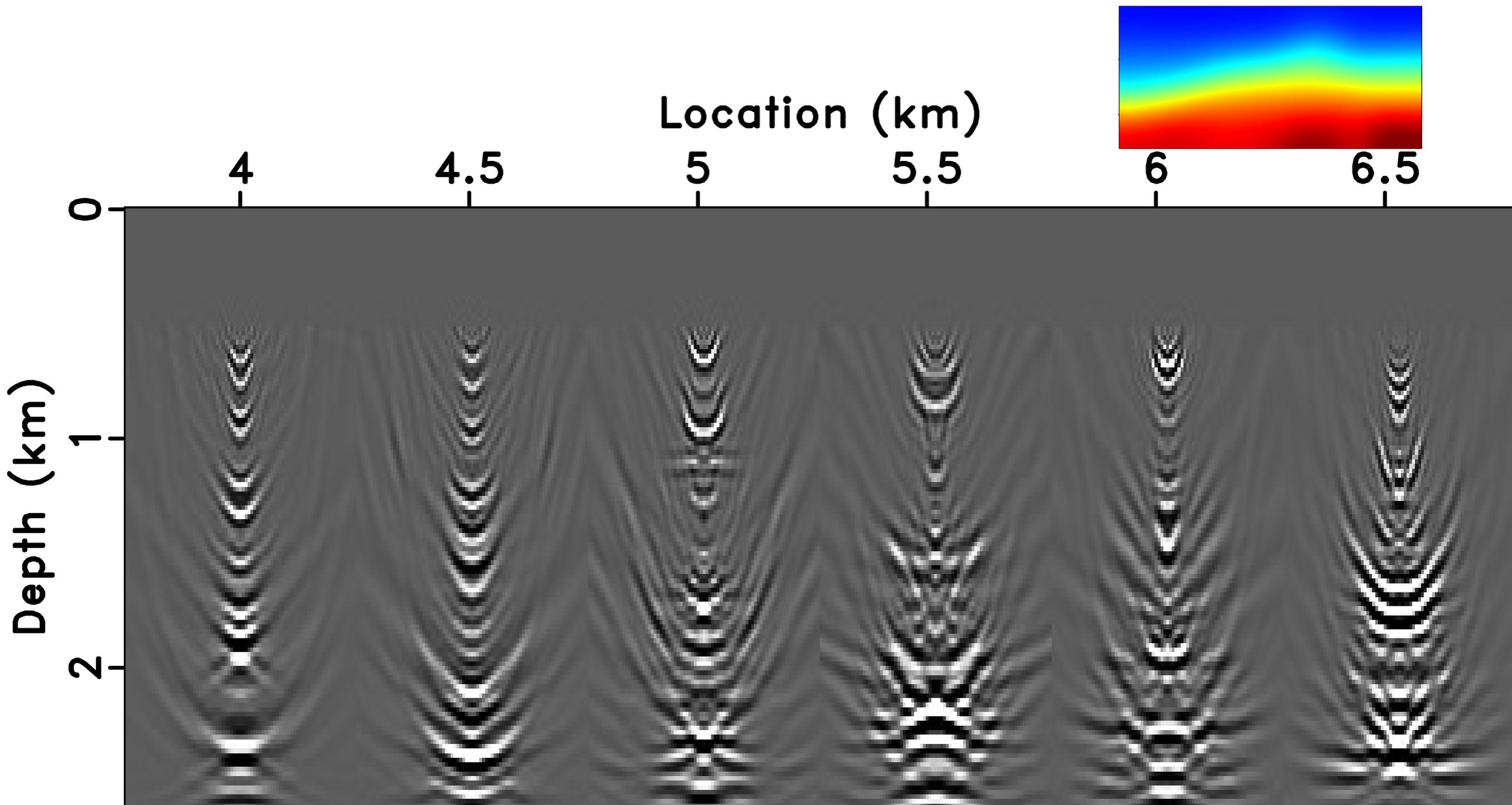


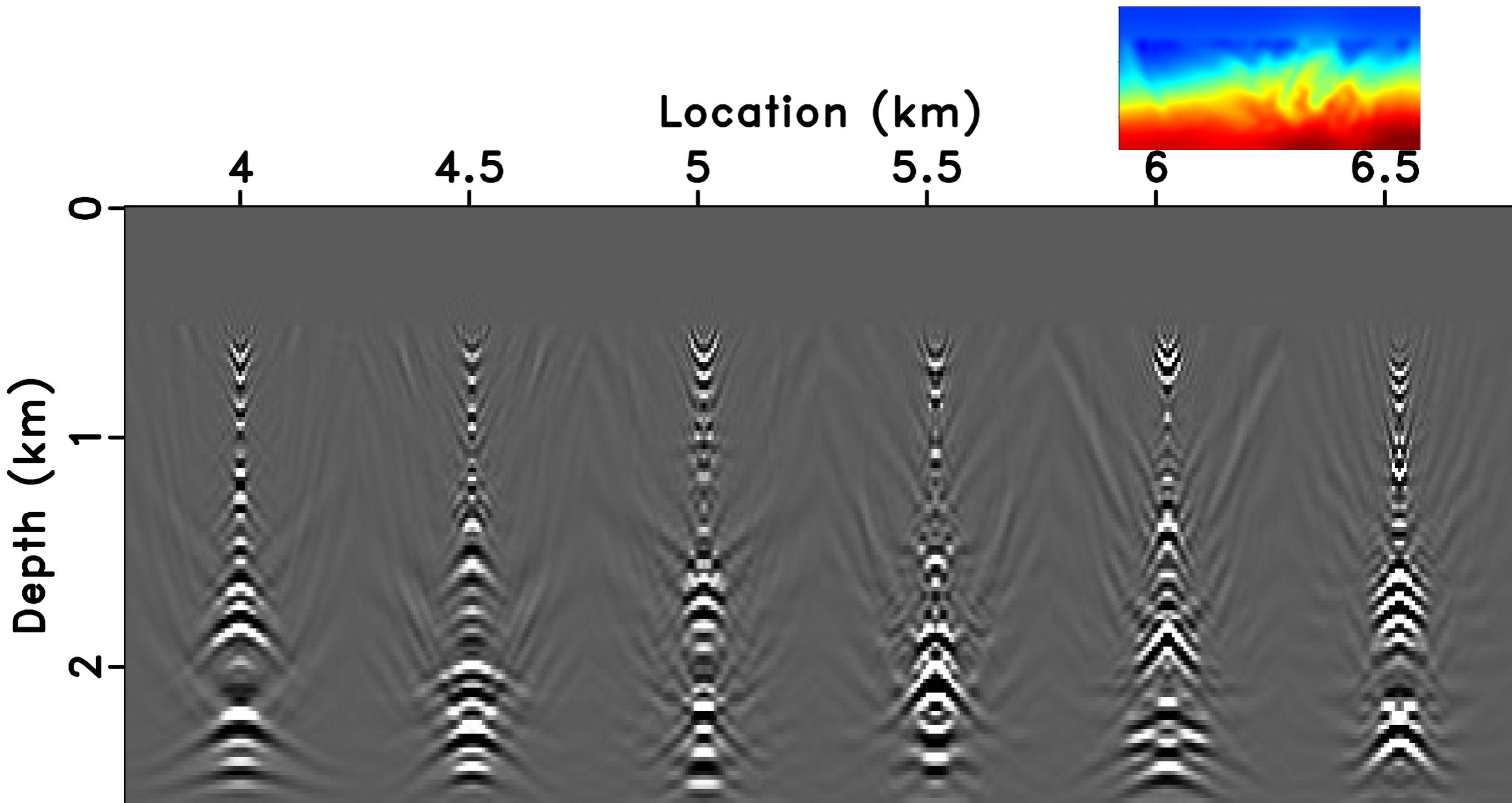
