

Asymptotic Inversion of the Variable Density Acoustic Model



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SEG, Anaheim 2018

Outline

- Introduction: Born Modeling \ Inversion
- Approximate Born Inversion:
 1. Constant Density.
 2. Variable Density.
- Numerical Examples
- Concluding Remarks

Seismic Inversion

Relate the model space M to the data space D through the forward map F

$$\mathcal{F}: M \rightarrow D$$

Find the model $m \in M$ that fits best the observed datum $d \in D$.

Minimization of the L_2 misfit function:

$$J[m, d] = \frac{1}{2} \|\mathcal{F}[m] - d\|^2$$

Full Waveform Inversion (FWI).

Extends into any modeling physics, or data geometry.

Nonlinear, ill-posed problem.

Born Scattering Theory

- **Scalar wave equation (**constant density**):**

$$\left(\frac{1}{v^2} \frac{\partial^2}{\partial t^2} - \nabla^2 \right) p(t, \mathbf{x}; \mathbf{x}_s) = f(t, \mathbf{x}; \mathbf{x}_s)$$

$$\mathcal{F}[v](t, \mathbf{x}_r; \mathbf{x}_s) = p(t, \mathbf{x}_r; \mathbf{x}_s)$$

- **Linearization:**

$$\begin{aligned} v &= v_0 + \delta v \\ p &= p_0 + \delta p \end{aligned}$$

$$\mathcal{F}[v_0] \cong \mathcal{F}[v_0] + F[v_0] \delta v$$

- **Born-type wave equation:**

$$\left(\frac{1}{v_0^2} \frac{\partial^2}{\partial t^2} - \nabla^2 \right) \delta p(t, \mathbf{x}; \mathbf{x}_s) = \frac{2\delta v}{v_0^3} \frac{\partial^2}{\partial t^2} p_0(t, \mathbf{x}; \mathbf{x}_s)$$

$$F[v_0] \delta v(t, \mathbf{x}_r; \mathbf{x}_s) = \delta p(t, \mathbf{x}_r; \mathbf{x}_s)$$

- **Linearized Misfit function:**

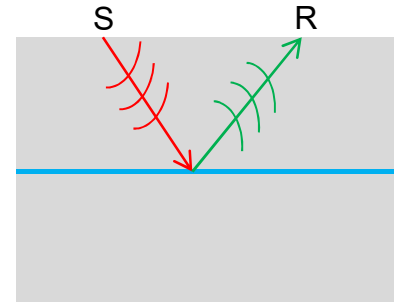
$$J_L[v_0, \delta v, \delta p] = \frac{1}{2} \| F[v_0] \delta v - (\delta p - \mathcal{F}[v_0]) \|^2$$

Born Scattering Theory

- Forward Modeling / Adjoint Imaging (**constant density**):

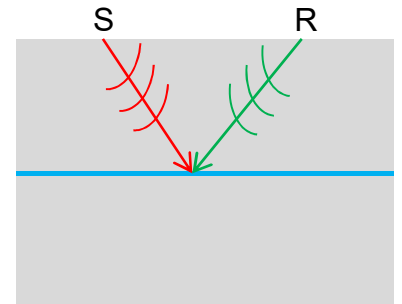
➤ Modeling:

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



➤ Imaging (migration):

$$(F^*[v_0]\delta p)(\mathbf{x}) = \frac{2}{v_0^3(\mathbf{x})} \int d\mathbf{x}_r \int d\mathbf{x}_s \int dt \frac{\partial^2}{\partial t^2} \underbrace{\delta p(t, \mathbf{x}_r; \mathbf{x}_s)}_{\text{blue}} \int \delta\tau \underbrace{G(t - \tau, \mathbf{x}; \mathbf{x}_r)}_{\text{green}} \underbrace{G(\tau, \mathbf{x}; \mathbf{x}_s)}_{\text{red}}$$



Model Extension

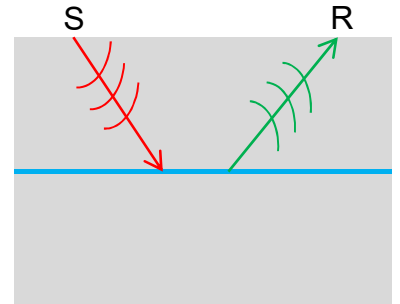
- Subsurface offset extension:

$$M \rightarrow \bar{M}$$

$$\mathcal{F}: \bar{M} \rightarrow D$$

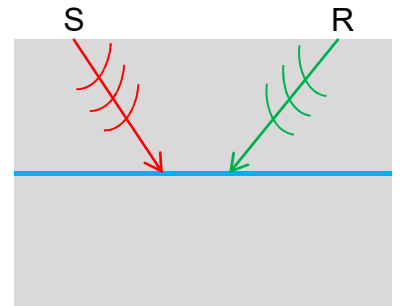
- Extended Modeling:

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



- Extended Imaging (migration):

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



Approximate Born Inversion

- **Approximate Inverse Operator (constant density):**

Modify migration (adjoint) into an inverse operator.

