

Parameter-selection strategy for density in frequency-domain elastic waveform inversion

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Motivation

- Needs for elastic full waveform inversion (FWI)
 - *multicomponent* data are *commonly* acquired
 - more *reliable* subsurface *information* is needed

Motivation

- Limitations of elastic FWI
 - *more* parameters than *acoustic* FWI
 - > (velocities and density
or Lamé constants and density)
 - easily stuck in *local* minima
 - > it is assumed that Poisson's ratio and density are constant (Brossier et al. 2009; Brossier et al. 2010; Bae et al. 2010; Lee et al. 2010).

Motivation

- Density
 - for *acoustic* or *elastic* impedance (Connolly, 1999), *density* is needed
 - in *conventional* elastic FWI, velocities are *properly* restored, but *density* is *very* difficult to recover
(Forgues and Lambare, 1997, Choi et al., 2008, Virieux and Operto, 2009).

Objective

- Develop an inversion strategy for density
 - Tarantola (1986) proposed a parameter-selection strategy based on *sensitivity* analysis.
 - propose a new parameter-selection strategy
 - > inversion is performed over two stages
 - > velocities and density are recovered *sequentially*

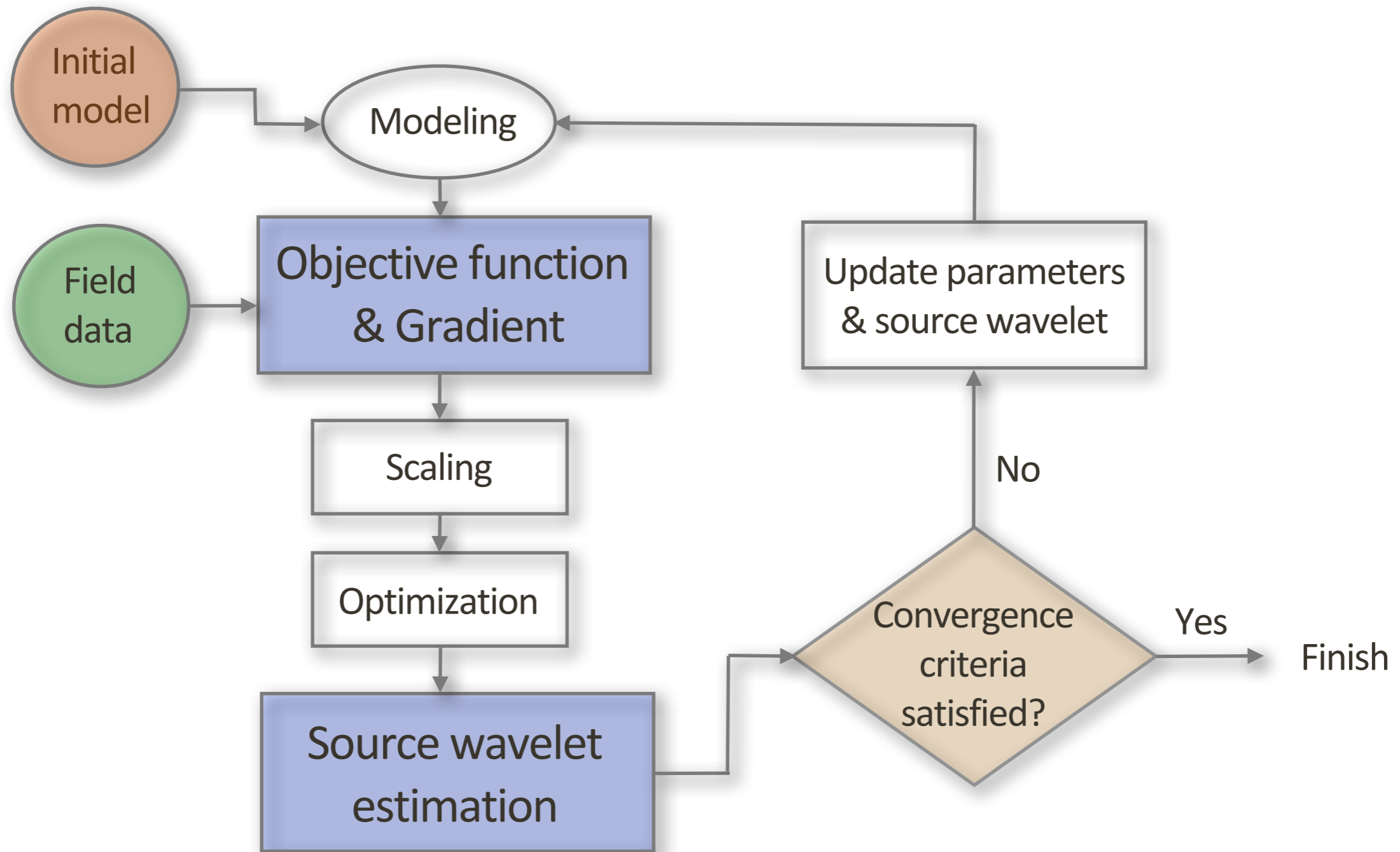
Contents

- FWI algorithm
- Conventional FWI & examples
- Parameter-selection strategy & examples
- Conclusions

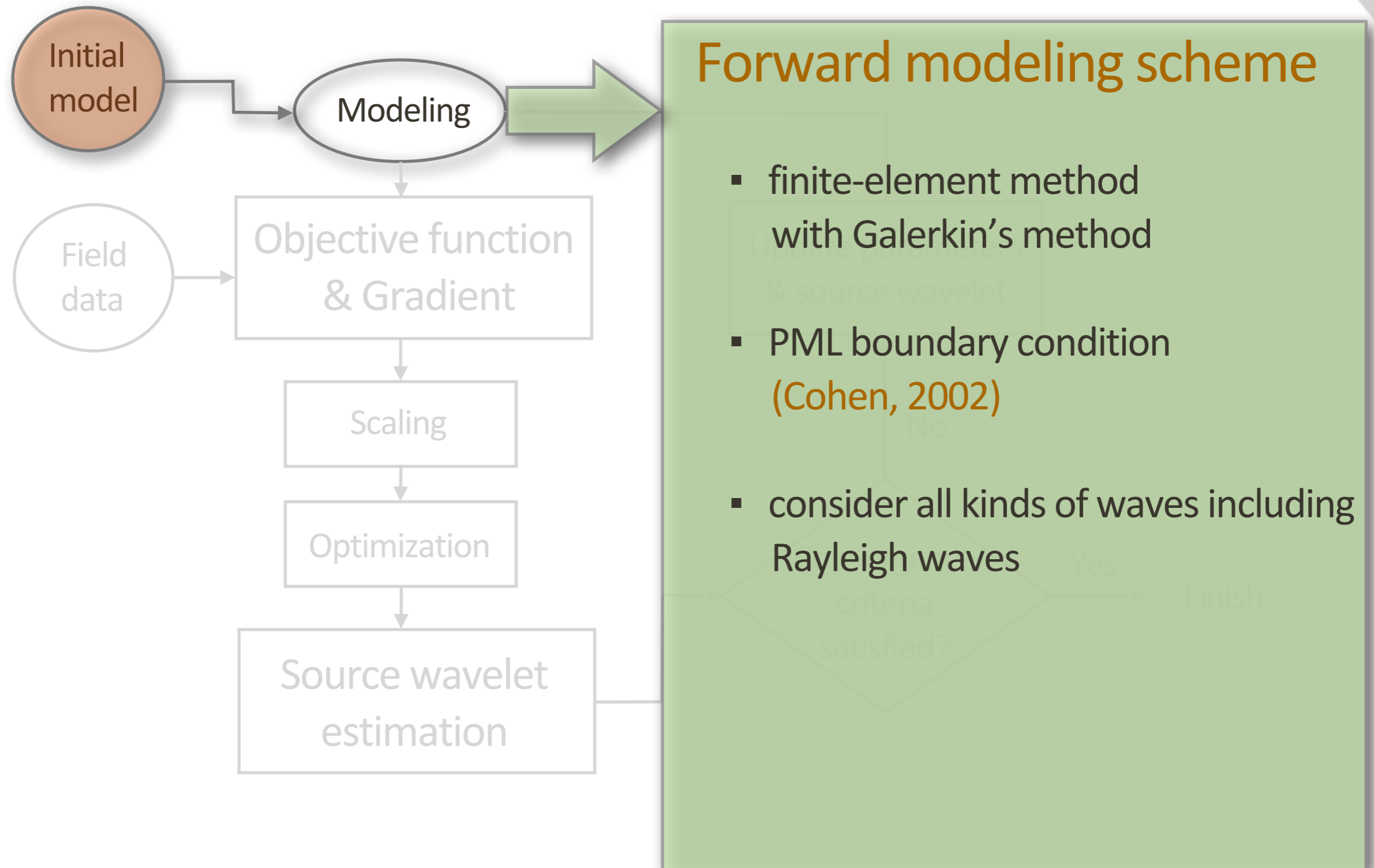
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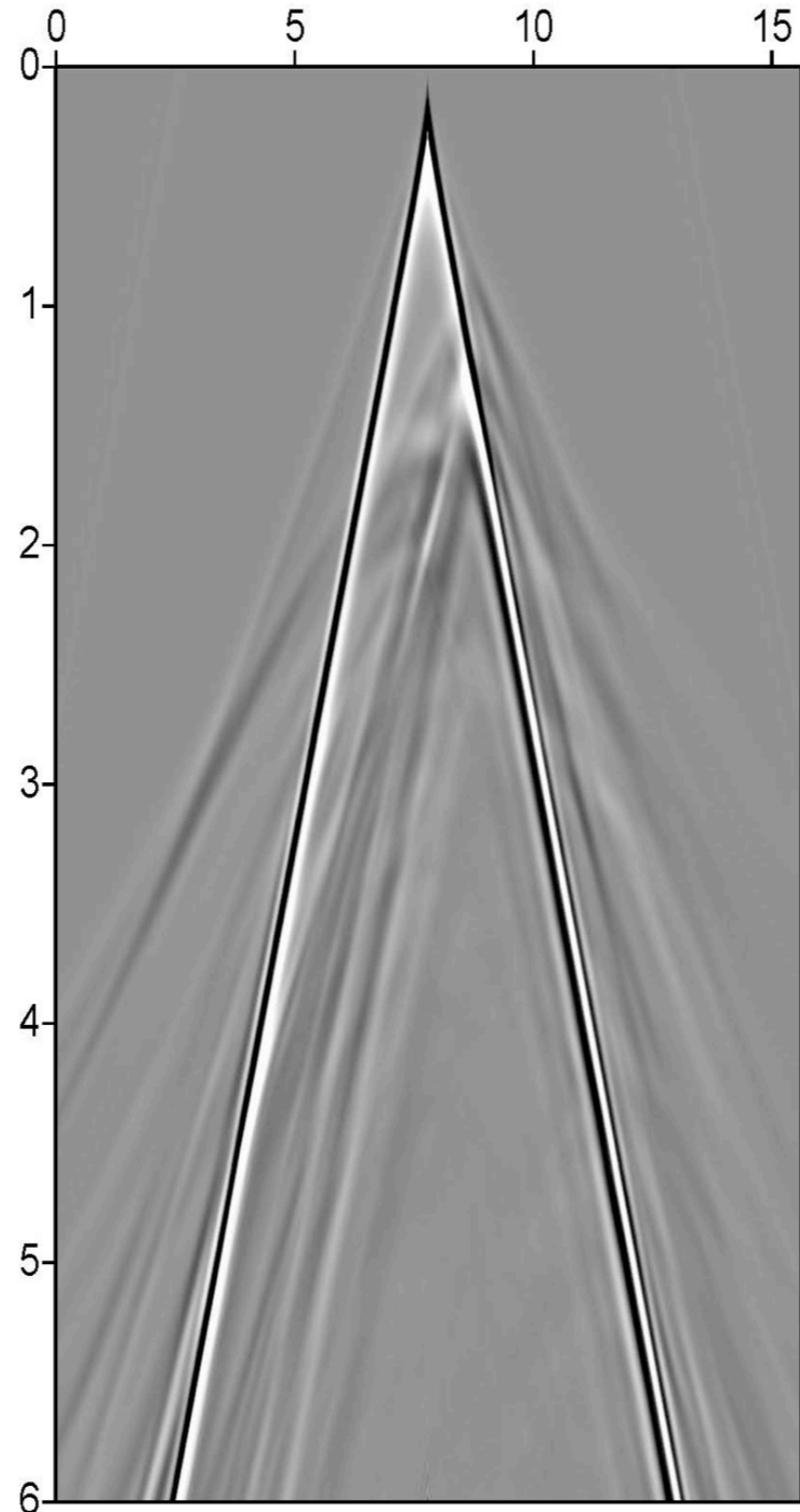
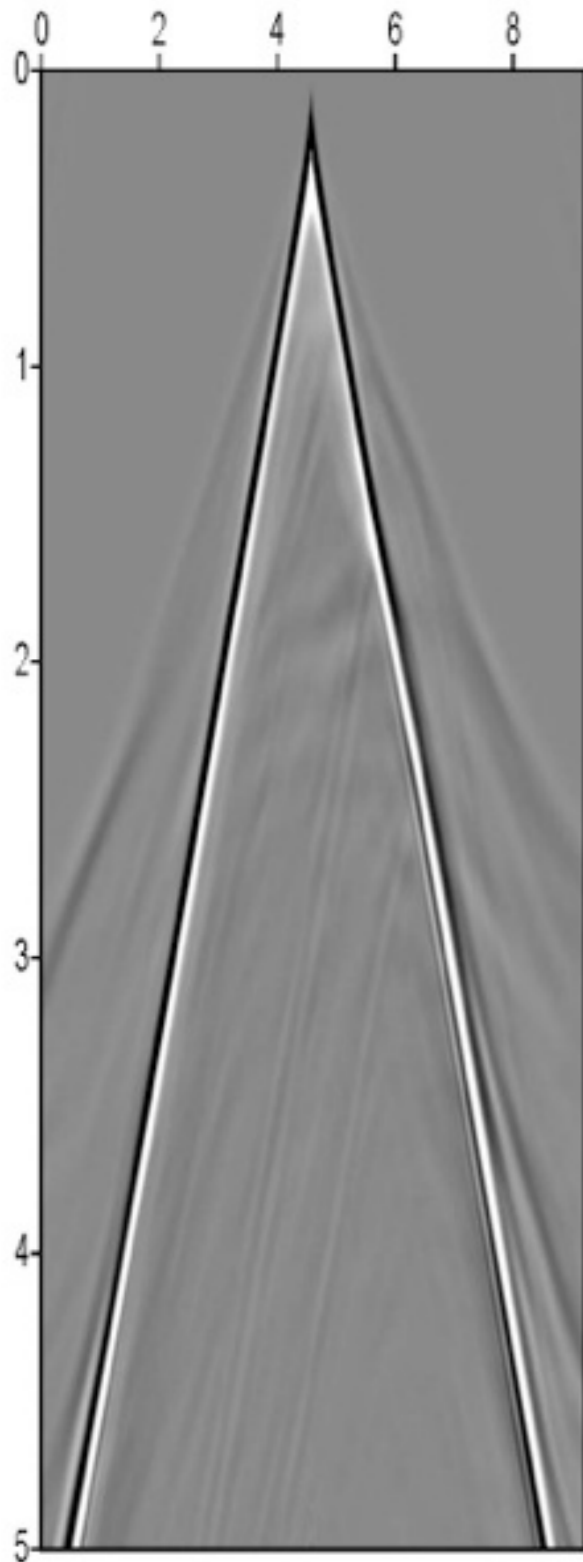
FWI algorithm



FWI algorithm



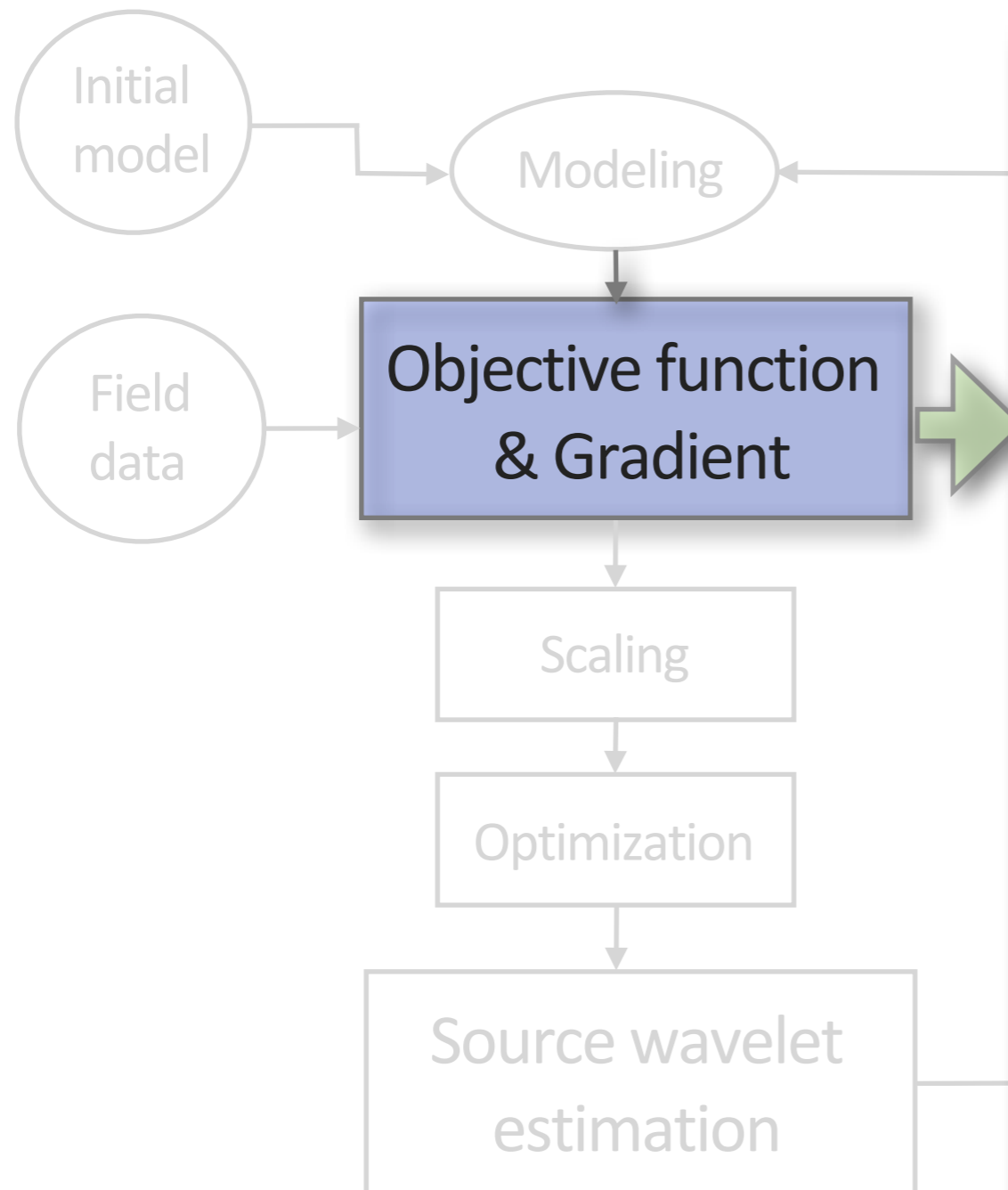
FWI algorithm



Forward modeling scheme

- finite-element method with Galerkin's method
- PML boundary condition
- consider all kinds of waves including Rayleigh waves

FWI algorithm



Objective function

$$\min_{\mathbf{p}} \frac{1}{2} \sum_{\omega} \sum_s \left\| \tilde{\mathbf{u}}_s(\mathbf{p}) - \tilde{\mathbf{d}}_s \right\|_2^2$$