Image with **True Model**

# Numerical Experiment



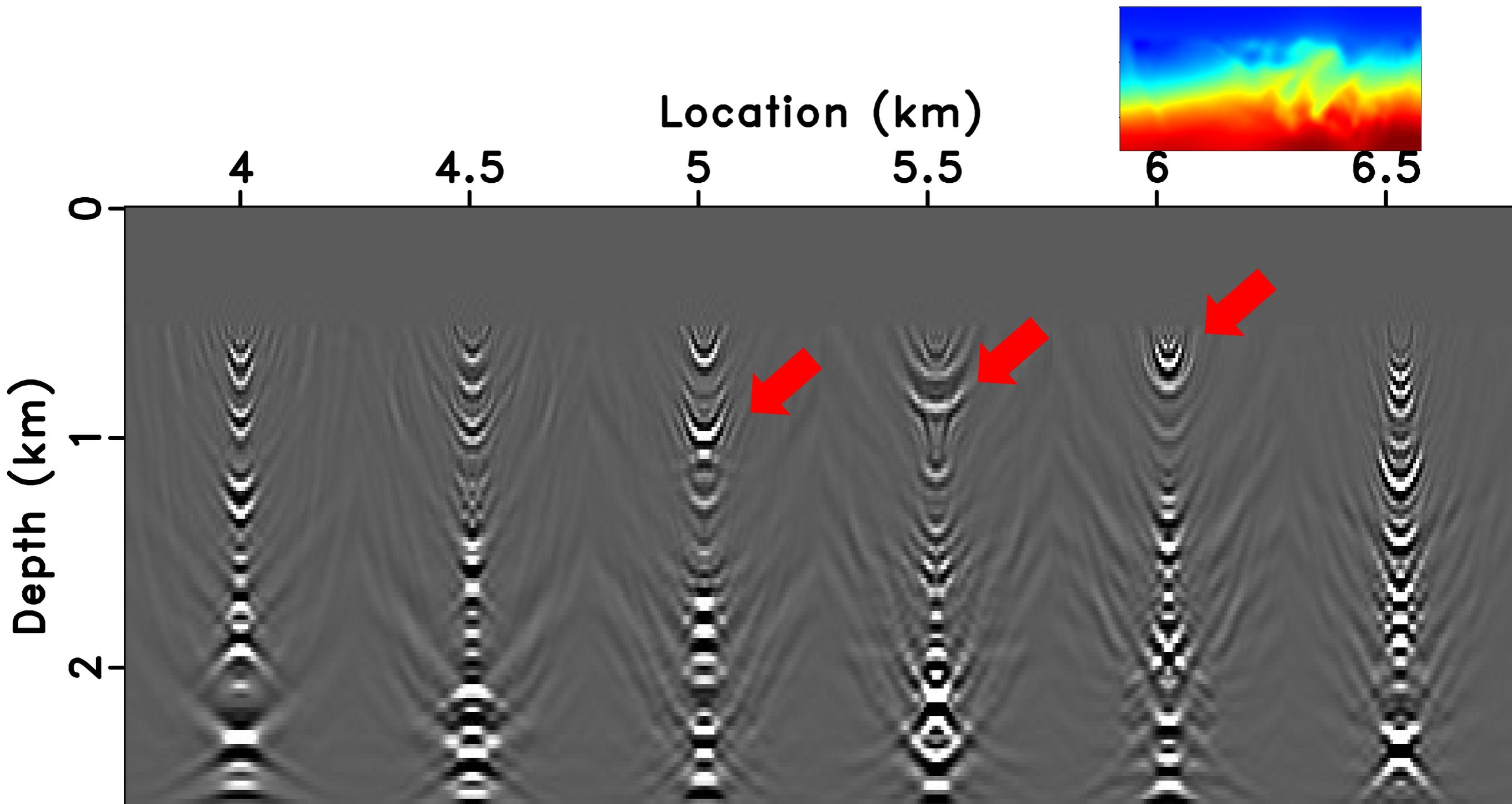
CIG with Initial Model

# Numerical Experiment