Introduced by **Hou and Symes (2017)** :

$$\bar{F}^\dagger(\mathbf{x}, \mathbf{h}) = W_m^{-1} \bar{F}^*(\mathbf{x}, \mathbf{h}) W_d$$

The diagram illustrates the decomposition of the inverse operator equation. A central equation box is connected by blue arrows to two sub-equations below it. The left arrow points to the model-side weight, and the right arrow points to the data-side weight.

model-side weight

$$W_m^{-1} = -8v_0^4 \partial_z$$

data-side weight

$$W_d = I_t(I_t \partial_{z_s})(I_t \partial_{z_r})$$

Variable Density Acoustics

- Dynamical Laws of Linear Acoustics:**

$$\left. \begin{array}{l} \text{Constitutive Law:} \\ \text{Momentum Balance:} \end{array} \right\} \begin{array}{l} \frac{1}{\kappa} \frac{\partial p}{\partial t} = -\nabla \cdot \mathbf{v} + f \\ \rho \frac{\partial \mathbf{v}}{\partial t} = -\nabla p \end{array}$$

- Linearization (extended):**

$$\begin{array}{ll} \kappa = \kappa_0 + \delta\bar{\kappa} & \mathbf{v} = \mathbf{v}_0 + \delta\bar{\mathbf{v}} \\ \rho = \rho_0 + \delta\bar{\rho} & p = p_0 + \delta\bar{p} \end{array}$$

- Born-type wave equation:**

$$\left\{ \begin{array}{l} \frac{1}{\kappa_0} \frac{\partial \delta\bar{p}}{\partial t} = -\nabla \cdot \delta\bar{\mathbf{v}} + \frac{1}{\kappa_0} \delta\bar{\kappa} \frac{1}{\kappa_0} \frac{\partial p_0}{\partial t} \\ \rho_0 \frac{\partial \delta\bar{\mathbf{v}}}{\partial t} = -\nabla \delta\bar{p} - \delta\bar{\rho} \frac{\partial \mathbf{v}_0}{\partial t} \end{array} \right.$$

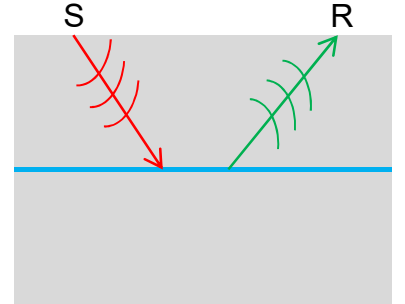
- Born forward map:**

$$(\bar{F}[\kappa_0, \rho_0][\delta\bar{\kappa}, \delta\bar{\rho}])(t, \mathbf{x}_r; \mathbf{x}_s) = \delta\bar{p}(t, \mathbf{x}_r; \mathbf{x}_s)$$

Variable Density Acoustics

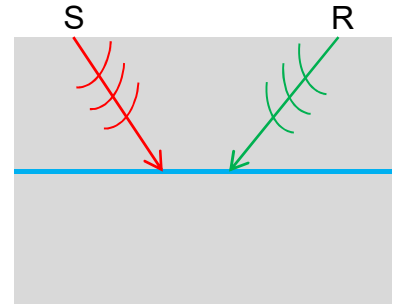
➤ Extended Modeling:

$$\begin{aligned}
 & (\bar{F}[\kappa_0, \rho_0][\delta\bar{\kappa}, \delta\bar{\rho}])(t, \mathbf{x}_r; \mathbf{x}_s) \\
 &= \int d\mathbf{x} \int d\mathbf{h} \int \delta\tau \left(G(t - \tau, \mathbf{x} + \mathbf{h}; \mathbf{x}_r) \frac{1}{\kappa_{0+}} \delta\bar{\kappa} \frac{1}{\kappa_{0-}} \frac{\partial^3}{\partial t^3} G(\tau, \mathbf{x} - \mathbf{h}; \mathbf{x}_s) \right. \\
 & \quad \left. + \nabla G(t - \tau, \mathbf{x} + \mathbf{h}; \mathbf{x}_r) \cdot \left(\frac{1}{\rho_{0+}} \delta\bar{\rho} \frac{1}{\rho_{0-}} \nabla \frac{\partial}{\partial t} G(\tau, \mathbf{x} - \mathbf{h}; \mathbf{x}_s) \right) \right)
 \end{aligned}$$



➤ Extended Imaging (migration):

$$\begin{aligned}
 & (\bar{F}^*[\kappa_0, \rho_0]\delta p)(\mathbf{x}, \mathbf{h}) \\
 &= \left[\frac{1}{\kappa_{0+}} \frac{1}{\kappa_{0-}} \int d\mathbf{x}_r \int d\mathbf{x}_s \int dt \int \delta\tau G(t - \tau, \mathbf{x} + \mathbf{h}; \mathbf{x}_r) \frac{\partial^3}{\partial t^3} G(\tau, \mathbf{x} - \mathbf{h}; \mathbf{x}_s) \delta p(t, \mathbf{x}_r; \mathbf{x}_s) \right. \\
 & \quad \left. \frac{1}{\rho_{0+}} \frac{1}{\rho_{0-}} \int d\mathbf{x}_r \int d\mathbf{x}_s \int dt \int \delta\tau \nabla G(t - \tau, \mathbf{x} + \mathbf{h}; \mathbf{x}_r) \nabla \frac{\partial}{\partial t} G(\tau, \mathbf{x} - \mathbf{h}; \mathbf{x}_s) \delta p(t, \mathbf{x}_r; \mathbf{x}_s) \right]
 \end{aligned}$$



Asymptotic Analysis

- **The (Modified) Normal Operator:**

$$M = (I_t \bar{F}[\kappa_0, \rho_0])^* \partial_{z_s} I_t \partial_{z_r} I_t \bar{F}[\kappa_0, \rho_0]$$