Gradient

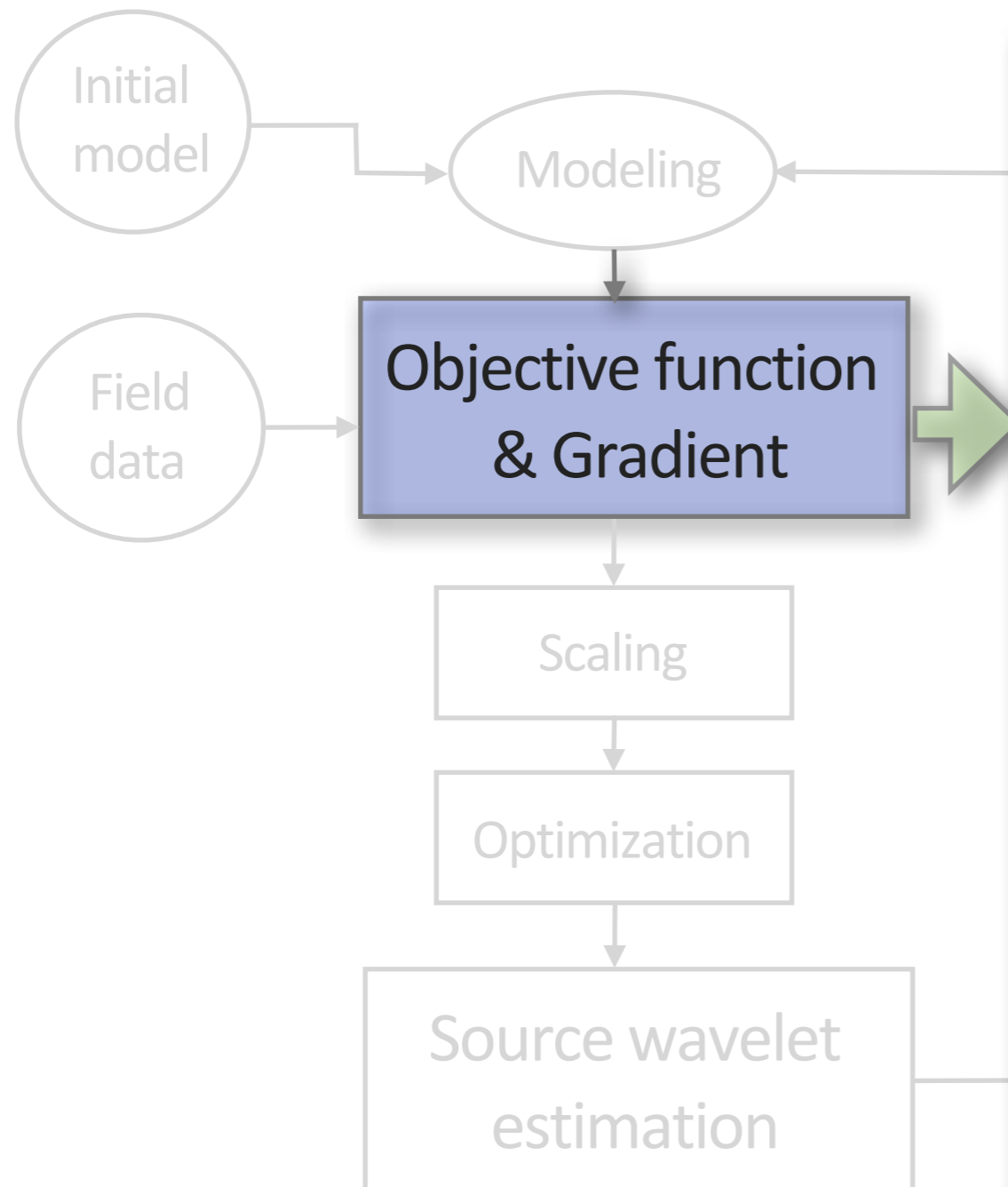
$$\frac{\partial E}{\partial p_k} = \sum_{\omega} \sum_s \operatorname{Re} \left[\left(\frac{\partial \tilde{\mathbf{u}}_s}{\partial p_k} \right)^T (\tilde{\mathbf{u}}_s - \tilde{\mathbf{d}}_s)^* \right]$$

$$\mathbf{S} \tilde{\mathbf{u}}_s = \mathbf{f}_s$$

$$\frac{\partial \tilde{\mathbf{u}}_s}{\partial p_k} = \mathbf{S}^{-1} \left(-\frac{\partial \mathbf{S}}{\partial p_k} \tilde{\mathbf{u}}_s \right) = \mathbf{S}^{-1} (\mathbf{f}_{s,k}^v)$$

$$\therefore \frac{\partial E}{\partial p_k} = \sum_{\omega} \sum_s \operatorname{Re} \left[(\mathbf{f}_{s,k}^v)^T (\mathbf{S}^{-1})^T (\tilde{\mathbf{u}}_s - \tilde{\mathbf{d}}_s)^* \right]$$

FWI algorithm



Objective function

$$\min_{\mathbf{p}} \frac{1}{2} \sum_{\omega} \sum_s \left\| \tilde{\mathbf{u}}_s(\mathbf{p}) - \tilde{\mathbf{d}}_s \right\|_2^2$$

Gradient

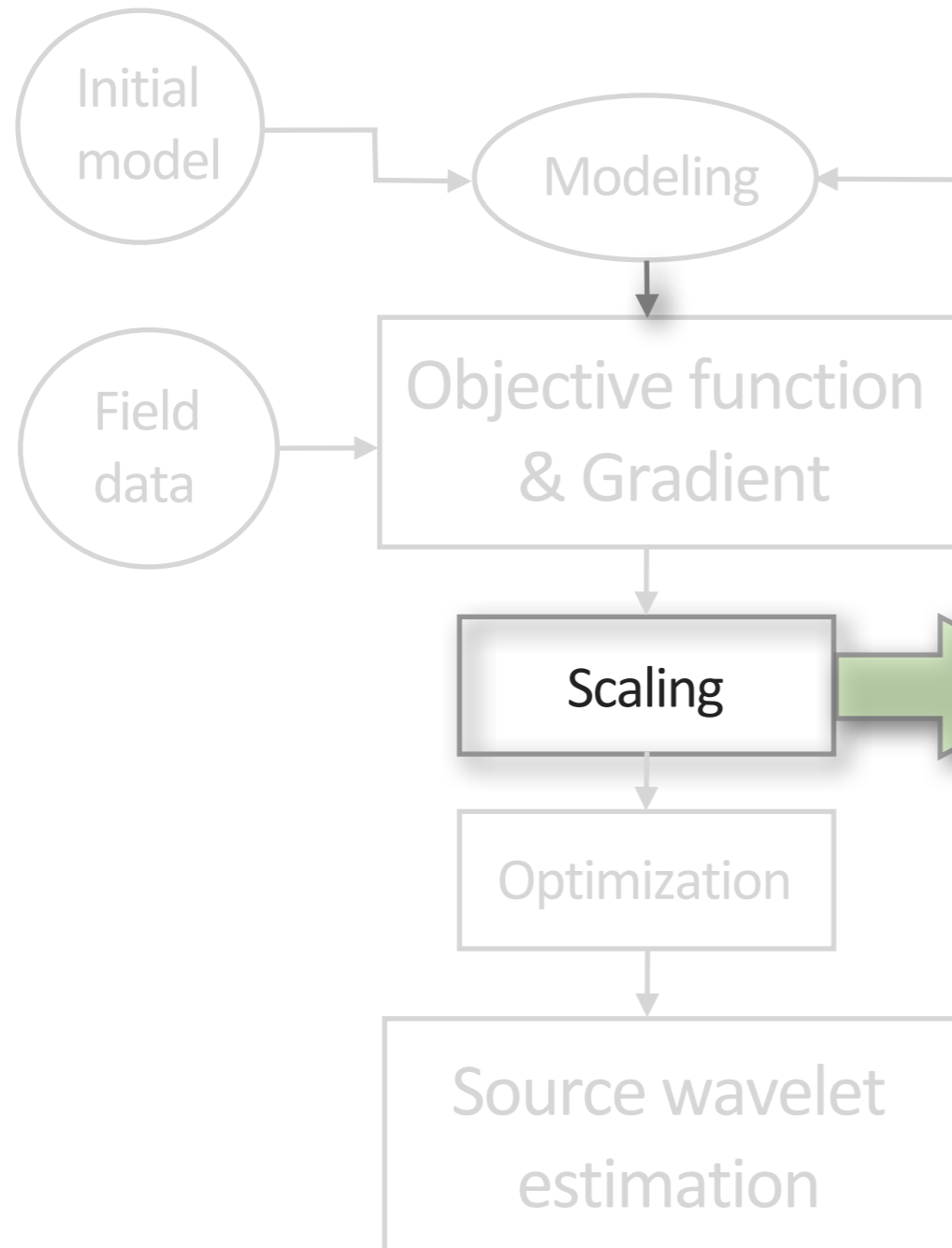
$$\frac{\partial E}{\partial p_k} = \sum_{\omega} \sum_s \operatorname{Re} \left[\left(\frac{\partial \tilde{\mathbf{u}}_s}{\partial p_k} \right)^T (\tilde{\mathbf{u}}_s - \tilde{\mathbf{d}}_s)^* \right]$$

$$\mathbf{S} \tilde{\mathbf{u}}_s = \mathbf{f}_s$$

$$\frac{\partial \tilde{\mathbf{u}}_s}{\partial p_k} = \mathbf{S}^{-1} \left(-\frac{\partial \mathbf{S}}{\partial p_k} \tilde{\mathbf{u}}_s \right) = \mathbf{S}^{-1} (\mathbf{f}_{s,k}^v)$$

$$\therefore \frac{\partial E}{\partial p_k} = \sum_{\omega} \sum_s \operatorname{Re} \left[\left(\mathbf{f}_{s,k}^v \right)^T (\mathbf{S}^{-1})^T (\tilde{\mathbf{u}}_s - \tilde{\mathbf{d}}_s)^* \right]$$

FWI algorithm



Approximate-Hessian matrix

$$\mathbf{H}_a = \sum_s \left(\frac{\partial \tilde{\mathbf{u}}_s}{\partial \mathbf{p}} \right)^T \left(\frac{\partial \tilde{\mathbf{u}}_s}{\partial \mathbf{p}} \right)^*$$

$$= \sum_s (\mathbf{F}_s^v)^T (\mathbf{S}^{-1})^T (\mathbf{S}^{-1})^* (\mathbf{F}_s^v)^*$$

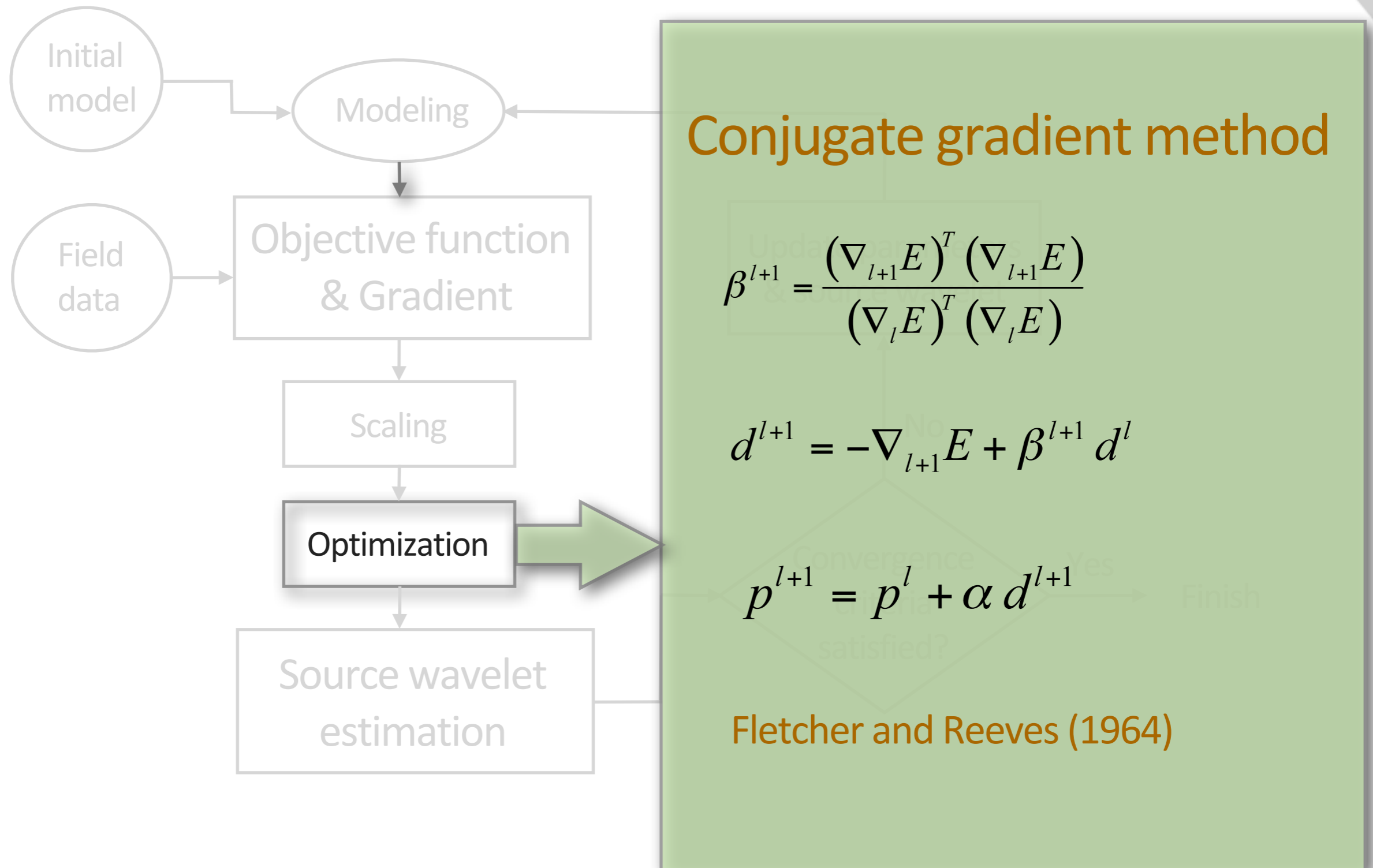
Pseudo-Hessian matrix

$$\mathbf{H}_p = \sum_s (\mathbf{F}_s^v)^T (\mathbf{F}_s^v)^*$$

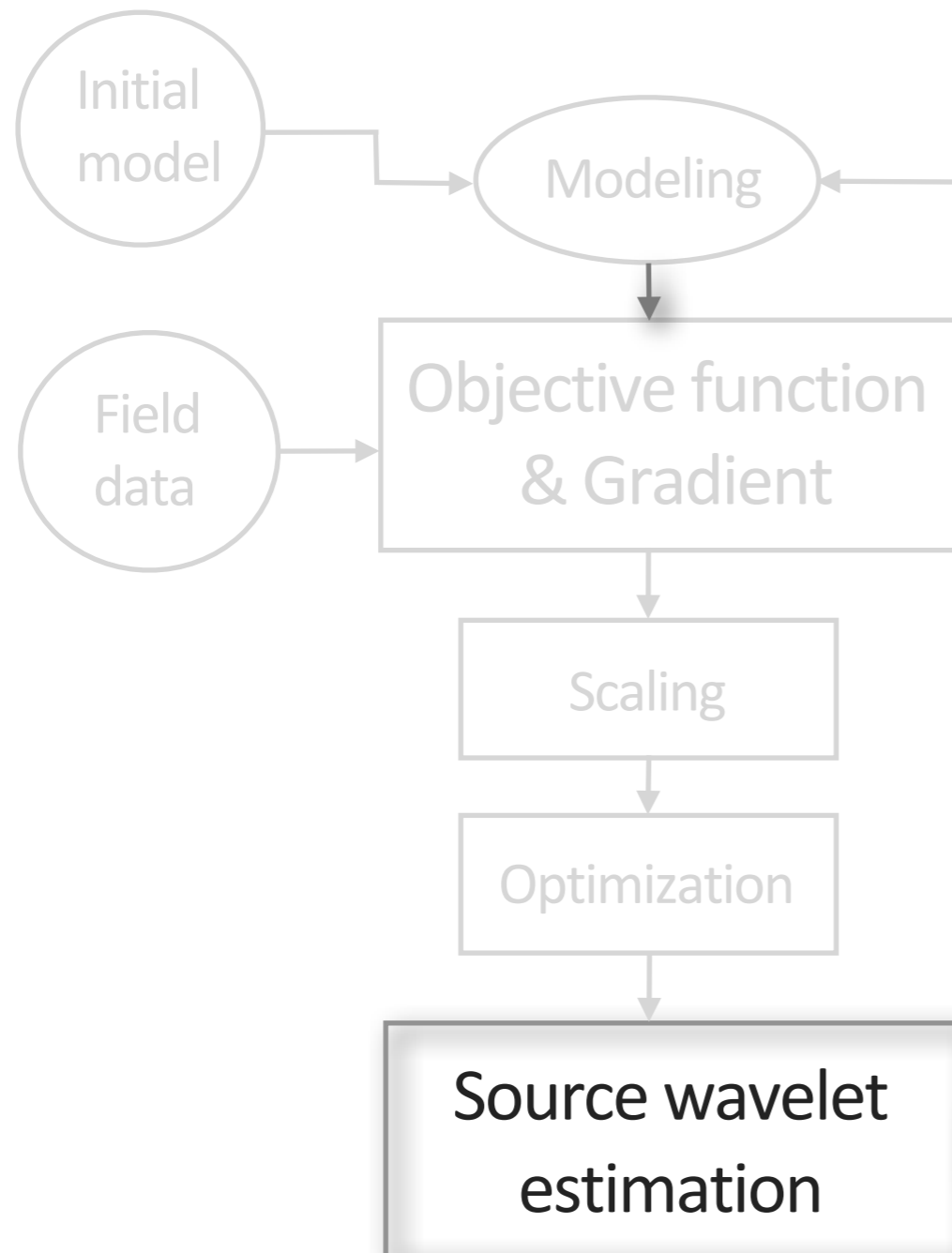
For scaling

$$\nabla E = \sum_{\omega} \left(\frac{\sum_s \text{Re} \left[(\mathbf{F}_s^v)^T (\mathbf{S}^{-1})^T (\tilde{\mathbf{u}} - \tilde{\mathbf{d}})^* \right]}{\sum_s \text{diag} \left\{ (\mathbf{F}_s^v)^T (\mathbf{F}_s^v)^* + \phi \mathbf{I} \right\}} \right)$$

FWI algorithm



FWI algorithm



Source wavelet estimation

L2-norm objective function and the full Hessian matrix

$$\min_w \frac{1}{2} \sum_{\omega} \sum_s \left\| \tilde{\mathbf{g}}_s^T w - \tilde{\mathbf{d}}_s \right\|_2^2$$

$$w = \frac{\sum_{\omega} \sum_s \tilde{\mathbf{g}}_s^T \tilde{\mathbf{d}}_s}{\sum_{\omega} \sum_s \tilde{\mathbf{g}}_s^T \tilde{\mathbf{g}}_s}$$

Yes

Finish

Contents

- FWI algorithm
- **Conventional FWI & examples**
- Parameter-selection strategy & examples
- Conclusions