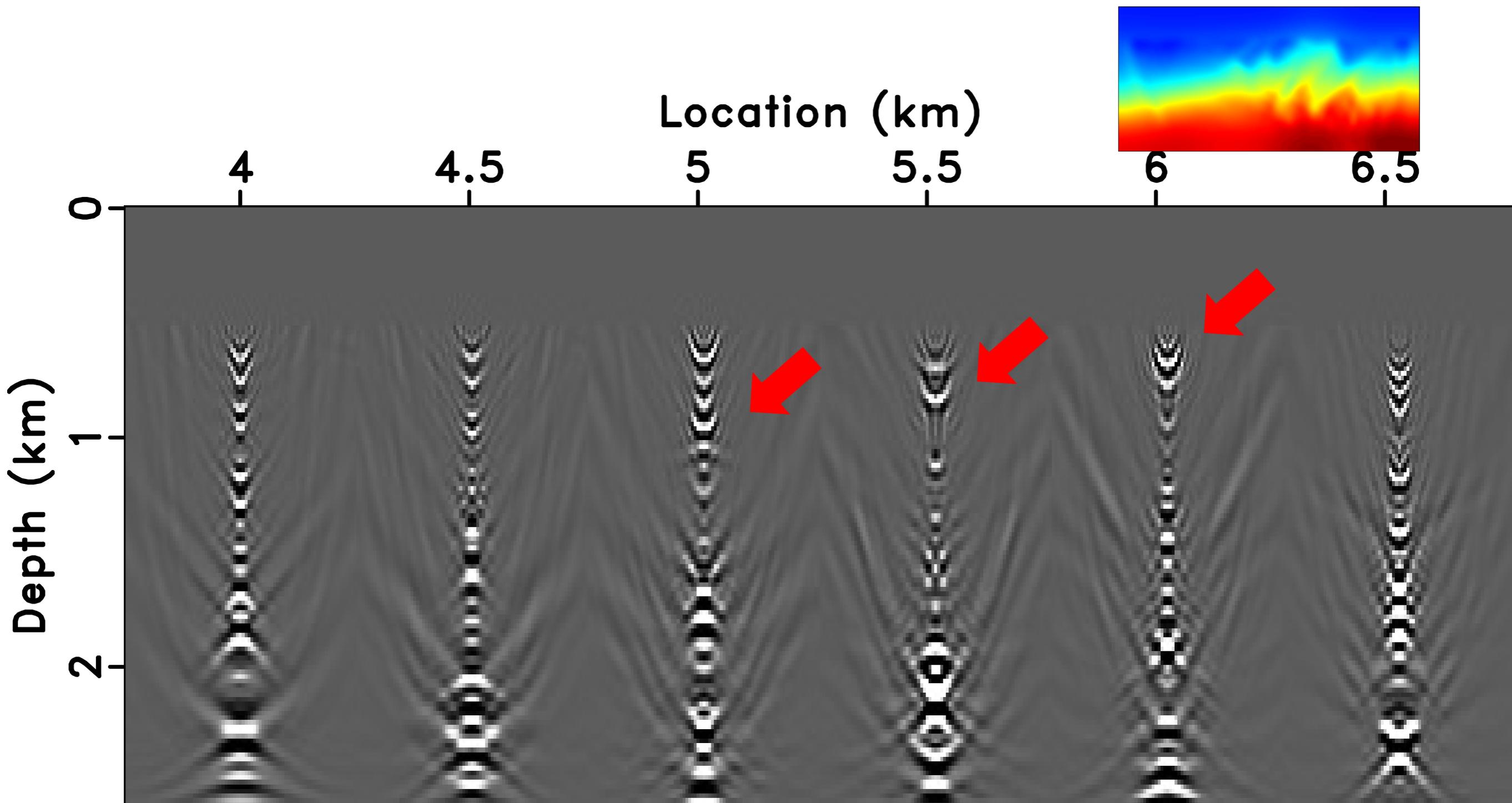
CIG with Recovered Model (RTM)