- **Has the form of a Generalized Radon Transform (GRT) operator:**

$$Mu(\mathbf{x}, \mathbf{h}) = \int d\mathbf{x}_r d\mathbf{x}_s d\mathbf{x}' d\mathbf{h}' A(\mathbf{x}_s, \mathbf{x}_r, \mathbf{x}, \mathbf{h}, \mathbf{x}', \mathbf{h}') \delta(\phi(\mathbf{x}_s, \mathbf{x}_r, \mathbf{x}, \mathbf{h}) - \phi(\mathbf{x}_s, \mathbf{x}_r, \mathbf{x}', \mathbf{h}')) u(\mathbf{x}', \mathbf{h}')$$

in which:

$$\phi = T_s + T_r$$

$$A_0(\mathbf{x}_s, \mathbf{x}_r, \mathbf{x}, \mathbf{h}, \mathbf{x}', \mathbf{h}') = \pi^2 a_s a_r \frac{\cos \theta_s}{v(\mathbf{x}_s)} \frac{\cos \theta_r}{v(\mathbf{x}_r)} a'_s a'_r$$

and:

$$A(\mathbf{x}_s, \mathbf{x}_r, \mathbf{x}, \mathbf{h}, \mathbf{x}', \mathbf{h}') = \begin{bmatrix} A_0 \frac{1}{\sqrt{\kappa_0 + \kappa_0 - \kappa'_0 + \kappa'_0 -}} & A_0 \frac{\nabla T_s \nabla T_r}{\sqrt{\rho_0 + \rho_0 - \kappa'_0 + \kappa'_0 -}} \\ A_0 \frac{\nabla T_s \nabla T_r}{\sqrt{\rho_0 + \rho_0 - \kappa'_0 + \kappa'_0 -}} & A_0 \frac{\nabla T_s \nabla T_r \nabla T'_s \nabla T'_r}{\sqrt{\rho_0 + \rho_0 - \rho'_0 + \rho'_0 -}} \end{bmatrix}$$

Asymptotic Analysis

- **Asymptotic analysis of the normal operator reveals:**

$$M[\bar{r}_\kappa, \bar{r}_\rho]^T \cong \int dk_z dk_x dk_h \frac{e^{i(zk_z + xk_x + hk_h)} (q_s + q_r)^2}{256\pi^3 (-ik_z) (s_+^2 q_s^2 + s_-^2 q_r^2 + (s_+^2 + s_-^2) q_s q_r) v_+^2 v_-^2} [\bar{r}_\kappa, \bar{r}_\rho]^T [1, \cos 2\theta] [1, \cos 2\theta]^T$$

where:

$$\bar{r}_\kappa = \frac{\delta \bar{\kappa}}{\sqrt{\kappa_{0+} \kappa_{0-}}} \quad \bar{r}_\rho = \frac{\delta \bar{\rho}}{\sqrt{\rho_{0+} \rho_{0-}}}$$

$$q_s = \frac{\partial T_s}{\partial z} \quad q_r = \frac{\partial T_r}{\partial z} \quad s = \frac{1}{v}$$

- **The κ component:**

$$(M[\delta \bar{\kappa}, \delta \bar{\rho}]^T)_\kappa \cong \int dk_z dk_x dk_h \frac{e^{i(zk_z + xk_x + hk_h)}}{256\pi^3 \kappa_0^3 \rho_0^{-1} (-ik_z)} PB(\delta \bar{\kappa} + (\kappa_0 \rho_0^{-1}) \delta \bar{\rho} \cos 2\theta)$$

in which θ is the scattering angle.

Approximate Born Inversion

- **The variable density acoustics Approximate Inverse:**

$$(I_t \bar{F}^*[\kappa_0, \rho_0]) \partial_{z_s} I_t \partial_{z_r} I_t \bar{F}[\kappa_0, \rho_0][\delta \bar{\kappa}, \delta \bar{\rho}]$$

$$\cong \int dk_z dk_x dk_h \frac{e^{i(zk_z + xk_x + hk_h)}}{256\pi^3 \kappa_0^3 \rho_0^{-1} (-ik_z)} PB(\delta \bar{\kappa} + (\kappa_0 \rho_0^{-1}) \delta \bar{\rho} \cos 2\theta)$$

Define the weights:

$$W_d = I_t (I_t \partial_{z_s}) (I_t \partial_{z_r})$$

$$W_m^{-1} = 32 \kappa_0^3 \rho_0^{-1} \partial_z$$

The approximate Inverse:

$$\bar{F}^\dagger \bar{F} [\delta \bar{\kappa}, \delta \bar{\rho}] = W_m^{-1} \bar{F}^* W_d \bar{F} [\delta \bar{\kappa}, \delta \bar{\rho}]$$

$$\cong 8\pi^3 \int dk_z dk_x dk_h e^{i(zk_z + xk_x + hk_h)} (\delta \bar{\kappa} + (\kappa_0 \rho_0^{-1}) \delta \bar{\rho} \cos 2\theta)$$

- **Angle-dependent Reflectivity:**

$$R(x, \theta) \cong \delta \bar{\kappa} + (\kappa_0 \rho_0^{-1}) \delta \bar{\rho} \cos 2\theta$$

Approximate Born Inversion

- **Approximate Inverse Operator (variable density):**

Similar construction as in Hou and Symes (2017) :

$$\bar{F}^\dagger(\mathbf{x}, \mathbf{h}) = W_m^{-1} \bar{F}^*(\mathbf{x}, \mathbf{h}) W_d$$

The diagram illustrates the construction of the approximate inverse operator. A central equation is enclosed in a blue box. Two blue arrows point from this box to two separate definitions below. The left arrow points to the 'model-side weight' definition, and the right arrow points to the 'data-side weight' definition.