Conventional FWI

● Conventional Method I

- elastic wave equations parameterized by Lamé constants and density:

$$-\rho\omega^2\tilde{u} = \frac{\partial}{\partial x} \left((\lambda + 2\mu) \frac{\partial \tilde{u}}{\partial x} + \lambda \frac{\partial \tilde{v}}{\partial z} \right) + \frac{\partial}{\partial z} \left(\mu \left(\frac{\partial \tilde{v}}{\partial x} + \frac{\partial \tilde{u}}{\partial z} \right) \right)$$

$$-\rho\omega^2\tilde{v} = \frac{\partial}{\partial x} \left(\mu \left(\frac{\partial \tilde{v}}{\partial x} + \frac{\partial \tilde{u}}{\partial z} \right) \right) + \frac{\partial}{\partial z} \left(\lambda \frac{\partial \tilde{u}}{\partial x} + (\lambda + 2\mu) \frac{\partial \tilde{v}}{\partial z} \right)$$

- virtual sources for each parameters:

$$\left(\mathbf{f}^v \right)_\lambda = -\frac{\partial \mathbf{S}}{\partial \lambda} \tilde{\mathbf{u}}, \quad \left(\mathbf{f}^v \right)_\mu = -\frac{\partial \mathbf{S}}{\partial \mu} \tilde{\mathbf{u}}, \quad \left(\mathbf{f}^v \right)_\rho = -\frac{\partial \mathbf{S}}{\partial \rho} \tilde{\mathbf{u}}$$

Conventional FWI

● Conventional Method II

- elastic wave equations parameterized by velocities and density:

$$-\rho\omega^2\tilde{u} = \frac{\partial}{\partial x} \left(\rho\alpha^2 \frac{\partial\tilde{u}}{\partial x} + (\rho\alpha^2 - 2\rho\beta) \frac{\partial\tilde{v}}{\partial z} \right) + \frac{\partial}{\partial z} \left(\rho\beta^2 \left(\frac{\partial\tilde{v}}{\partial x} + \frac{\partial\tilde{u}}{\partial z} \right) \right)$$

$$-\rho\omega^2\tilde{v} = \frac{\partial}{\partial x} \left(\rho\beta^2 \left(\frac{\partial\tilde{v}}{\partial x} + \frac{\partial\tilde{u}}{\partial z} \right) \right) + \frac{\partial}{\partial z} \left((\rho\alpha^2 - 2\rho\beta^2) \frac{\partial\tilde{u}}{\partial x} + \rho\alpha^2 \frac{\partial\tilde{v}}{\partial z} \right)$$

- virtual sources for each parameters:
assumption: parameters are independent of each other

$$\left(\mathbf{f}^v\right)_\alpha = -\frac{\partial\mathbf{S}}{\partial\alpha}\tilde{\mathbf{u}}, \quad \left(\mathbf{f}^v\right)_\beta = -\frac{\partial\mathbf{S}}{\partial\beta}\tilde{\mathbf{u}}, \quad \left(\mathbf{f}^v\right)_\rho = -\frac{\partial\mathbf{S}}{\partial\rho}\tilde{\mathbf{u}}$$

Conventional FWI

● Conventional Method II

- virtual sources using the chain rule (Mora, 1987)
assumption: velocities are dependent on density

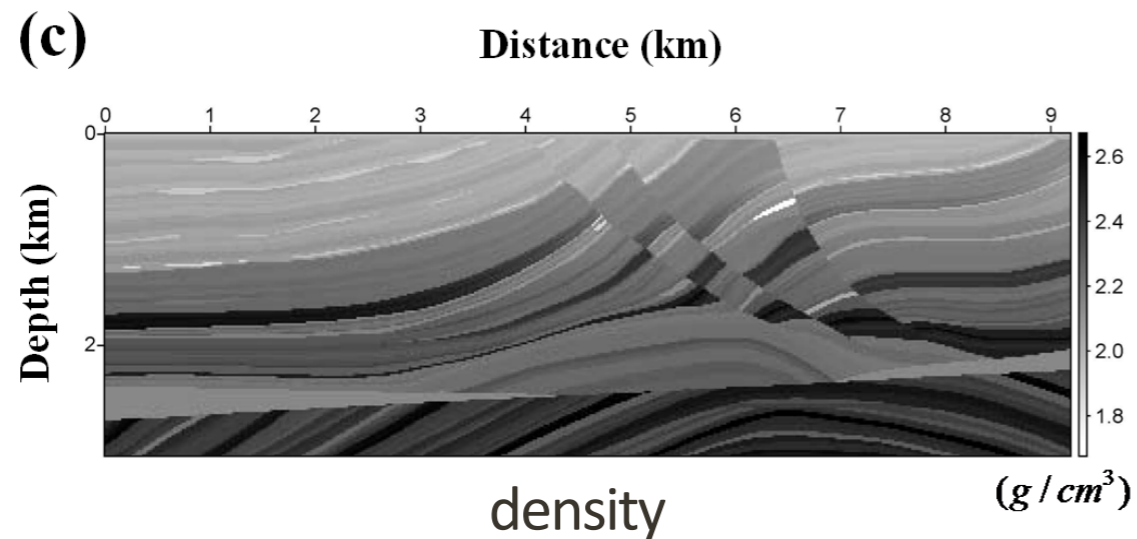
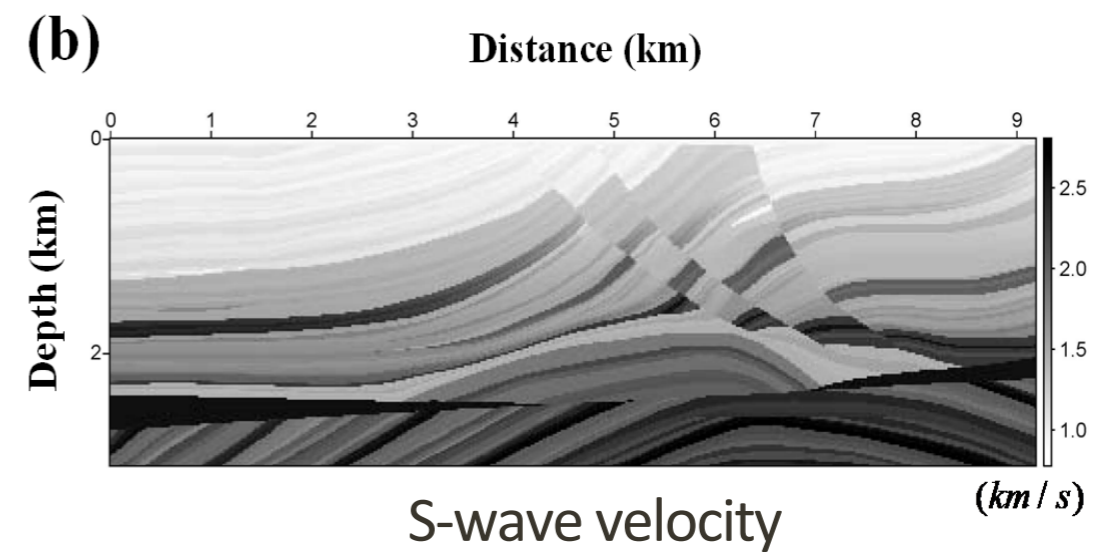
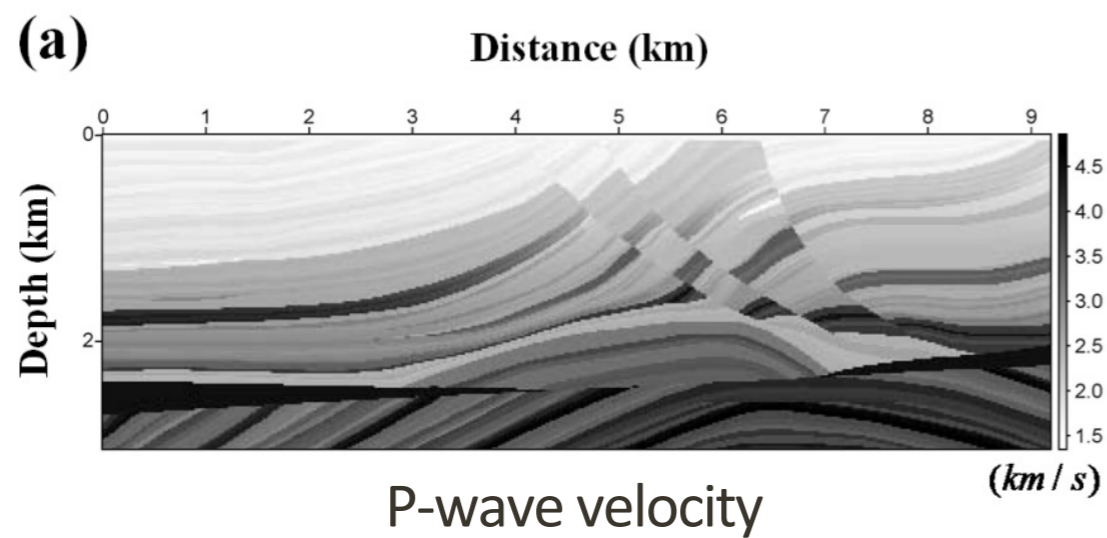
$$\left(\mathbf{f}^v\right)_\alpha = -\left[\frac{\partial \mathbf{S}}{\partial \lambda} \frac{\partial \lambda}{\partial \alpha} + \frac{\partial \mathbf{S}}{\partial \mu} \frac{\partial \mu}{\partial \alpha} + \frac{\partial \mathbf{S}}{\partial \rho} \frac{\partial \rho}{\partial \alpha}\right] \tilde{\mathbf{u}} = -\left[\frac{\partial \mathbf{S}}{\partial \lambda} 2\rho\alpha\right] \tilde{\mathbf{u}}$$

$$\left(\mathbf{f}^v\right)_\beta = -\left[\frac{\partial \mathbf{S}}{\partial \lambda} \frac{\partial \lambda}{\partial \beta} + \frac{\partial \mathbf{S}}{\partial \mu} \frac{\partial \mu}{\partial \beta} + \frac{\partial \mathbf{S}}{\partial \rho} \frac{\partial \rho}{\partial \beta}\right] \tilde{\mathbf{u}} = -\left[-\frac{\partial \mathbf{S}}{\partial \lambda} 4\rho\beta + \frac{\partial \mathbf{S}}{\partial \mu} 2\rho\beta\right] \tilde{\mathbf{u}}$$

$$\left(\mathbf{f}^v\right)_\rho = -\left[\frac{\partial \mathbf{S}}{\partial \lambda} \frac{\partial \lambda}{\partial \rho} + \frac{\partial \mathbf{S}}{\partial \mu} \frac{\partial \mu}{\partial \rho} + \frac{\partial \mathbf{S}}{\partial \rho}\right] \tilde{\mathbf{u}} = -\left[-\frac{\partial \mathbf{S}}{\partial \lambda} (\alpha^2 - 2\beta^2) + \frac{\partial \mathbf{S}}{\partial \mu} \beta^2 + \frac{\partial \mathbf{S}}{\partial \rho}\right] \tilde{\mathbf{u}}$$

Elastic Marmousi-2 model

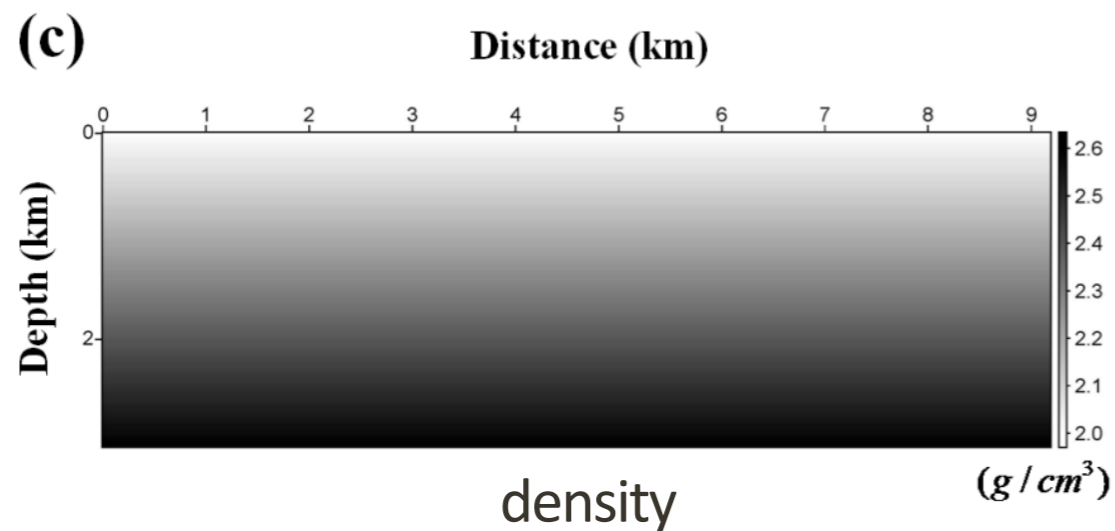
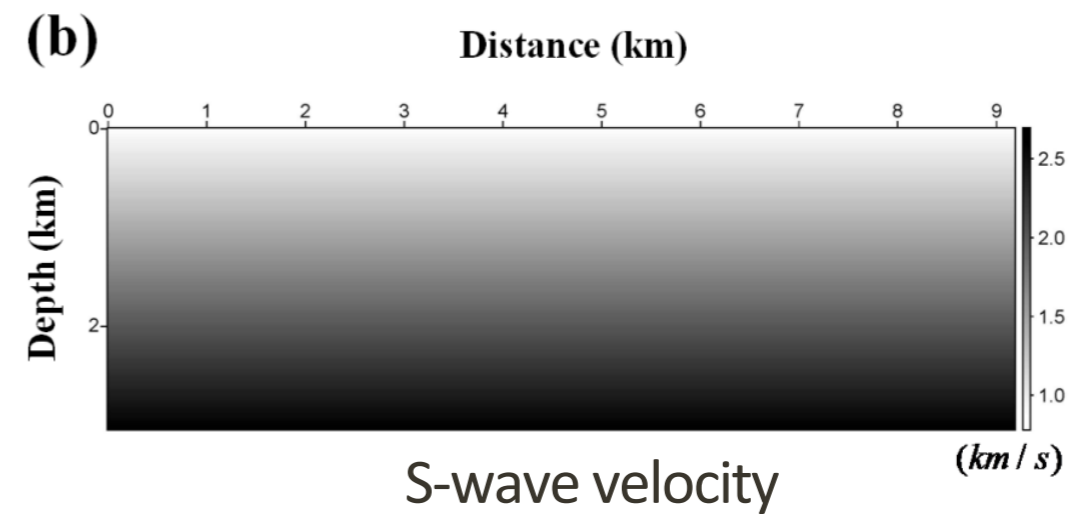
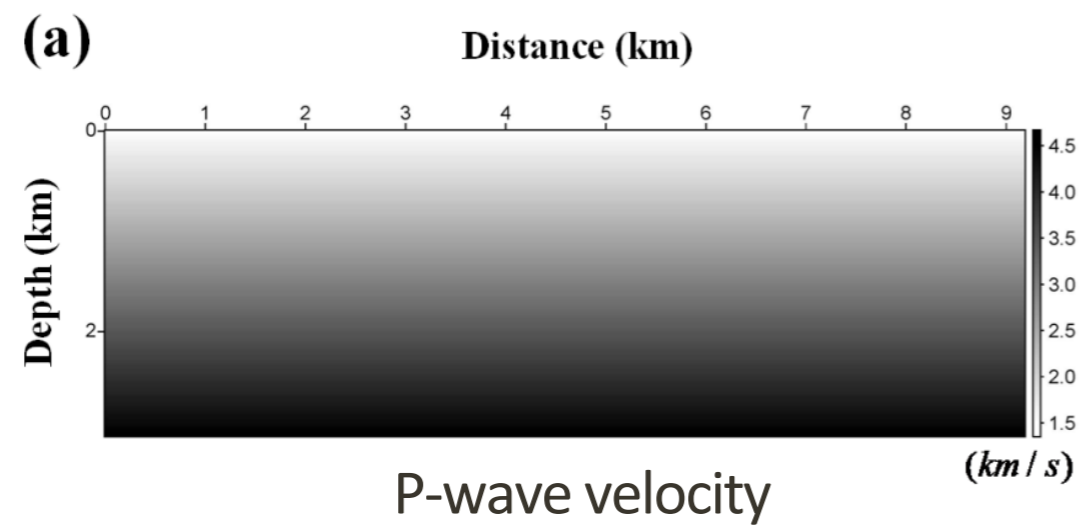
● True model



Parameters	Values
dimension	9.2 km x 3.04 km
no. of source	219
source interval	0.04 km
no. of receiver	461
receiver interval	0.02 km
recording time	5 s
frequency range	0.2 - 10 Hz

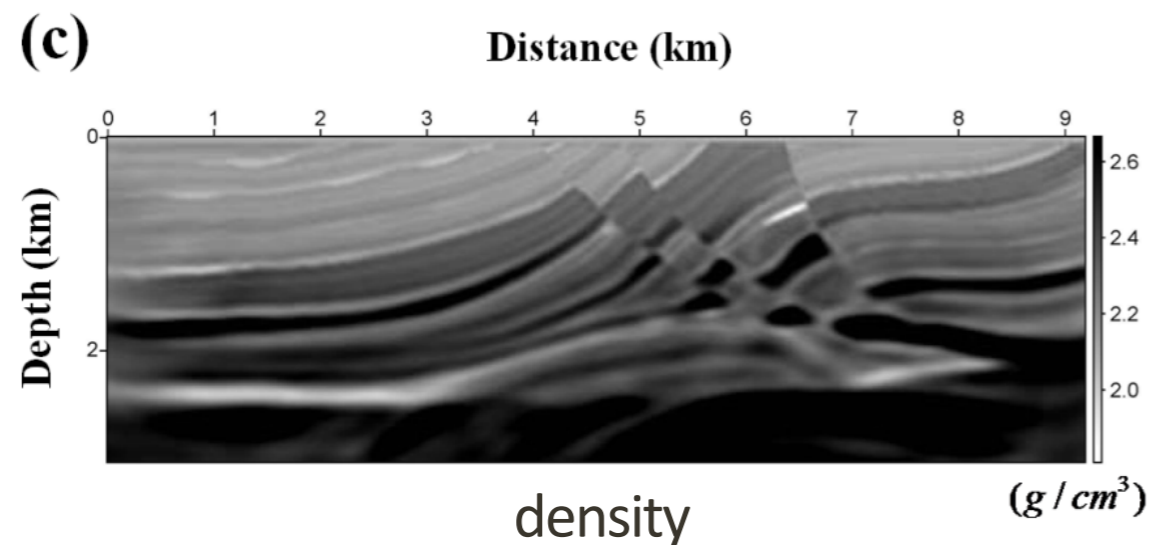
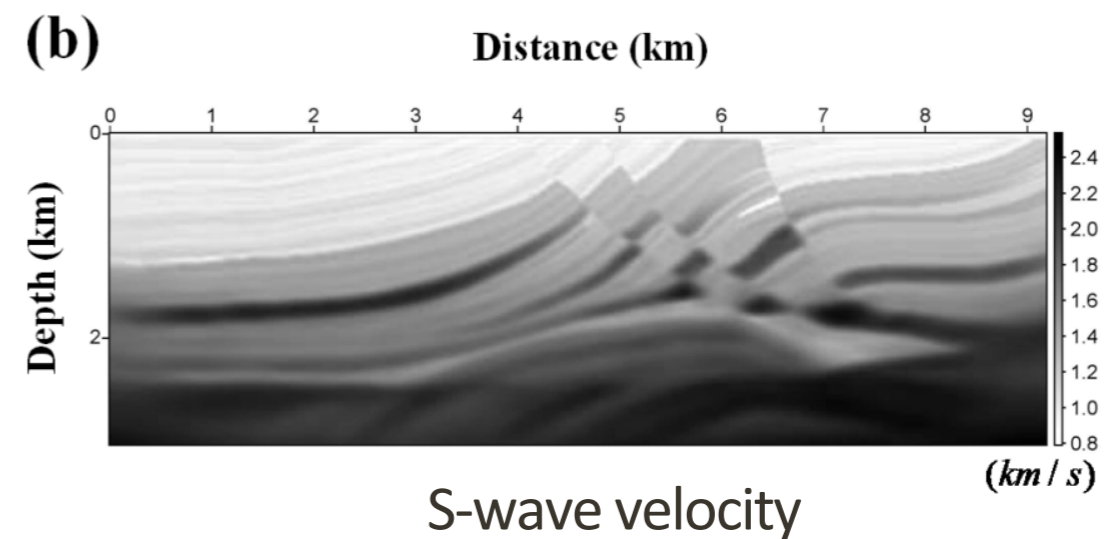
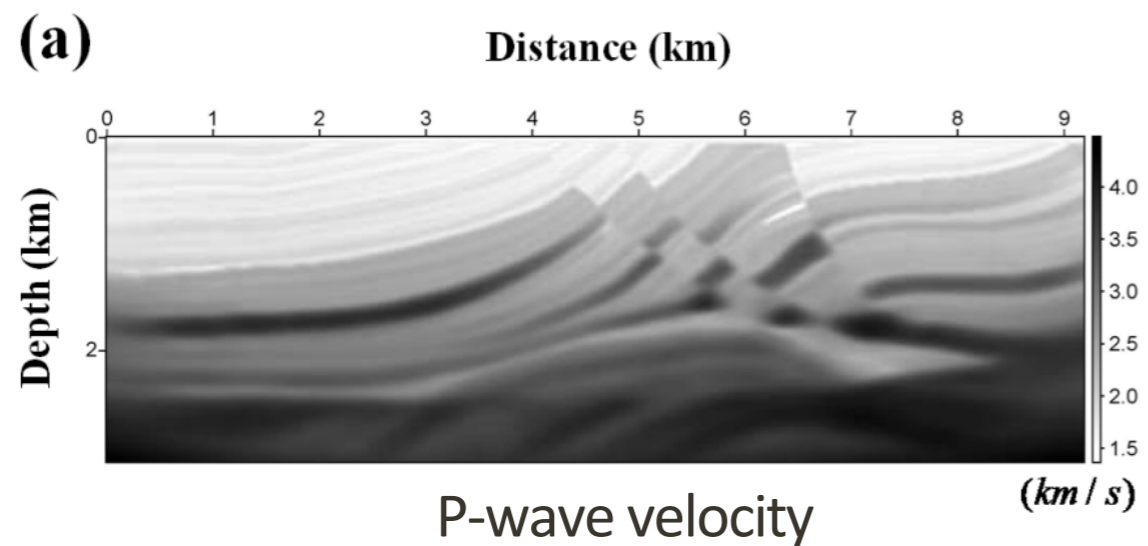
Elastic Marmousi-2 model