# Numerical Experiment



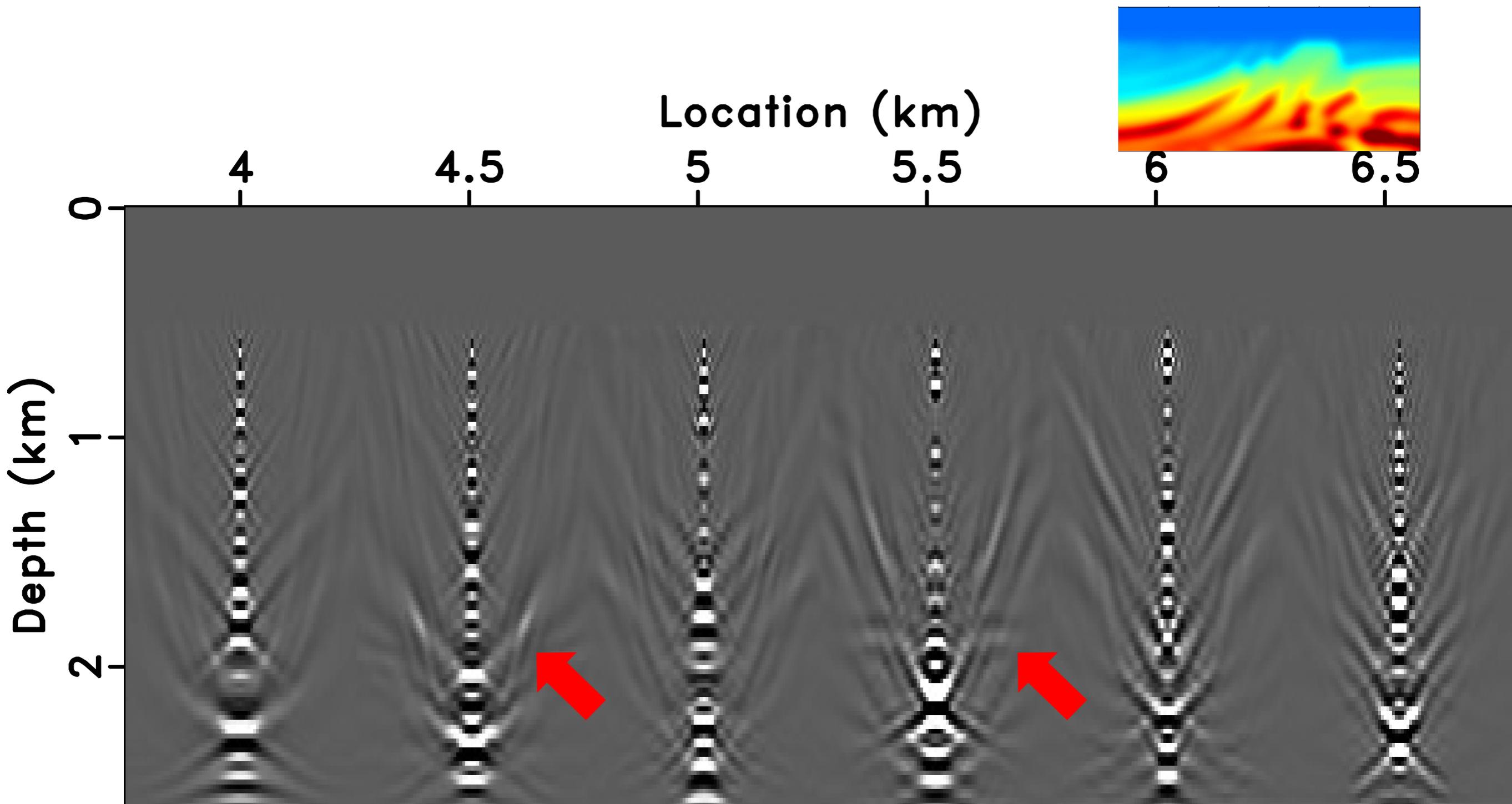
CIG with Recovered Model (Adjoint)

# Numerical Experiment



CIG with Recovered Model (Appinv)

# Numerical Experiment



CIG with True Model

# Conclusion

- MVA complements FWI by extracting **long-scale** information
- “Gradient Artifacts” are features of the objective function, **not gradient**
- Approximate inverse operator improve the performance of velocity analysis

# Acknowledgement

- Fons ten Kroode for inspiring our work
- Jon Sheiman, Henning Kuehl, Peng Shen
- Shell International Exploration & Production
- TRIP members and sponsors
- TACC and RCSG for HPC resources
- **Thank you for listening**

*Thank  
You*