model-side weight

$$W_m^{-1} = 32\kappa_0^3 \rho_0^{-1} \partial_z$$

data-side weight

$$W_d = I_t(I_t \partial_{z_s})(I_t \partial_{z_r})$$

Net Result

- Born data (or simulation)

$$(\bar{F}[\kappa_0, \rho_0][\delta\bar{\kappa}, \delta\bar{\rho}])(t, \mathbf{x}_r; \mathbf{x}_s) = d(t, \mathbf{x}_r; \mathbf{x}_s)$$
- Data-side weight operator

$$W_d d(t, \mathbf{x}_r; \mathbf{x}_s)$$
- Migration (subsurface offset extended)

$$\bar{F}^*[\kappa_0, \rho_0]W_d d(\mathbf{x}, h)$$
- Model-side weight operator

$$W_m^{-1}\bar{F}^*[\kappa_0, \rho_0]W_d d(\mathbf{x}, h)$$
- Angle-domain transformation

$$\mathcal{R}[W_m^{-1}\bar{F}^*[\kappa_0, \rho_0]W_d d(\mathbf{x}, h)](\mathbf{x}, \theta)$$
- AVA inversion

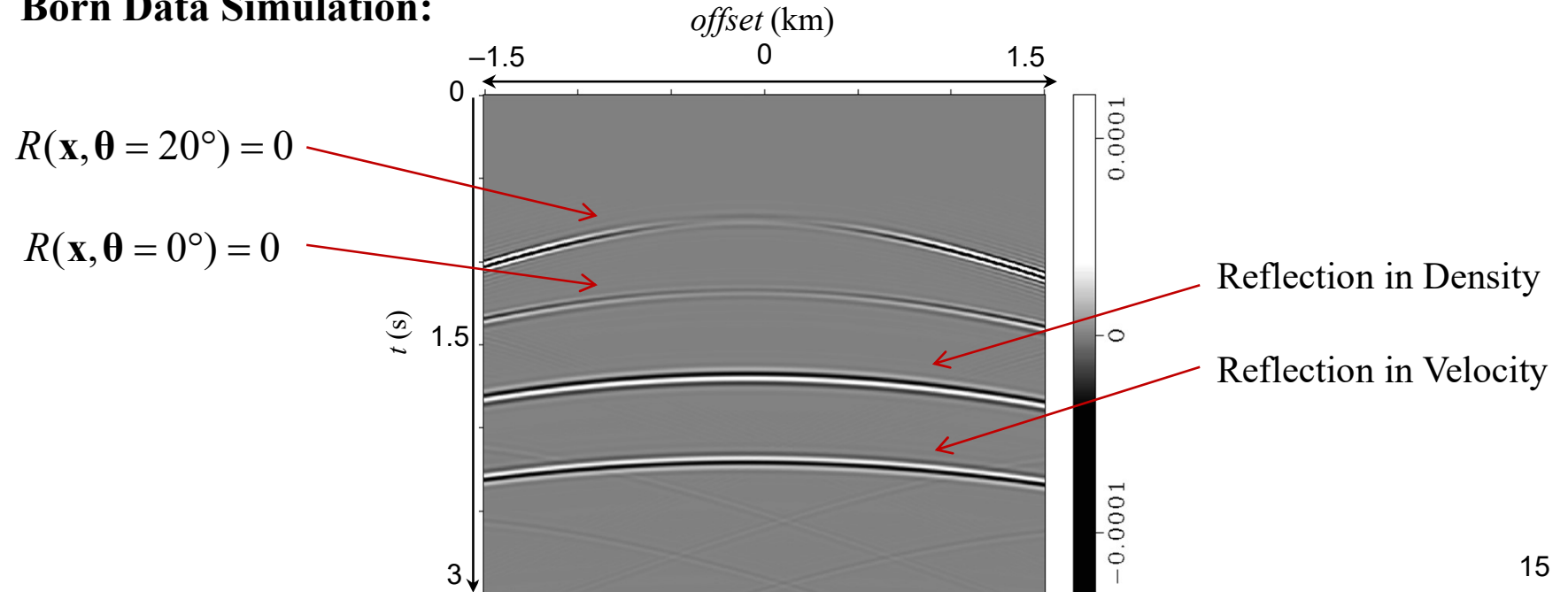
$$R(x, \theta) \cong \delta\bar{\kappa} + (\kappa_0\rho_0^{-1})\delta\bar{\rho}\cos 2\theta$$

Example #1

- Layered Variable Density Media:

Layer #	Depth (km)	Velocity (km/s)	Density (gr/cm ³)
1	0.750	2.000	2.300
2	1.250	2.300	1.963
3	1.750	1.963	2.300
4	2.250	1.963	2.000
5	3.000	2.200	2.000

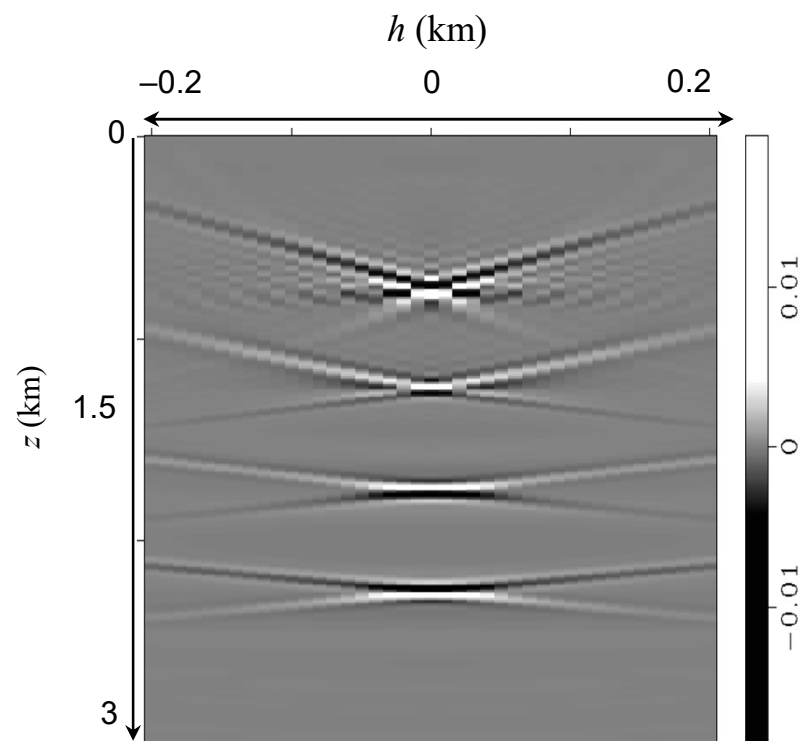
- Born Data Simulation:



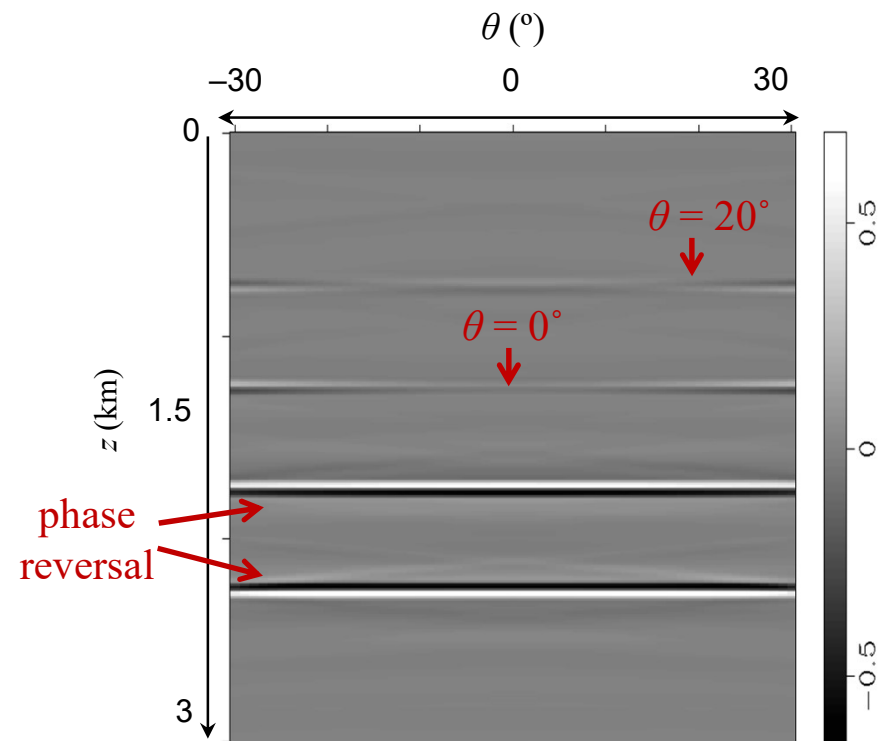
Example #1

- Approximate Inverse:

subsurface offset gather



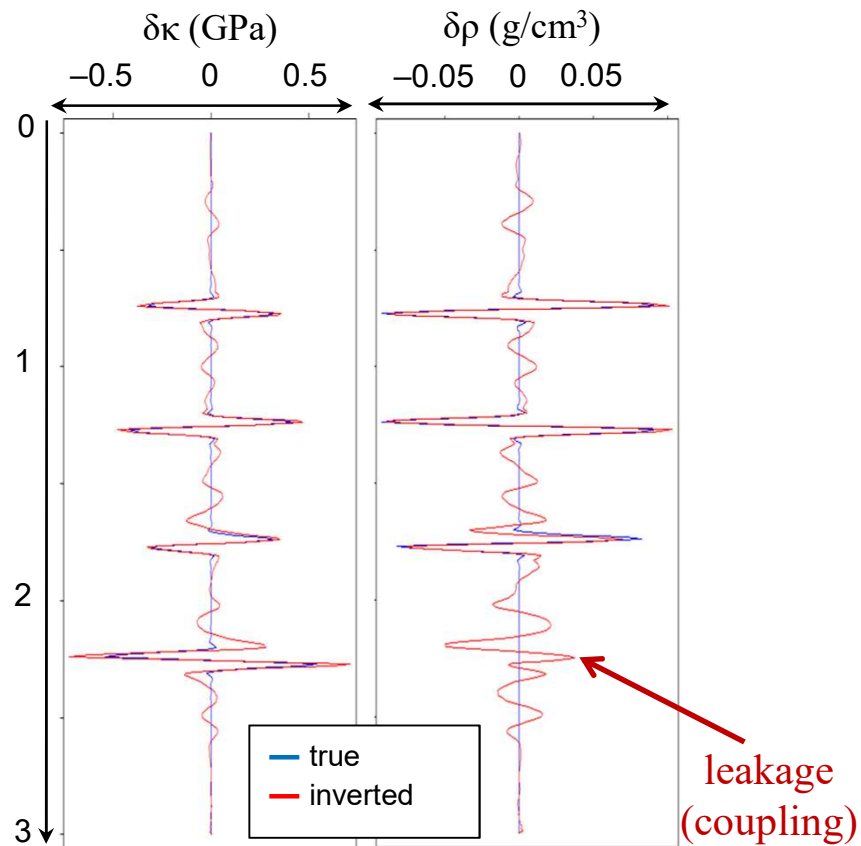
scattering-angle gather



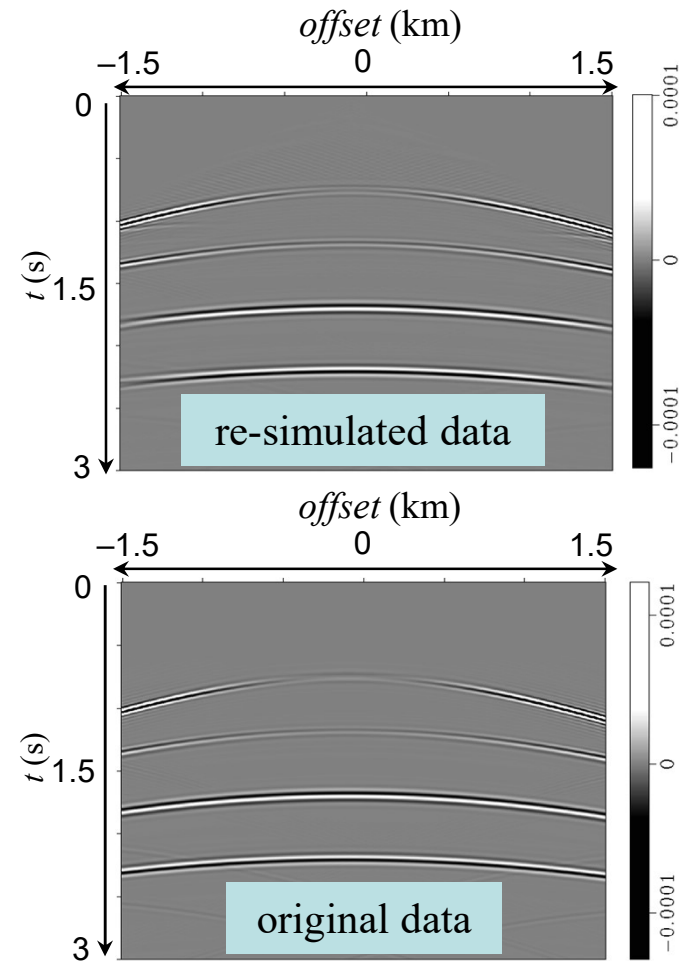
Example #1

- AVA Inversion:

model perturbations

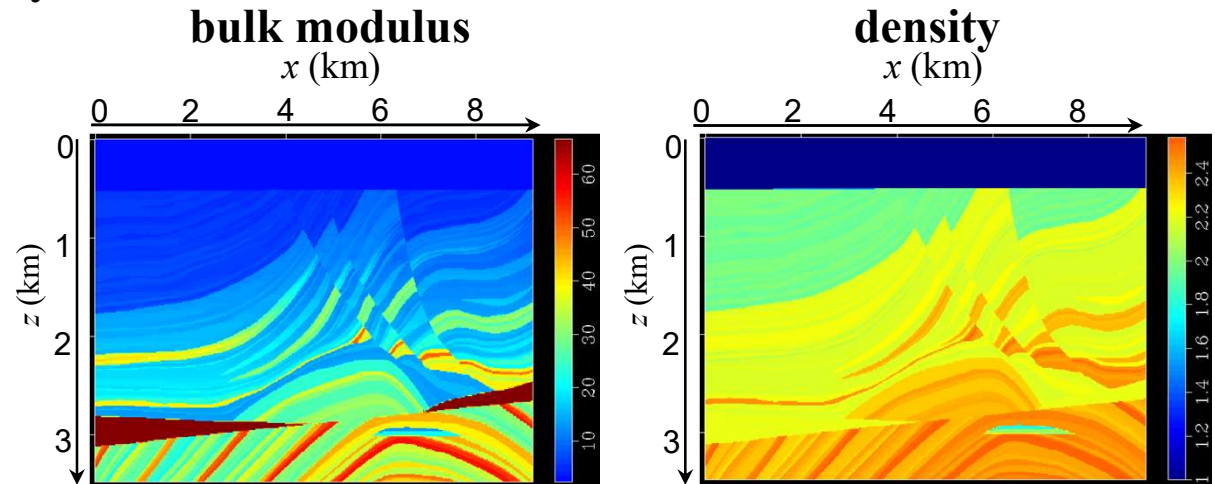


Born data re-simulation



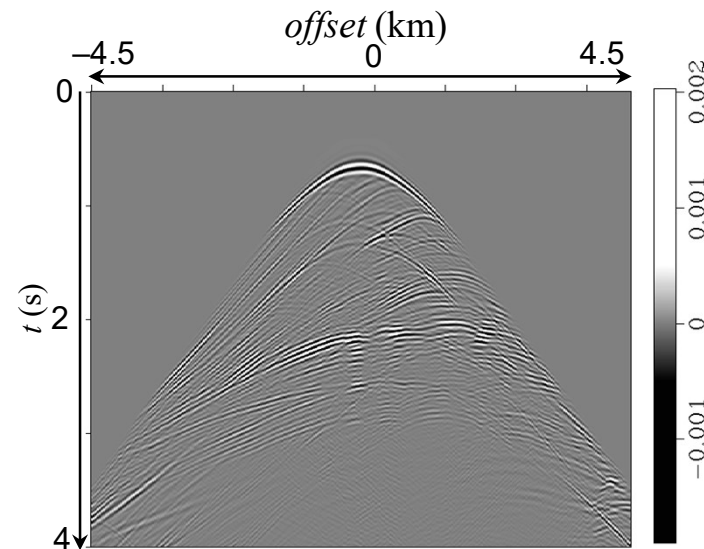
Example #2

- Marmousi Variable Density Model:**



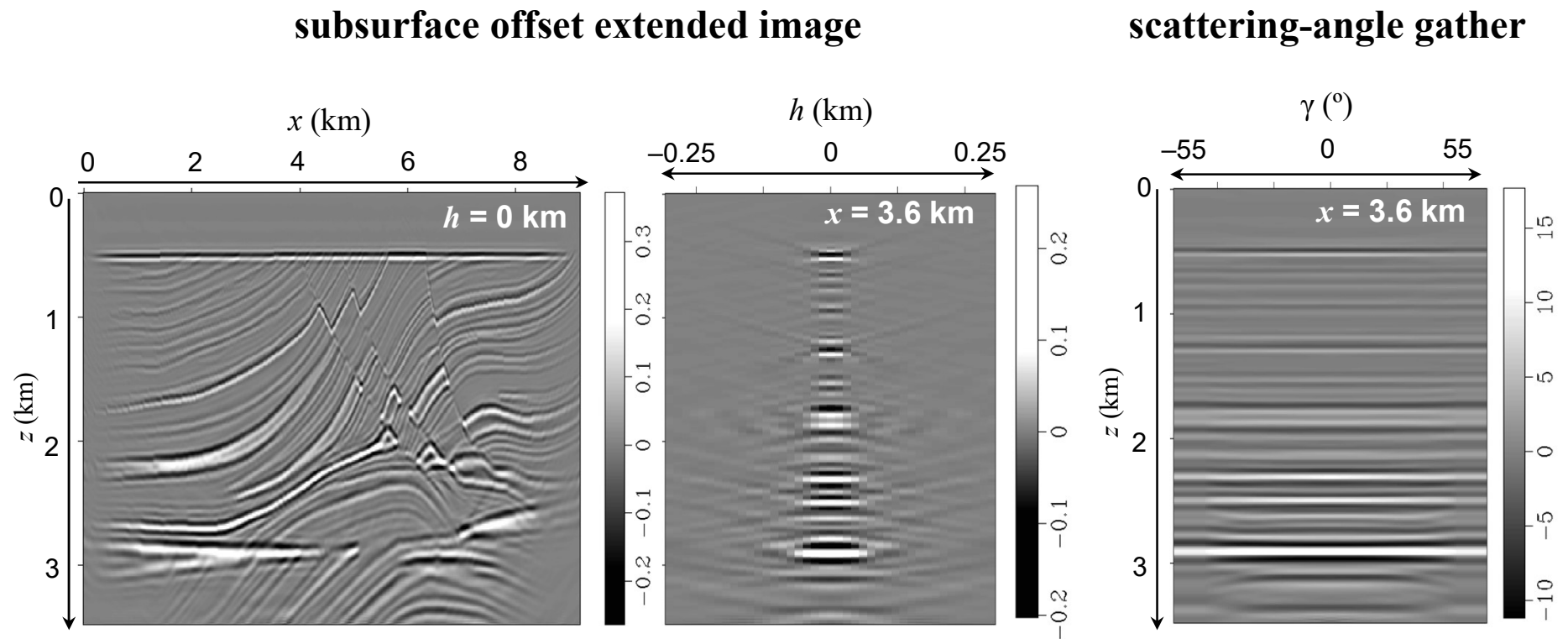
- Born data simulation:**

Shot gather at $x = 4.56\text{km}$:



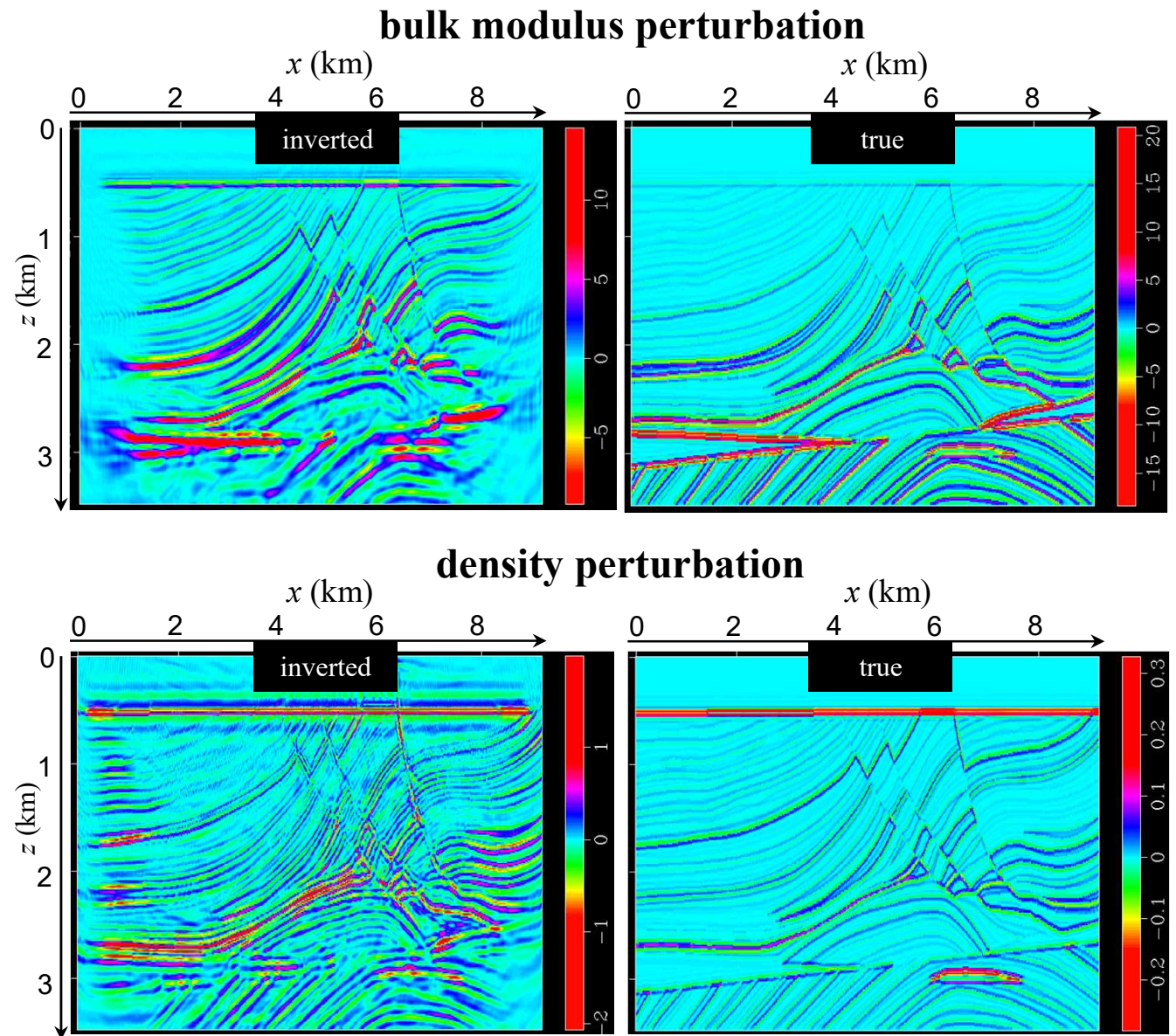
Example #2

- Approximate Inverse:



Example #2

- AVA Inversion:

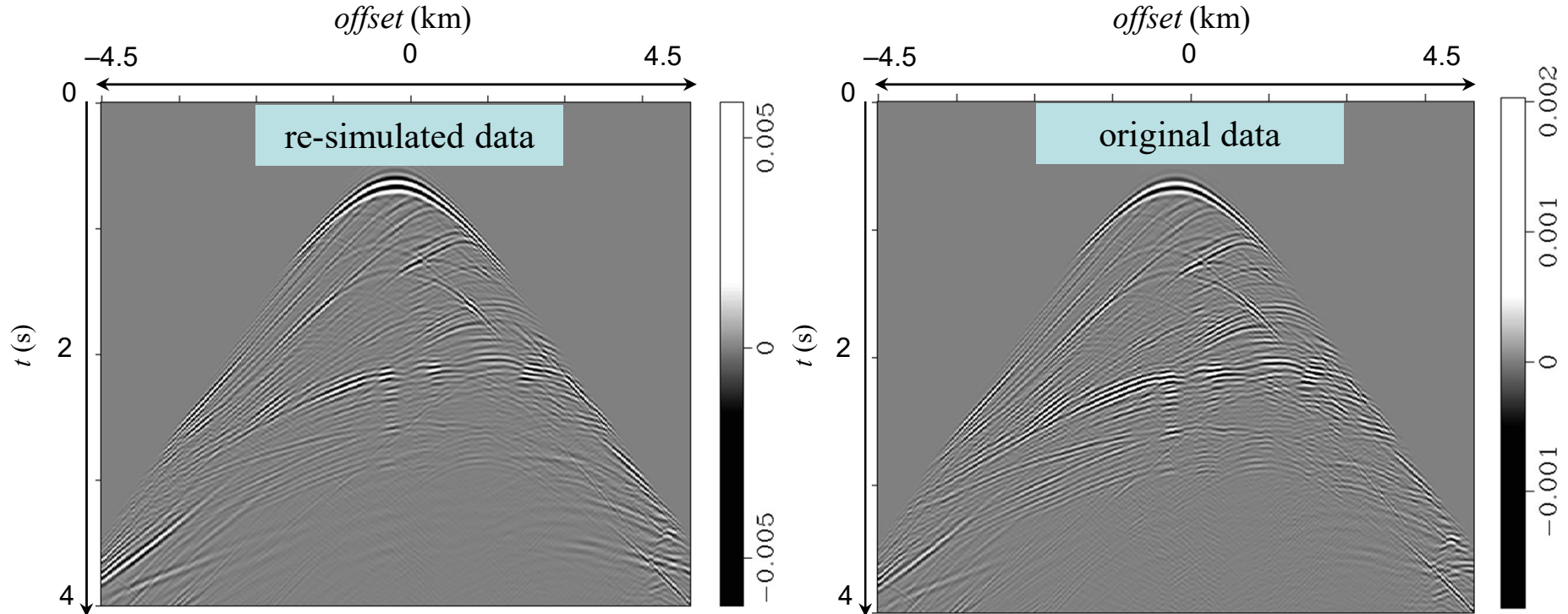


Example #2

- AVA Inversion:

Born data re-simulation

Shot gather at $x = 4.56\text{km}$:



Concluding Remarks

- Approximate Born Inversion = weighted variation of RTM (subsurface offset extended)
- Explicit data and model side weights, independent of ray-theory properties.
- Post-migration angle-transformation naturally exposes the angle-dependent reflectivity.
- Simultaneous inversion of model perturbations: $\delta\kappa$, $\delta\rho$
- May be extended similarly to Elastic Media (pure and converted modes).
- Main Limitations:
 - 1) asymptotic approximation (high-frequency, short-scale)
 - 2) data bandwidth
 - 3) image is assumed to be well focused (optimal velocity)
 - 3) coupling between inverted parameters
- The Approximate Inverse application may accelerate the convergence of LSM
(SEG 2018, Tue, SPMI2, “Accelerated Acoustic LSM” by Dafni & Symes).

Acknowledgments:

- ❖ The Rice University Inversion Project (**TRIP**) members and sponsors.
- ❖ The Rice University Research Computing Support Group (**RCSG**).
- ❖ The Texas Advanced Computing Center (**TACC**).
- ❖ **Shell International Exploration and Production Inc.** for their sponsorship.

Thank you!