● Initial model



Conventional FWI - Examples

● Inversion results - Conventional methods I

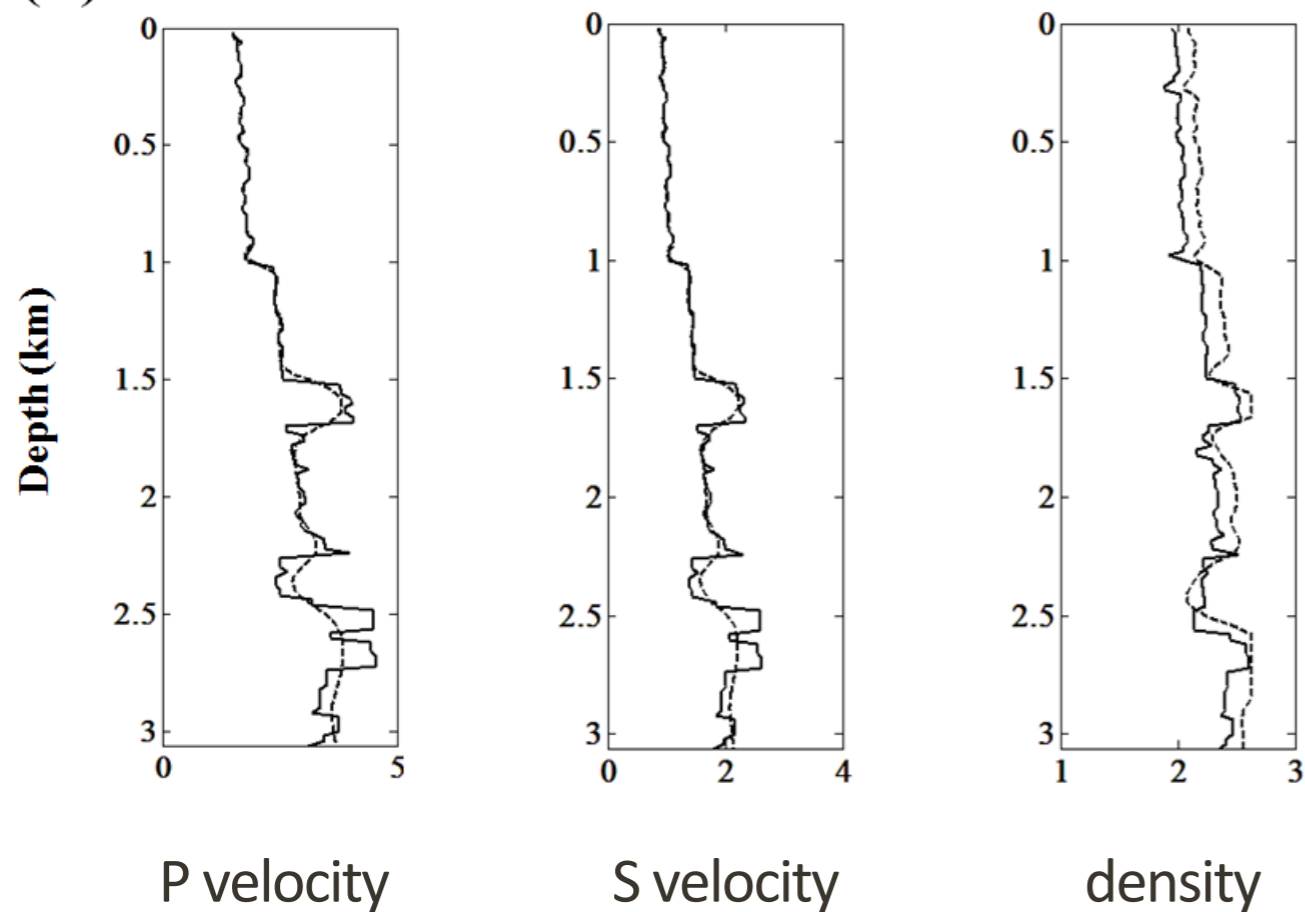


Conventional FWI - Examples

- Depth profiles - Conventional methods I

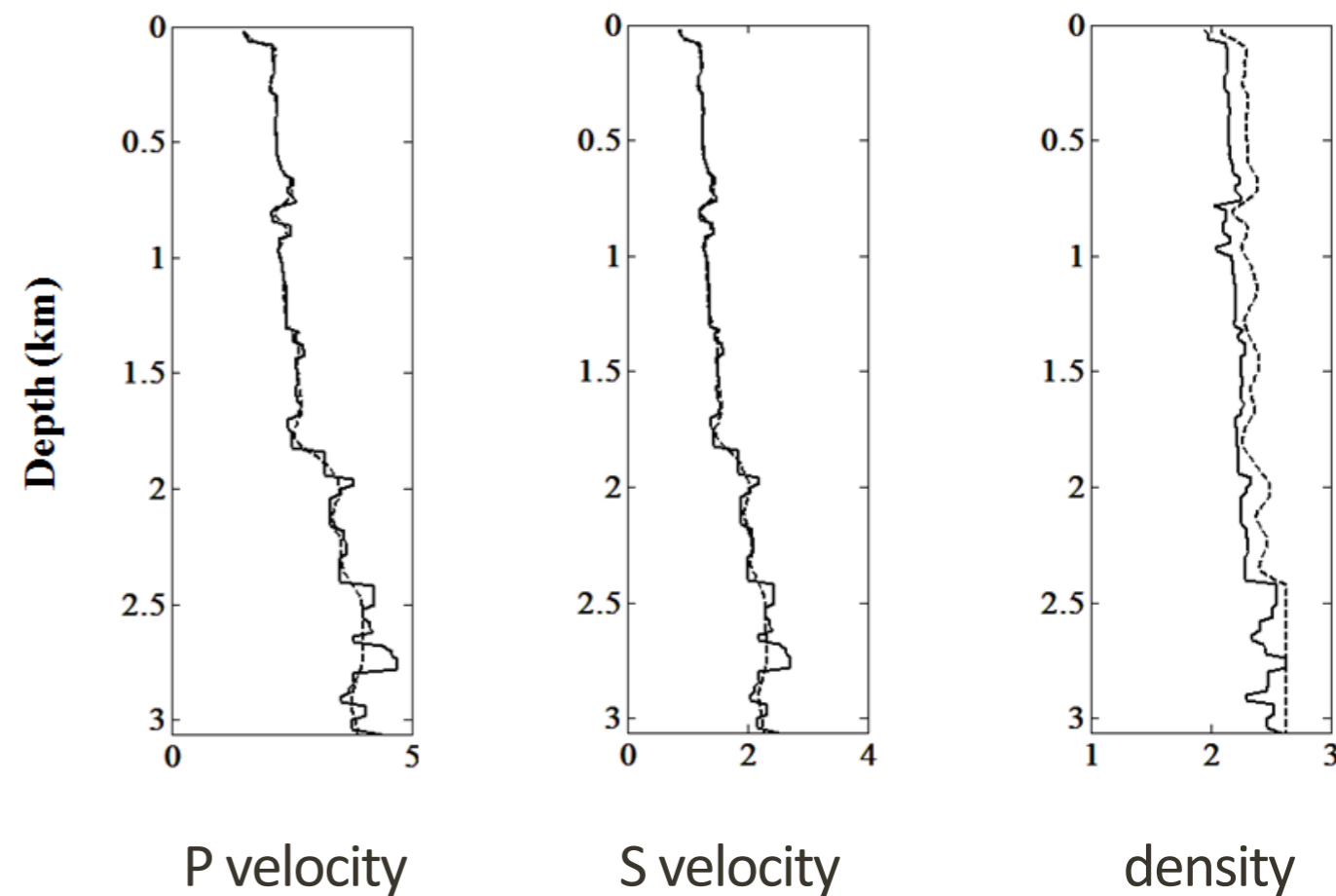
At a distance of 3 km

(a)



At a distance of 6 km

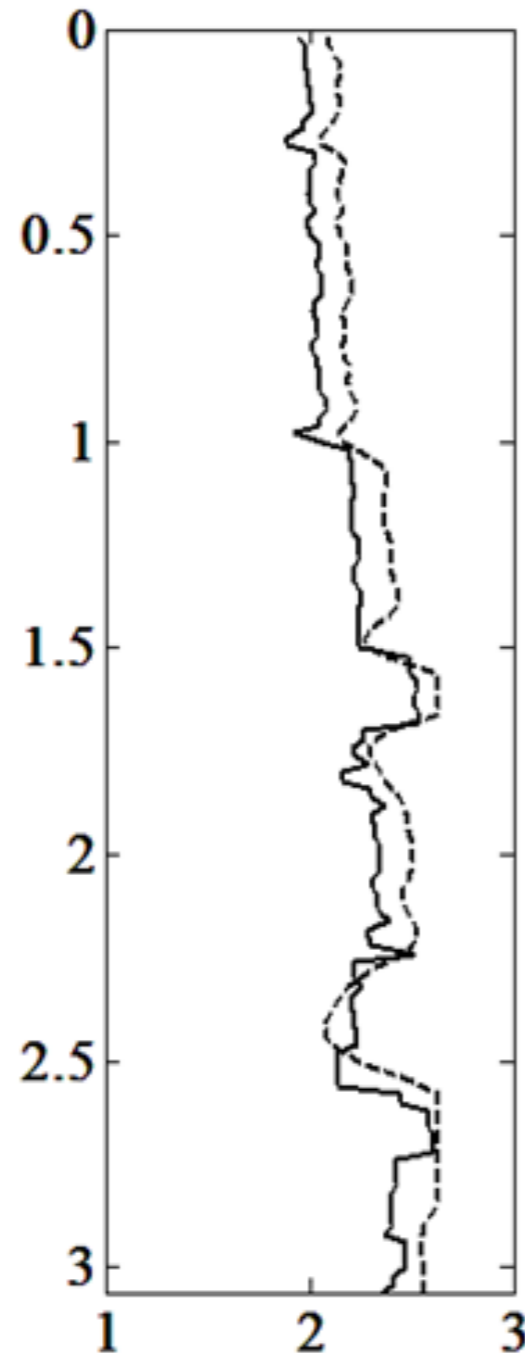
(b)



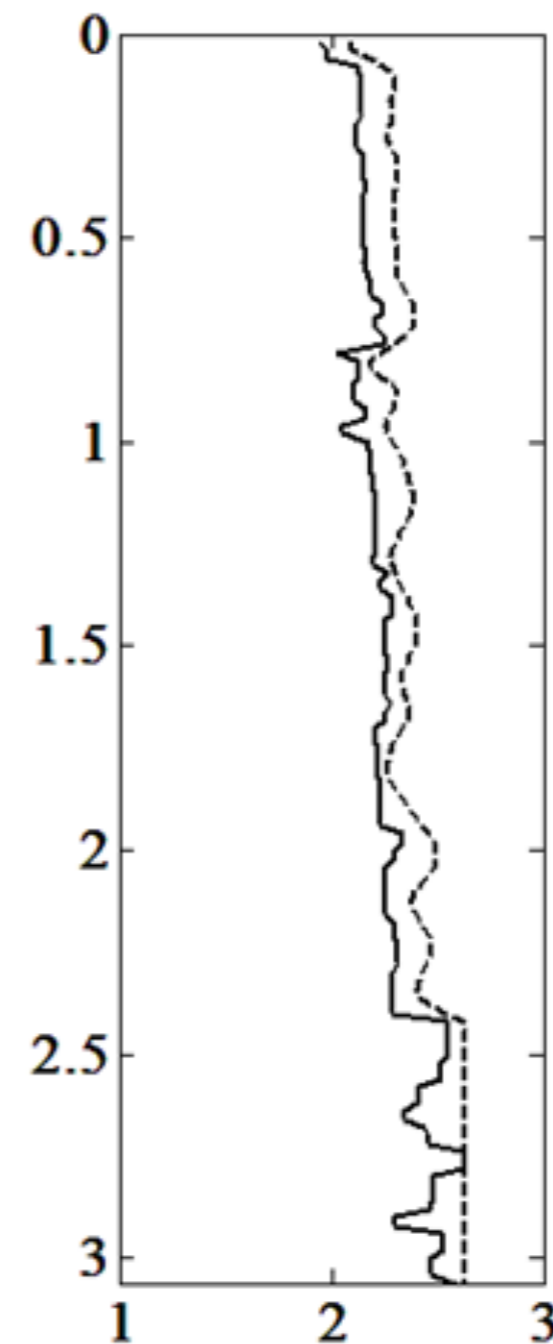
Conventional FWI - Examples

- Density profiles - Conventional methods I

at a distance of 3 km

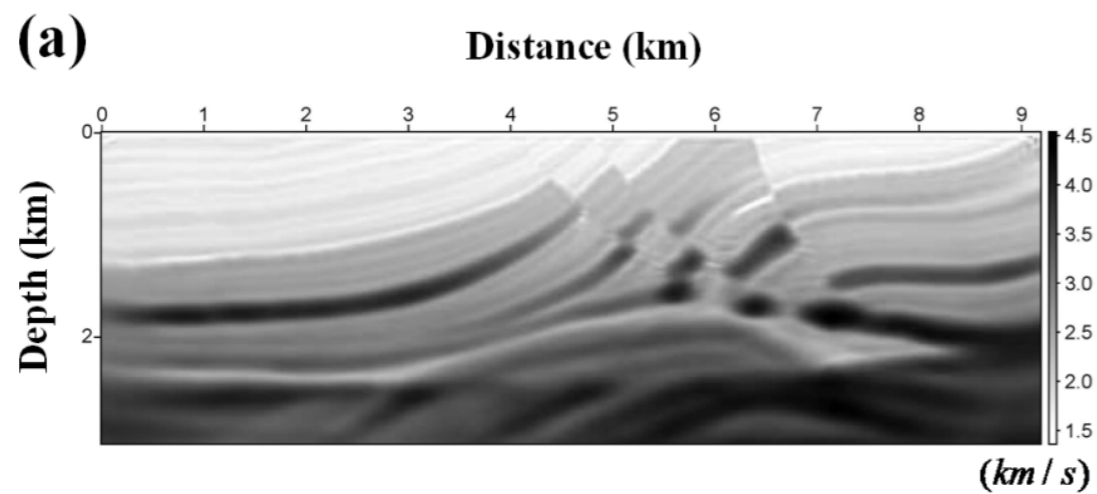


at a distance of 6 km

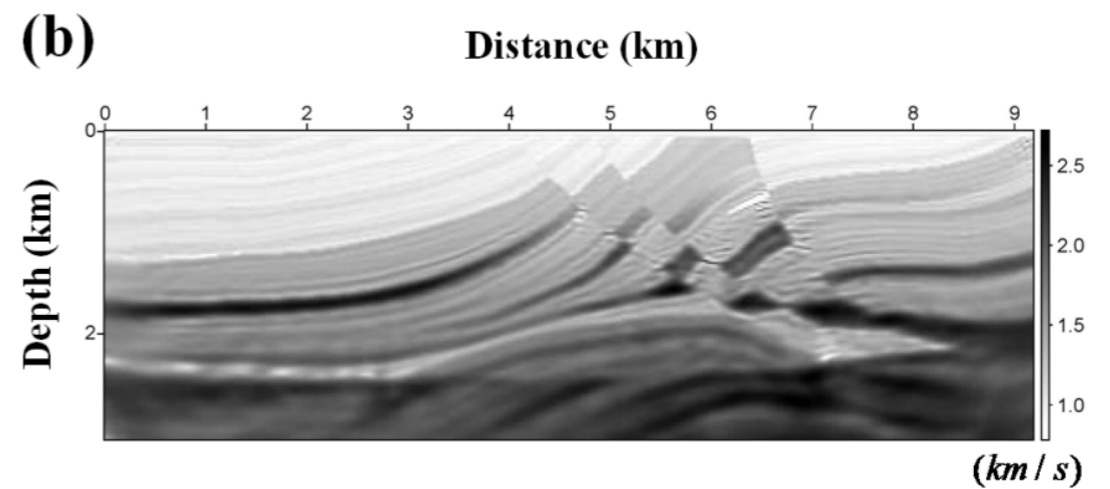


Conventional FWI - Examples

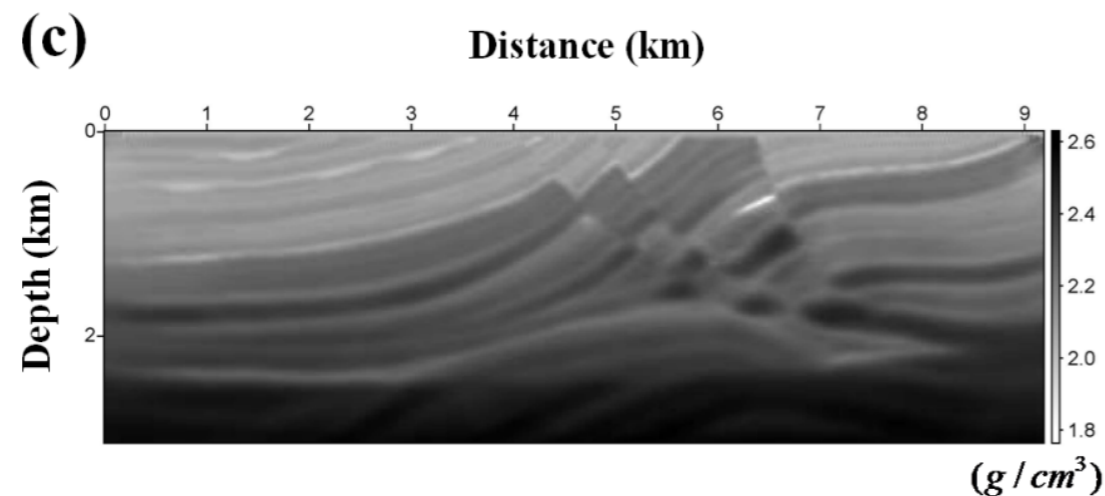
- Inversion results - Conventional methods II



P-wave velocity



S-wave velocity



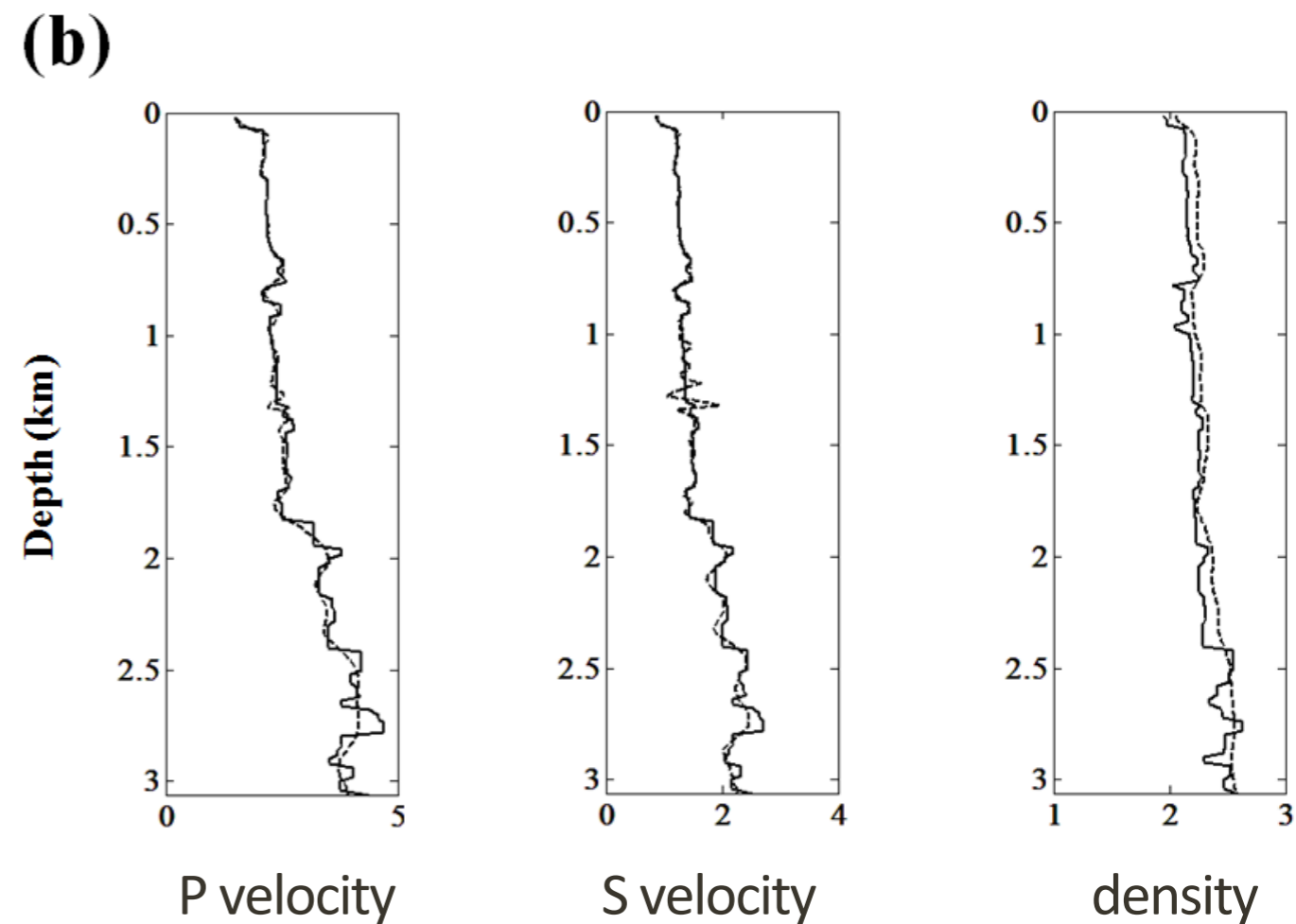
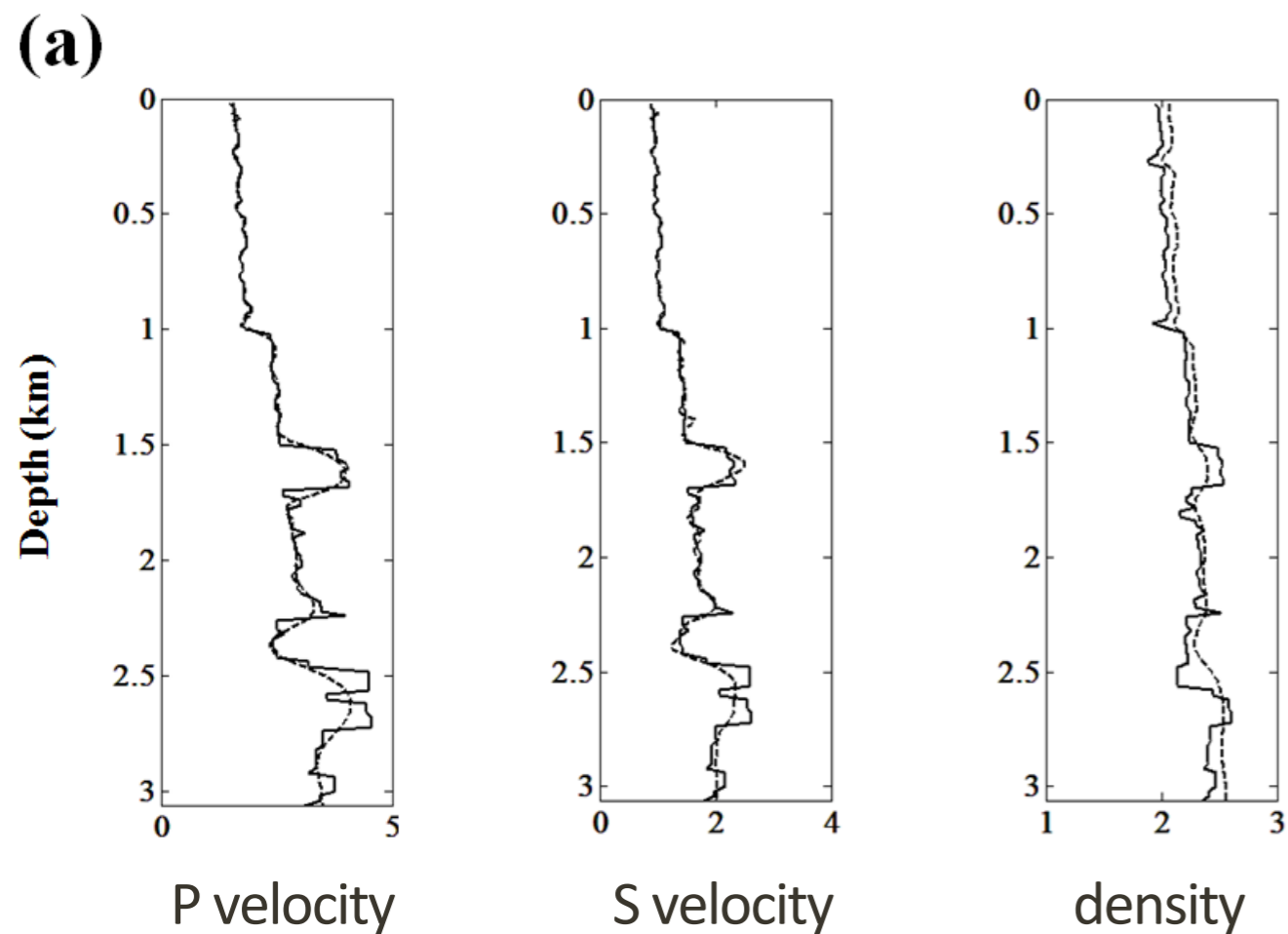
density

Conventional FWI - Examples

- Depth profiles - Conventional methods II

at a distance of 3 km

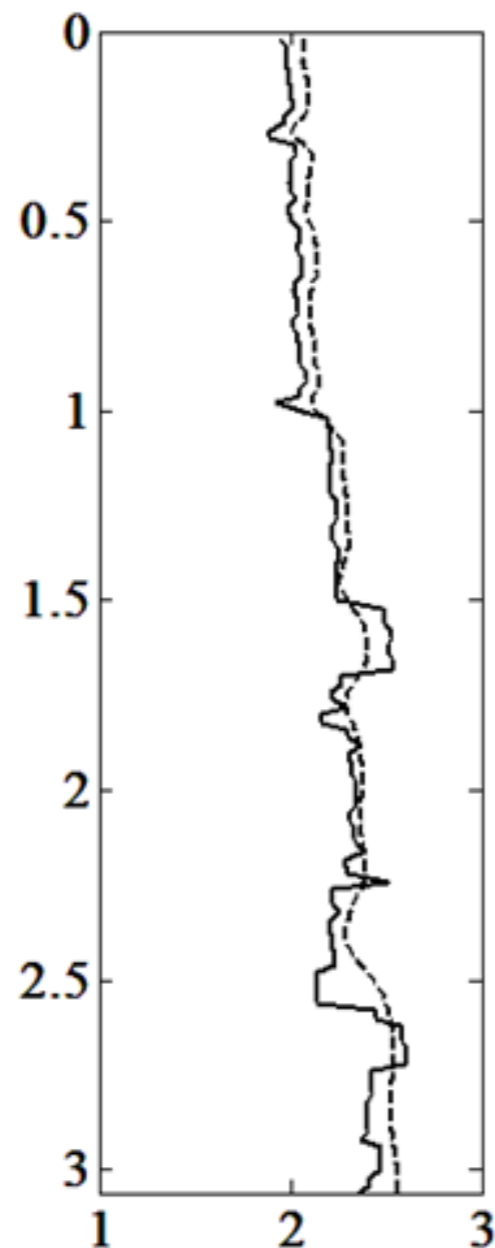
at a distance of 6 km



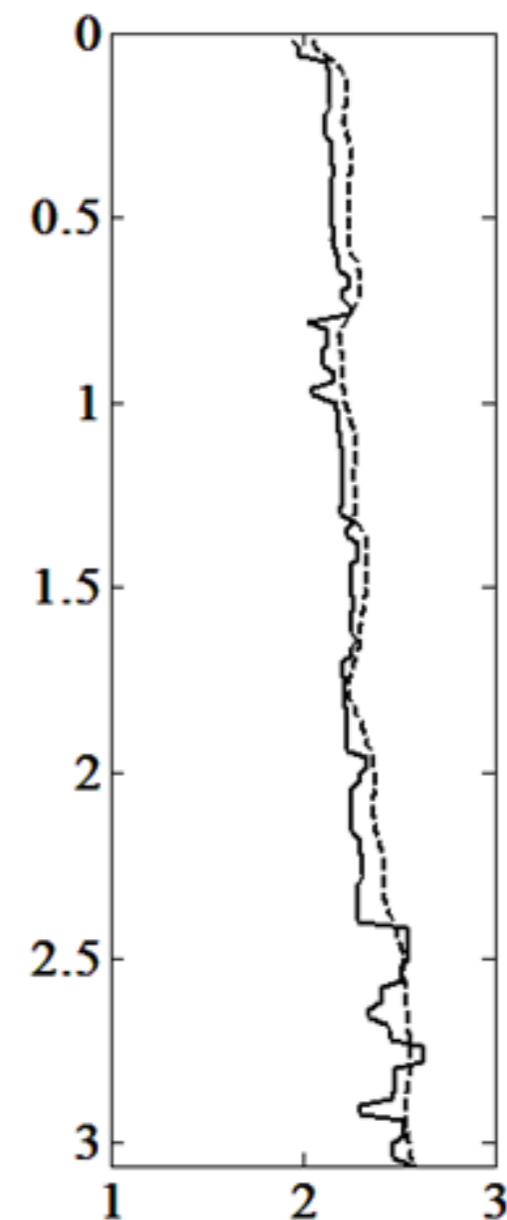
Conventional FWI - Examples

- Density profiles - Conventional methods II

at a distance of 3 km



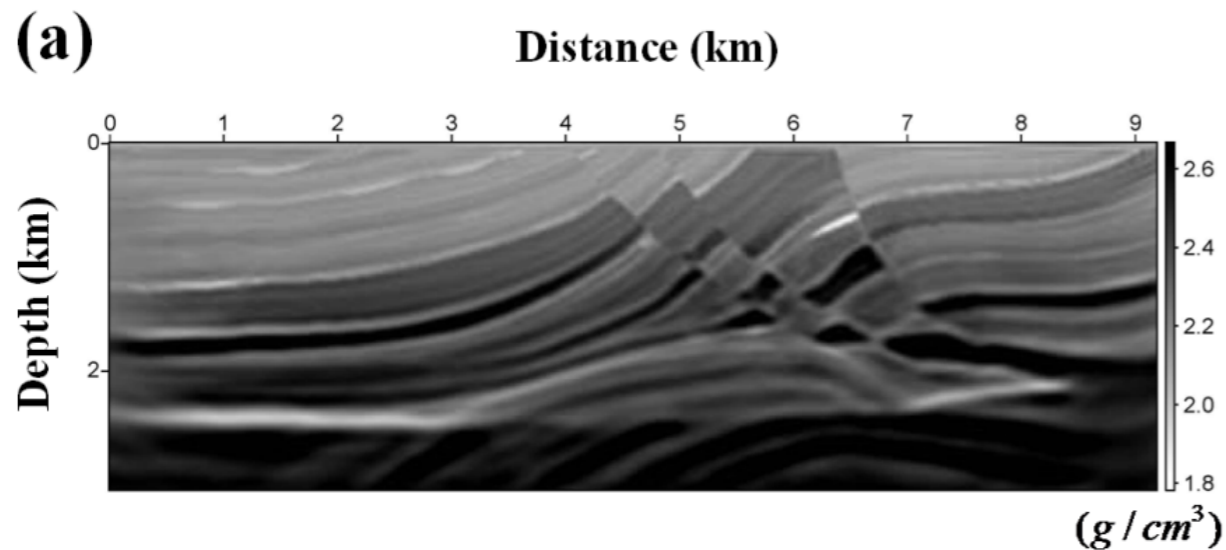
at a distance of 6 km



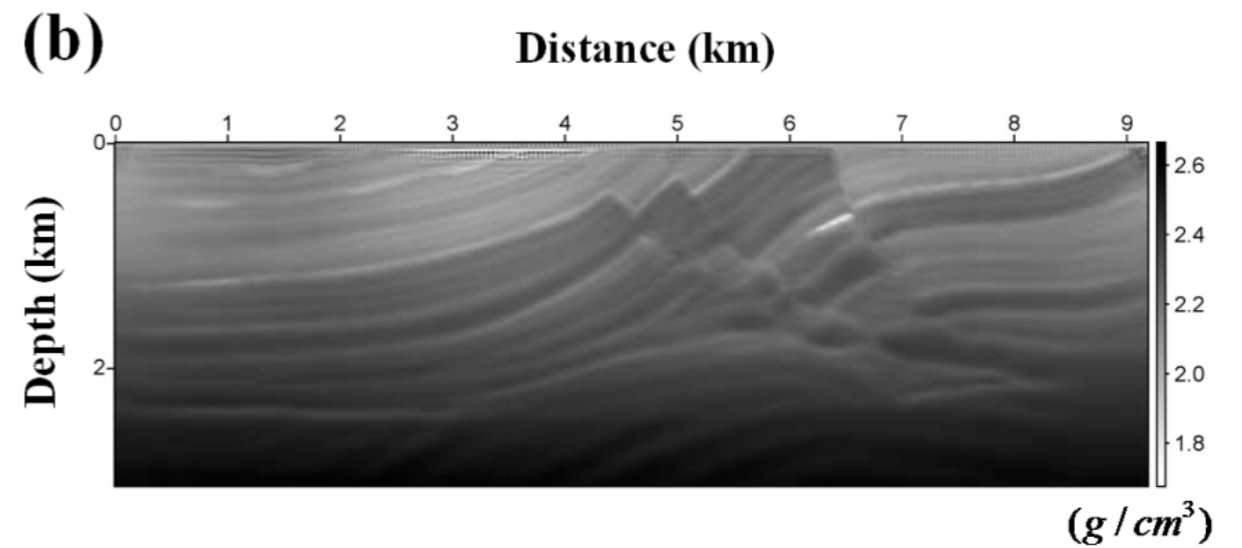
Conventional FWI - Examples

- Density results **in the second stage**

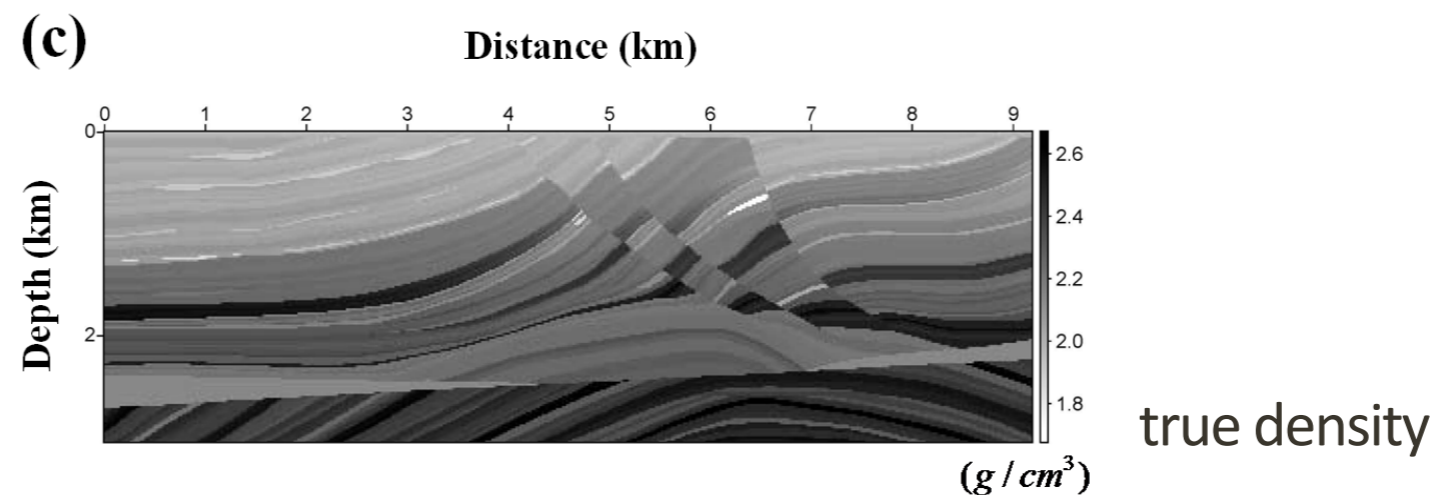
Conventional method I



Conventional method II



The initial guess for density is a gradually increasing model.

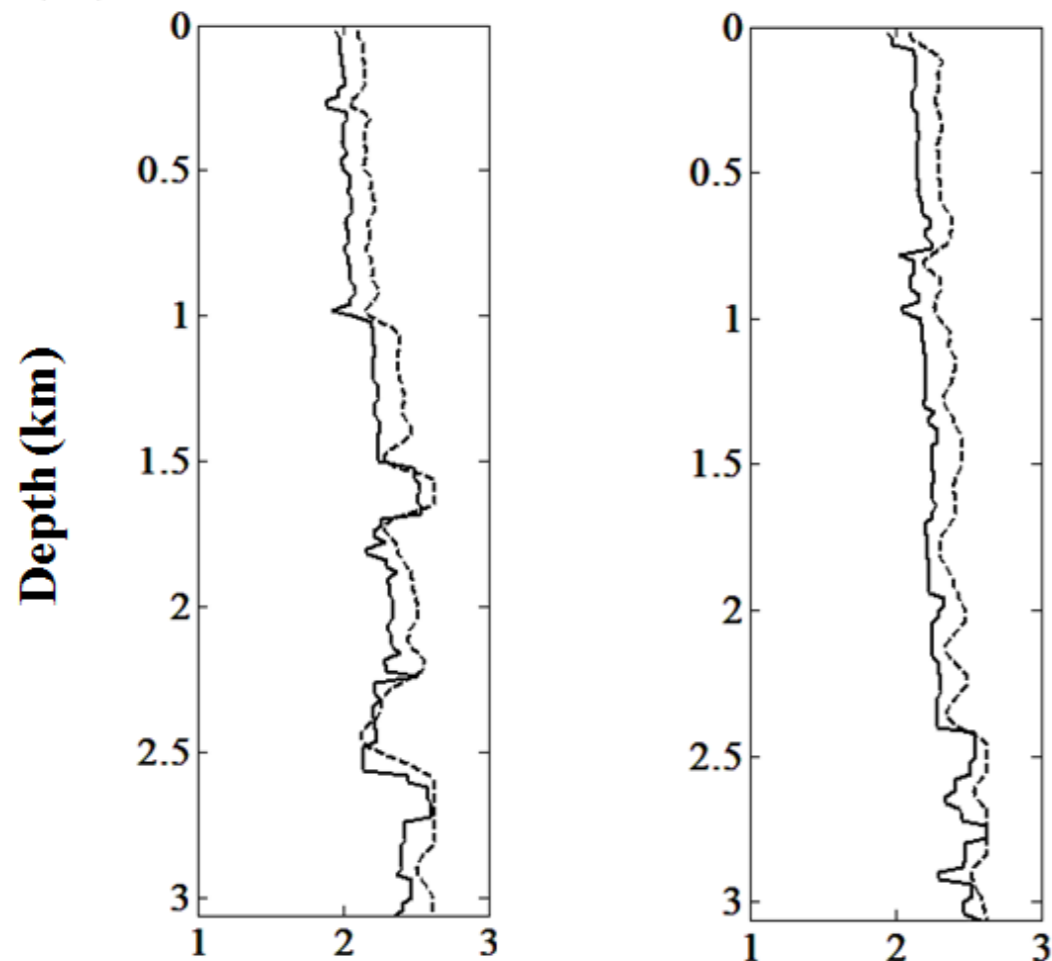


Conventional FWI - Examples

- Density profiles **in the second stage**

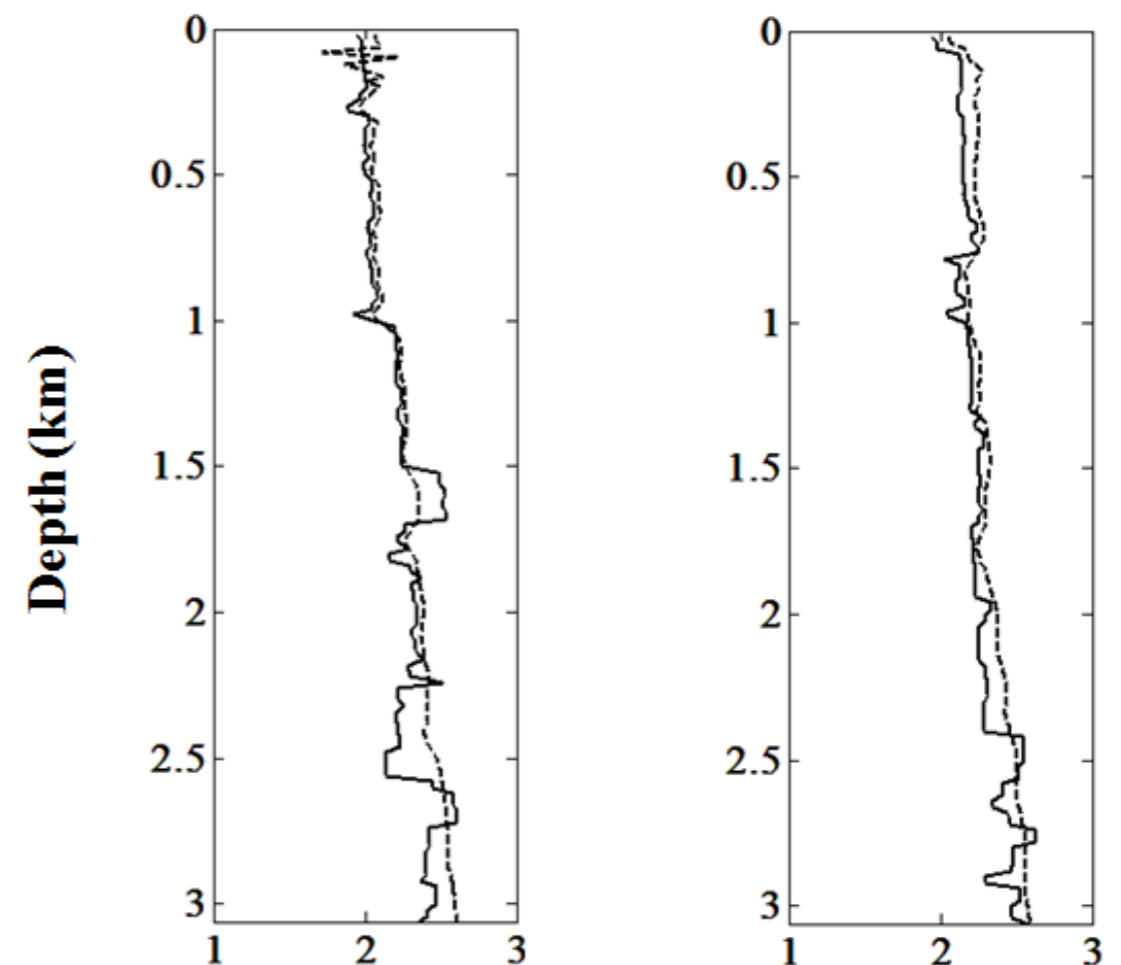
Conventional I

(a)



Conventional II

(b)



Contents

- FWI algorithm
- Conventional FWI & examples
- **Parameter-selection strategy & examples**
- Conclusions

New strategy for density

● Parameter-selection strategy

- consider that velocities are properly recovered but density is distorted in the conventional inversion results.
- inversion is performed over two stages.
 - > first stage: velocities are described
 - > second stage: Lamé constants and density are inverted

New strategy for density

● Parameter-selection strategy

- First stage

--> inversion is conducted for Lamé constants **with density fixed**
based on the wave equations parameterized by Lamé constants
and density

--> **wrong Lamé constants and density**

--> **but reliable velocities can be extracted from wrong information**

$$\alpha = \sqrt{\frac{\lambda_V + 2\mu_V}{\rho_C}} \quad \beta = \sqrt{\frac{\mu_V}{\rho_C}} \quad V: \text{virtual, } C: \text{constant}$$

- Virtual sources

$$\left(\mathbf{f}^v\right)_\lambda = -\frac{\partial \mathbf{S}}{\partial \lambda} \tilde{\mathbf{u}}, \quad \left(\mathbf{f}^v\right)_\mu = -\frac{\partial \mathbf{S}}{\partial \mu} \tilde{\mathbf{u}},$$

New strategy for density

● Parameter-selection strategy

- Second stage

- > velocities obtained in the first stage are used as initial guesses
- > for initial guess for density, a linearly increasing model is used
- > both Lamé constants and density are inverted based on the wave equation parameterized by velocities and density

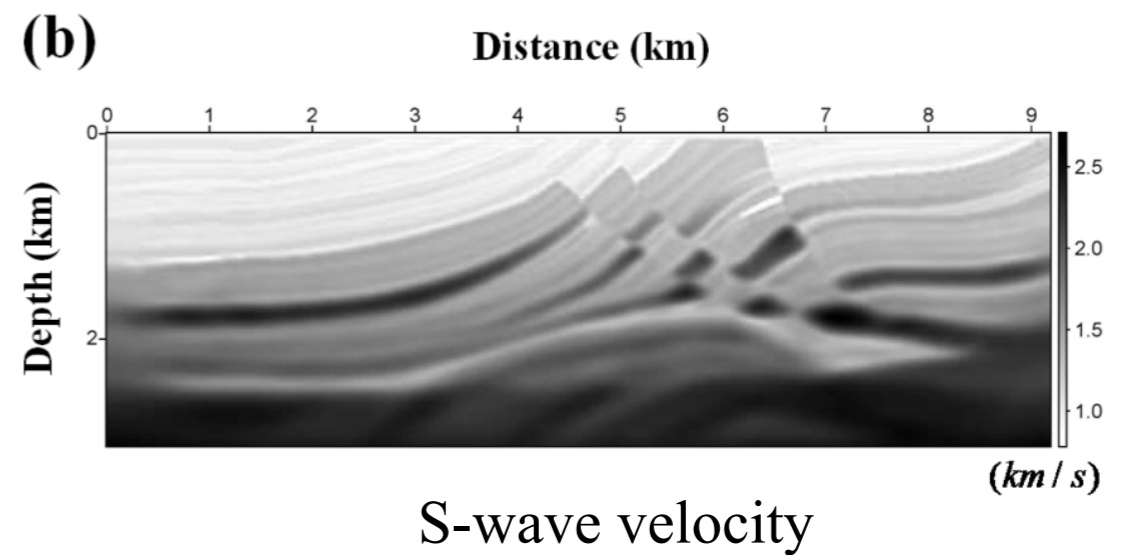
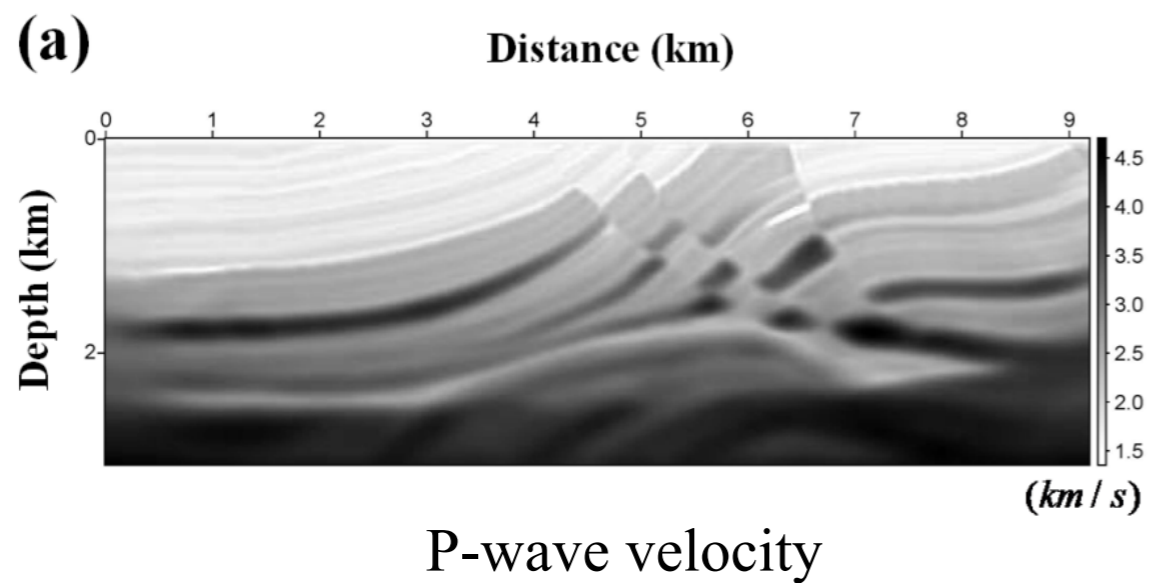
- Virtual sources (using the chain rule reversed to [Mora \(1987\)](#))

$$\left(\mathbf{f}^v\right)_\lambda = -\left(\frac{\partial \mathbf{S}}{\partial \alpha} \frac{\partial \alpha}{\partial \lambda} + \frac{\partial \mathbf{S}}{\partial \beta} \frac{\partial \beta}{\partial \lambda} + \frac{\partial \mathbf{S}}{\partial \rho} \frac{\partial \rho}{\partial \lambda}\right) \tilde{\mathbf{u}} \quad \left(\mathbf{f}^v\right)_\rho = -\frac{\partial \mathbf{S}}{\partial \rho} \tilde{\mathbf{u}}$$

$$\left(\mathbf{f}^v\right)_\mu = -\left(\frac{\partial \mathbf{S}}{\partial \alpha} \frac{\partial \alpha}{\partial \mu} + \frac{\partial \mathbf{S}}{\partial \beta} \frac{\partial \beta}{\partial \mu} + \frac{\partial \mathbf{S}}{\partial \rho} \frac{\partial \rho}{\partial \mu}\right) \tilde{\mathbf{u}}$$

New inversion results

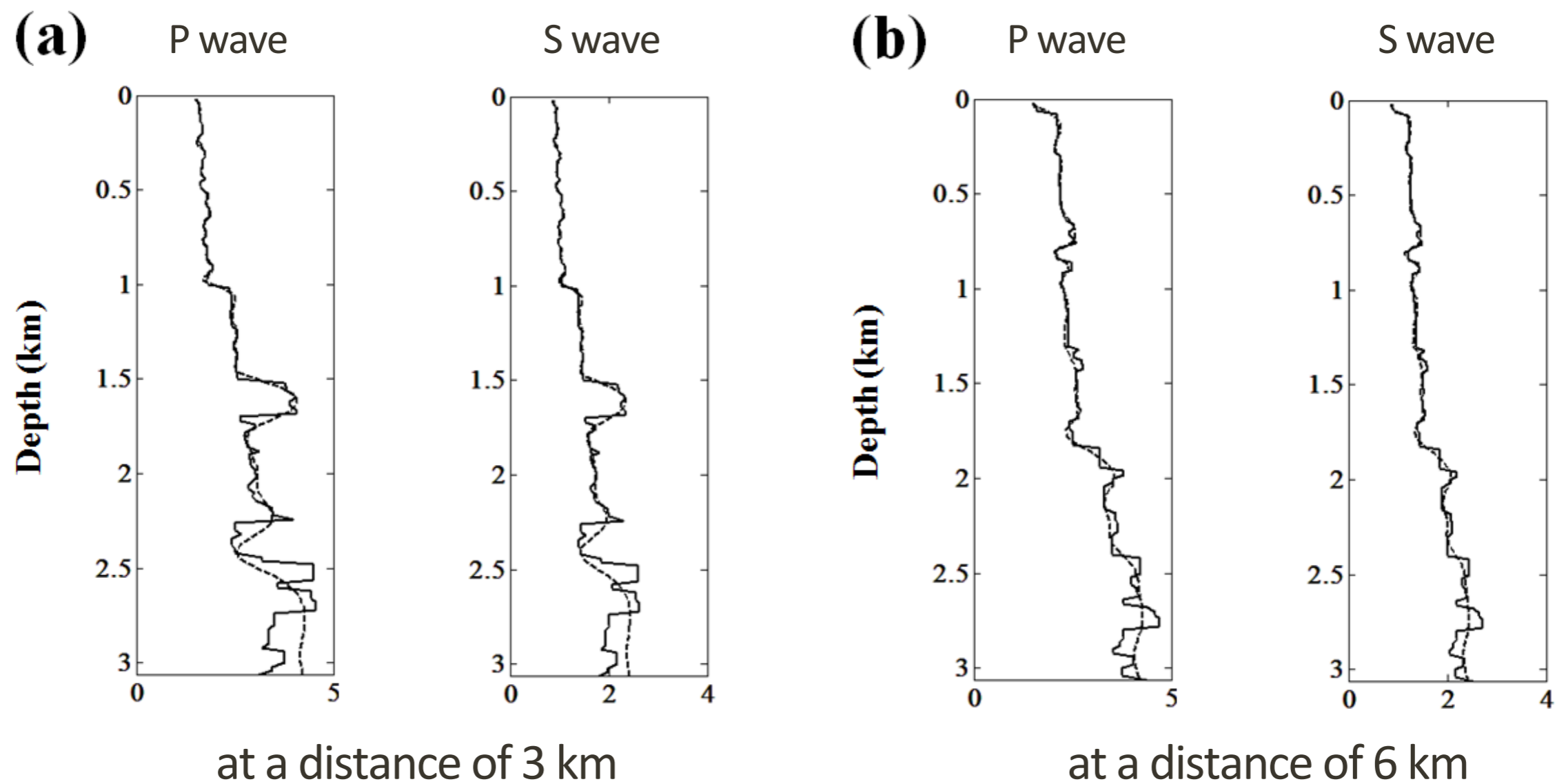
- Velocities inverted in the first stage



The density is fixed as 2 g/cm^3

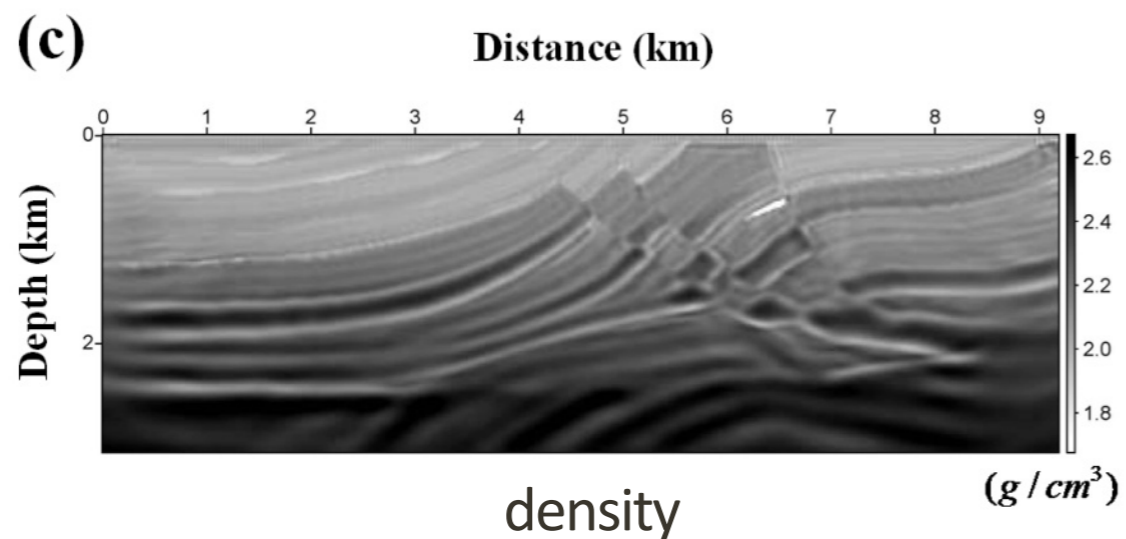
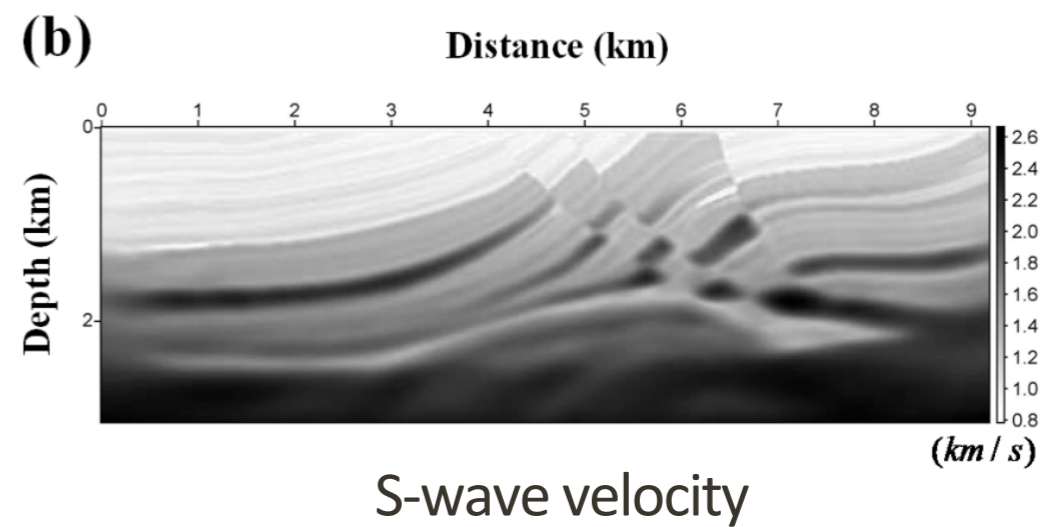
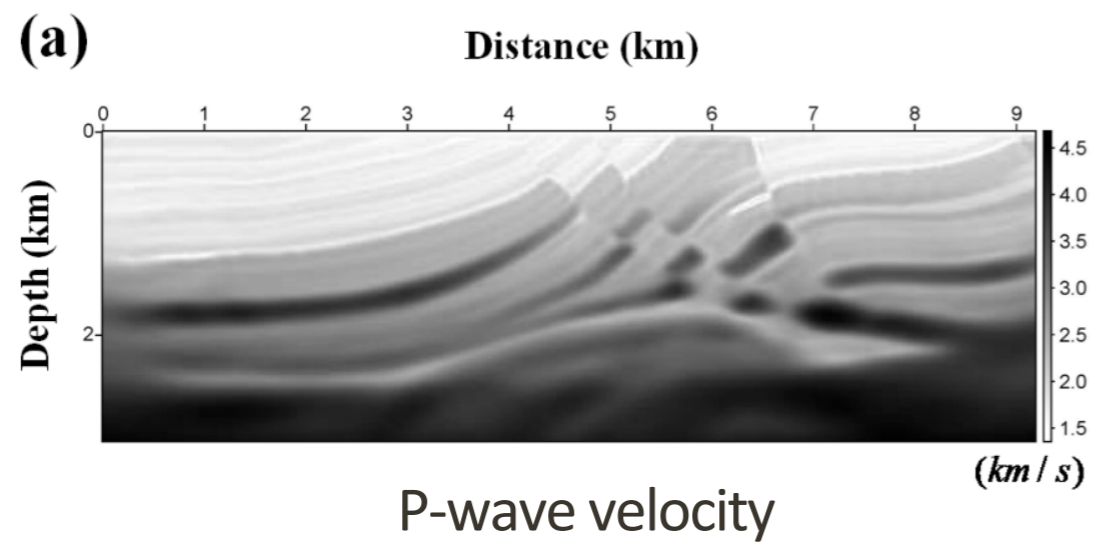
New inversion results

- Velocity profiles **in the first stage**



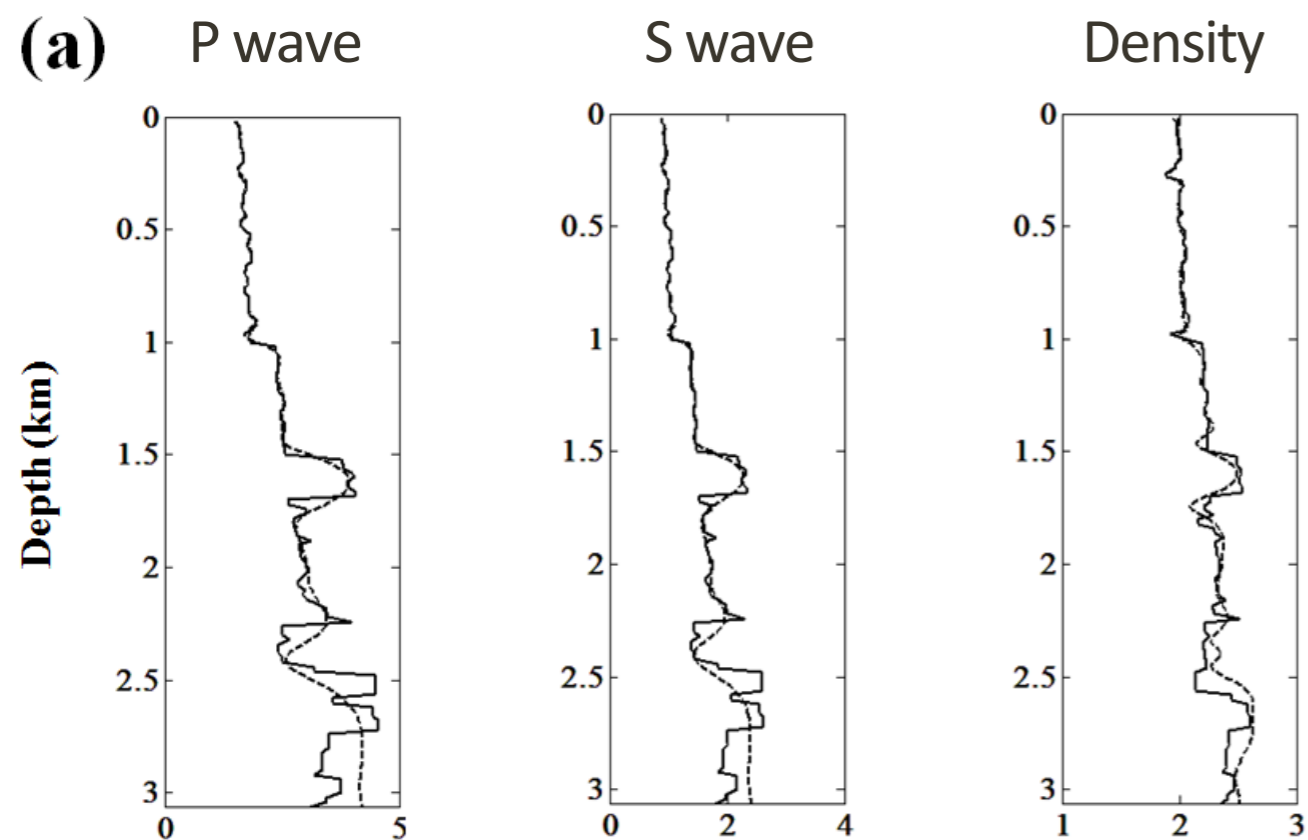
New inversion results

- Inversion results in the second stage

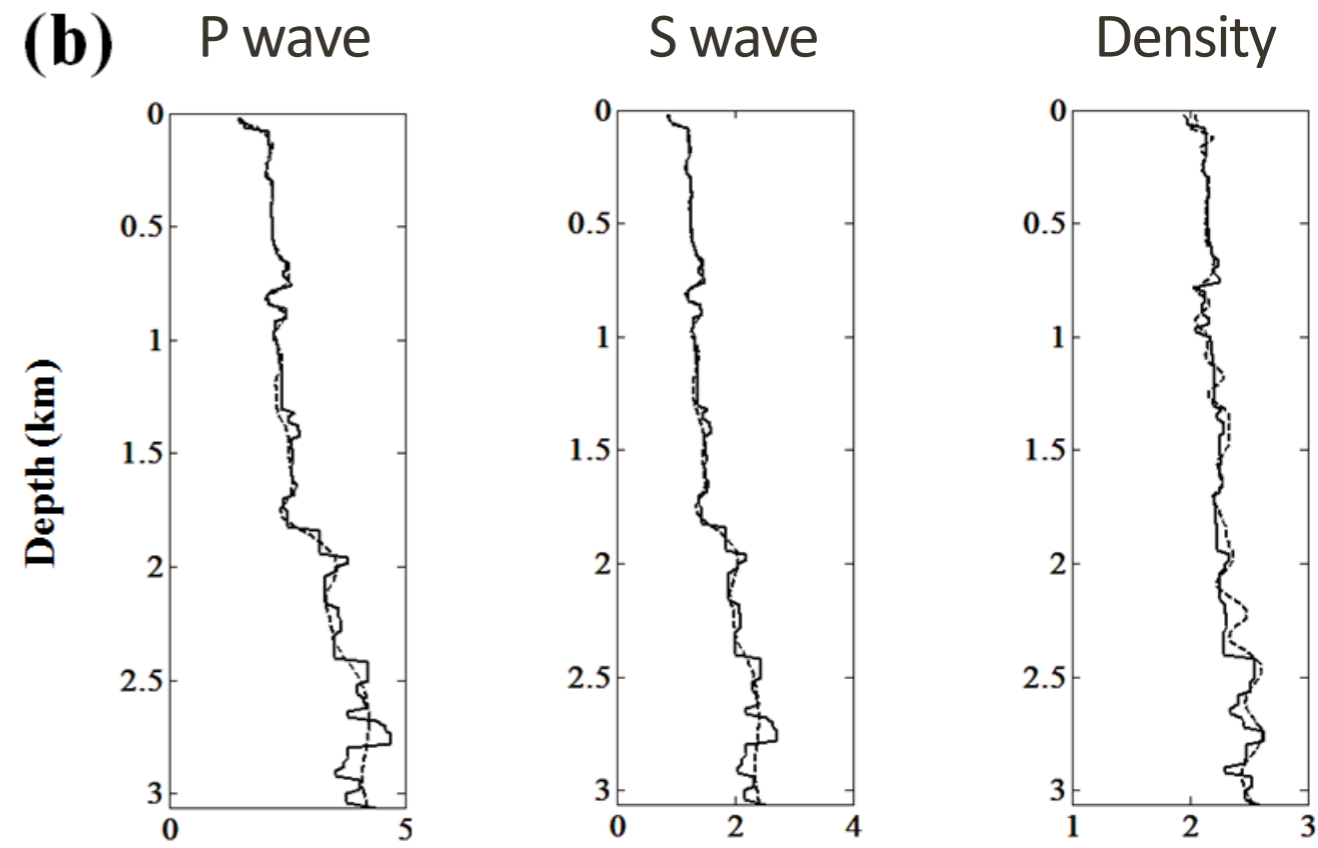


New inversion results

- Depth profiles in the second stage



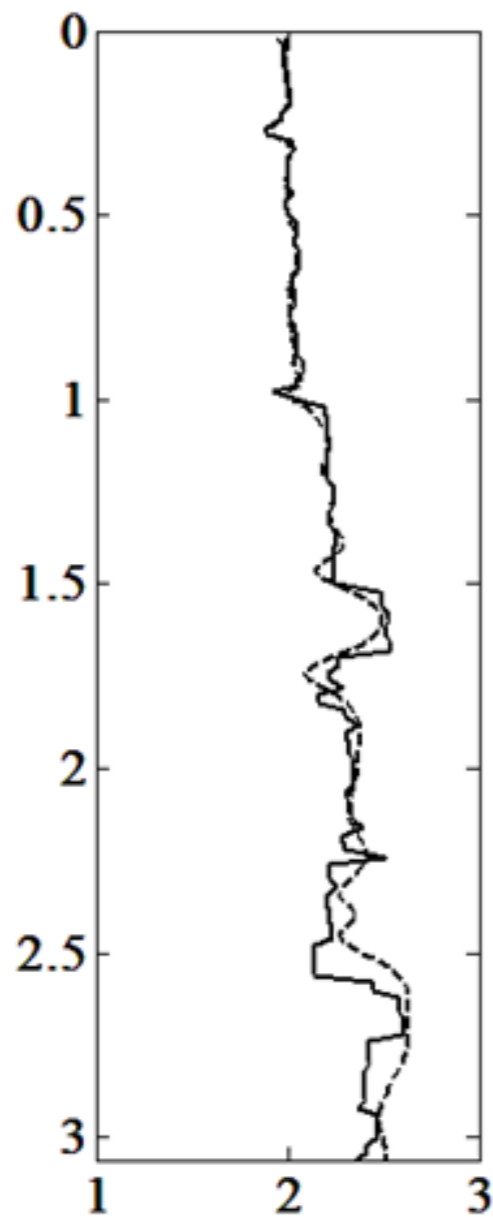
at a distance of 3 km



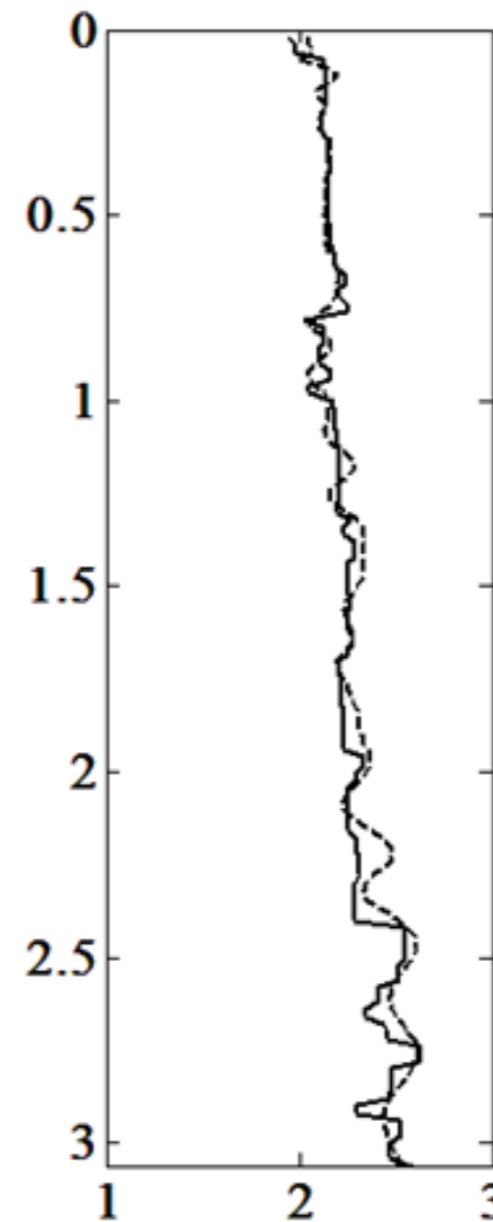
at a distance of 6 km

New inversion results

- Density profiles **in the second stage**



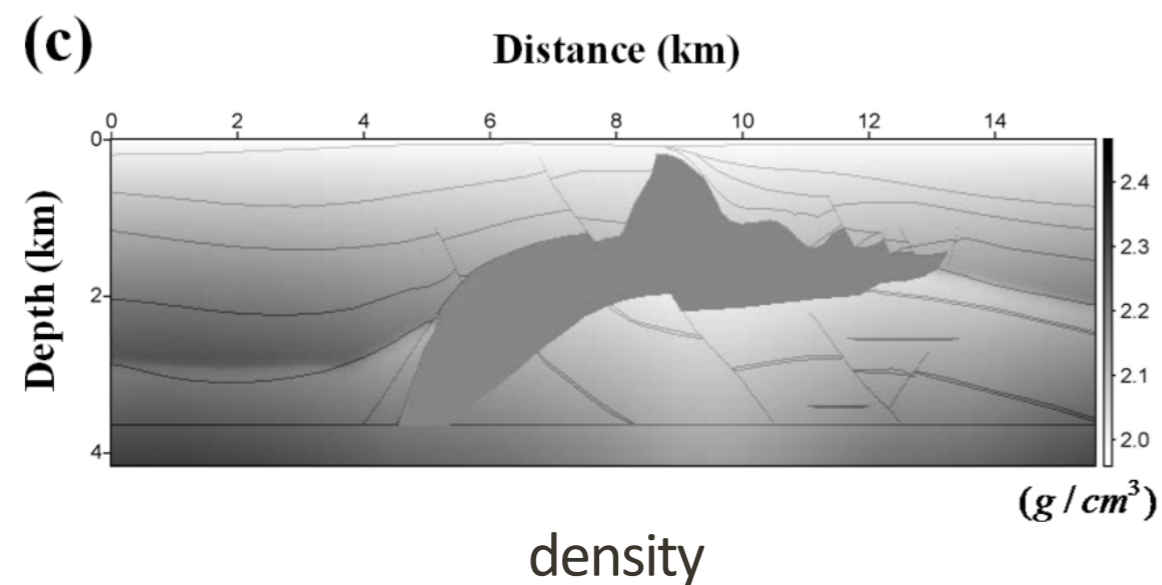
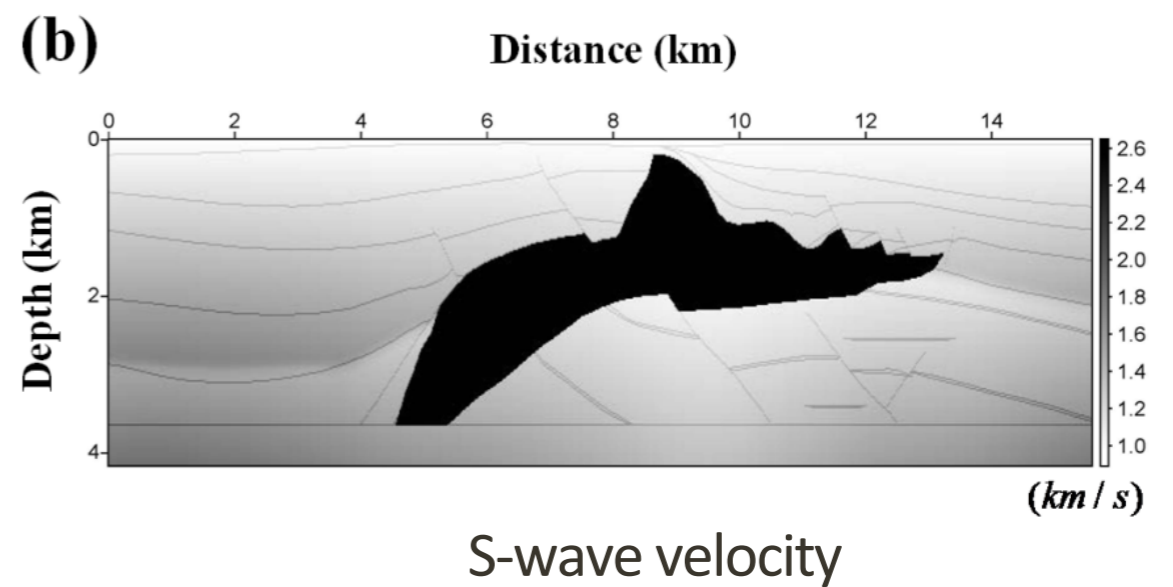
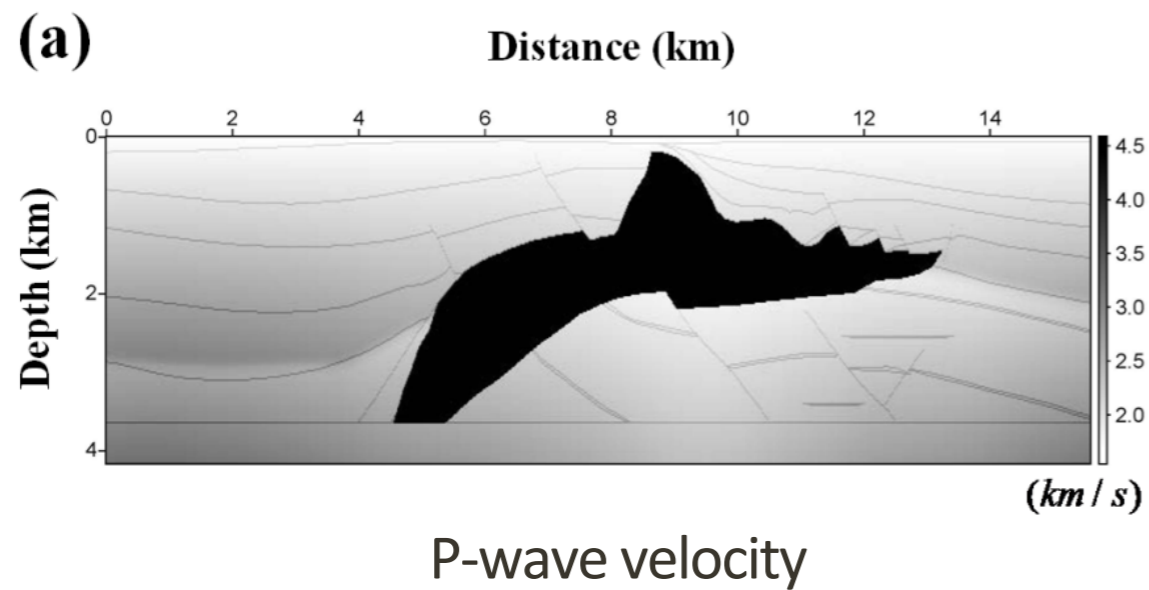
at a distance of 3 km



at a distance of 6 km

SEG/EAGE salt model

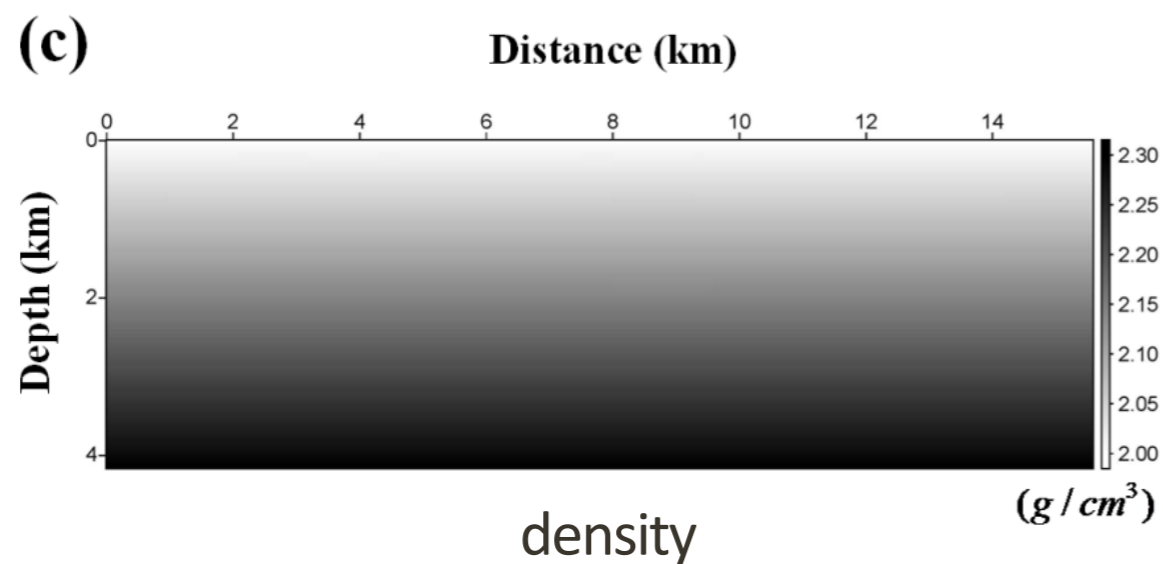
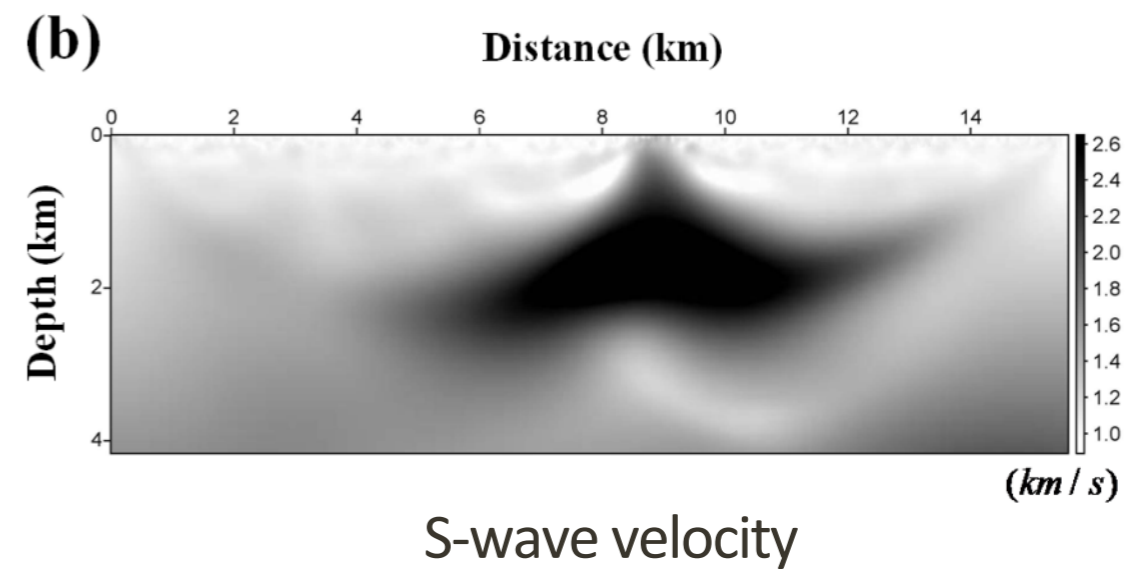
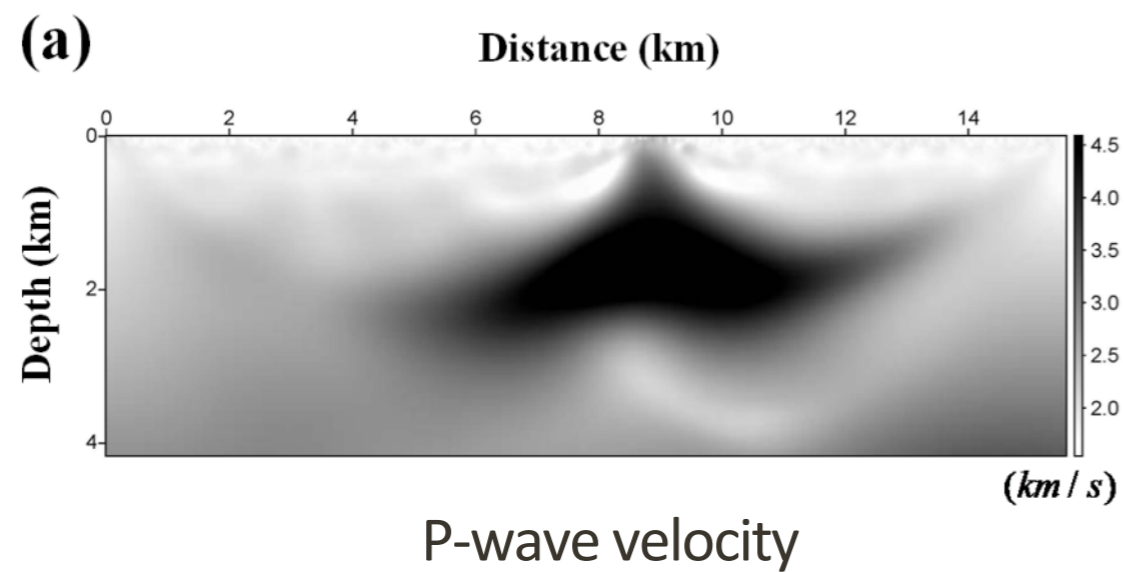
● True model



Parameters	Values
dimension	15.6 km x 4.2 km
no. of source	379
source interval	0.04 km
no. of receiver	781
receiver interval	0.02 km
recording time	6 s
frequency range	0.167 - 10 Hz

SEG/EAGE salt model

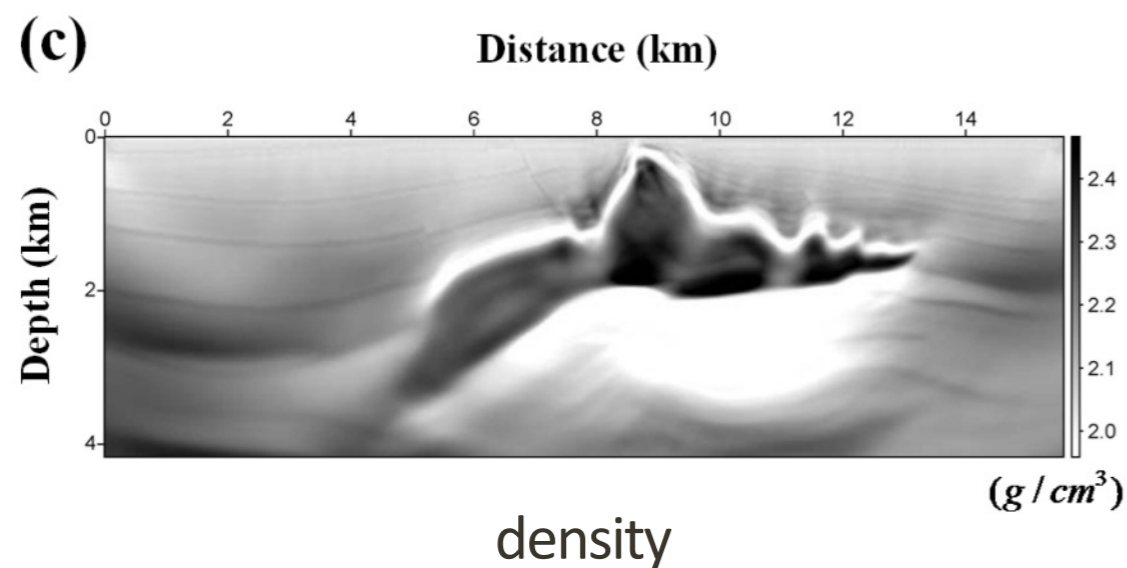
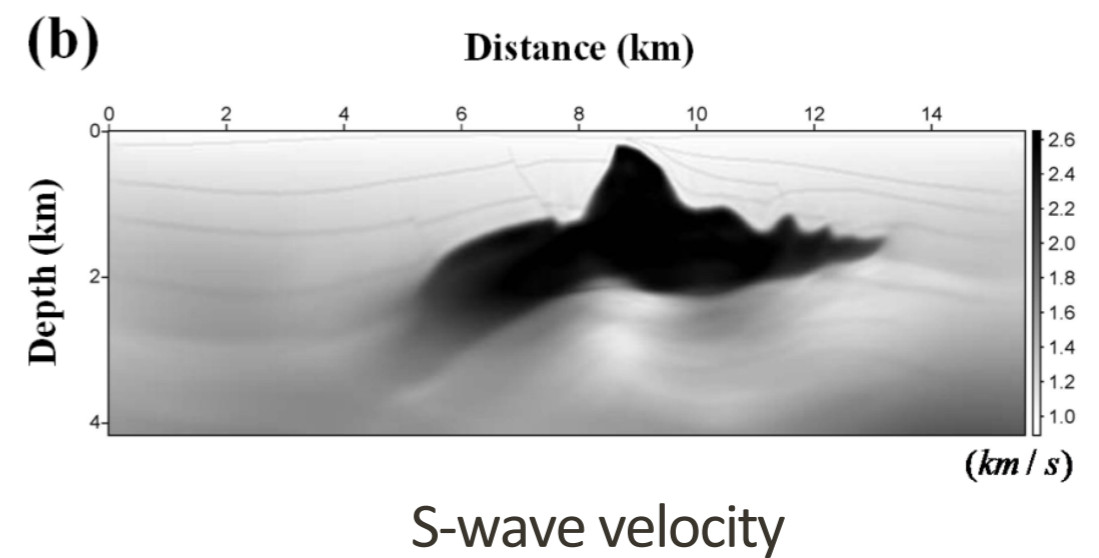
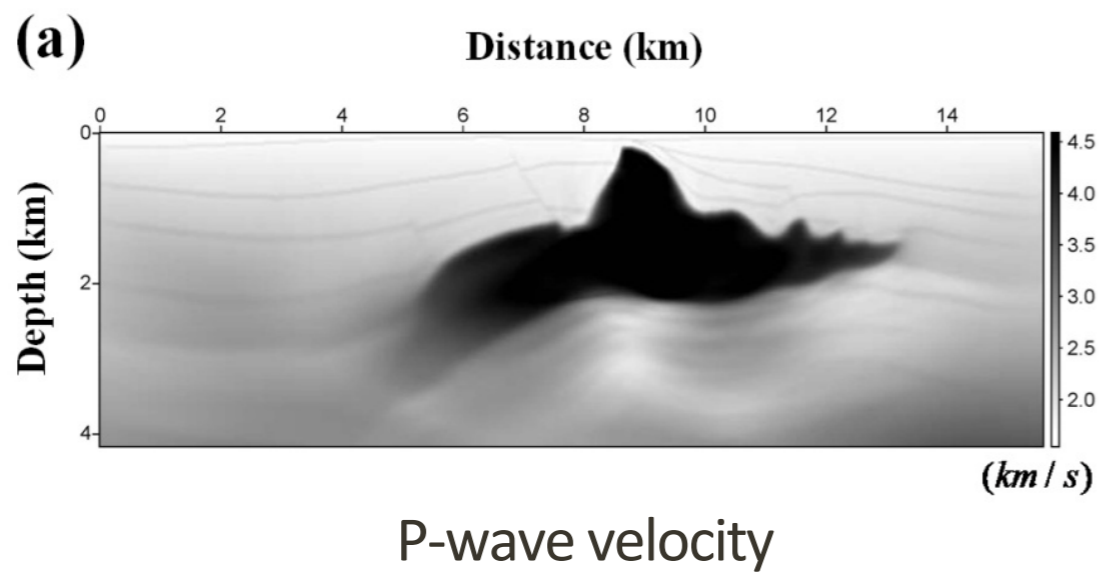
● Initial model



Laplace-domain inversion results
(Chung et al. 2010)

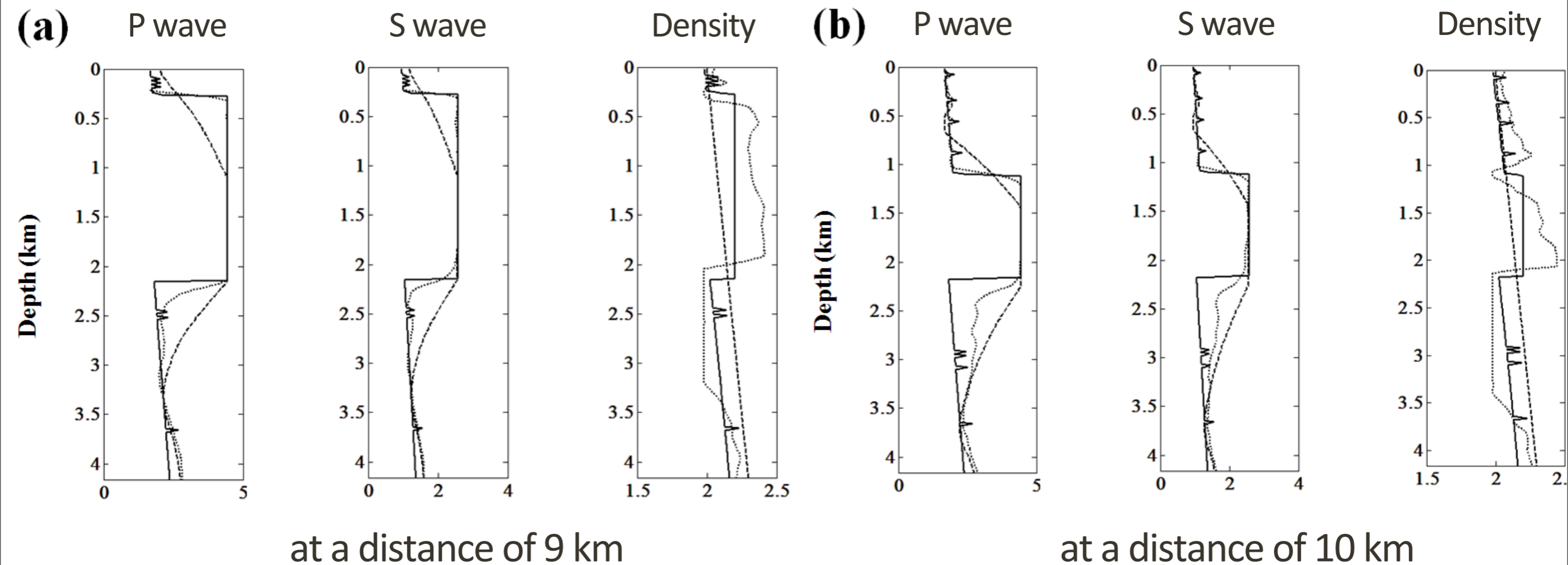
Conventional inversion results

● Inversion results - Conventional method I



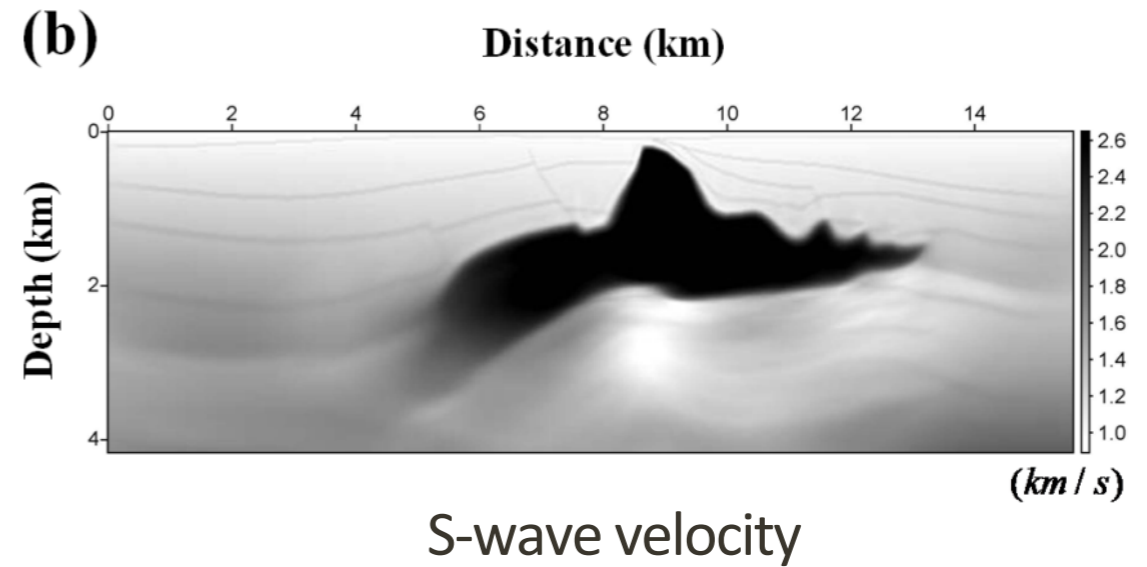
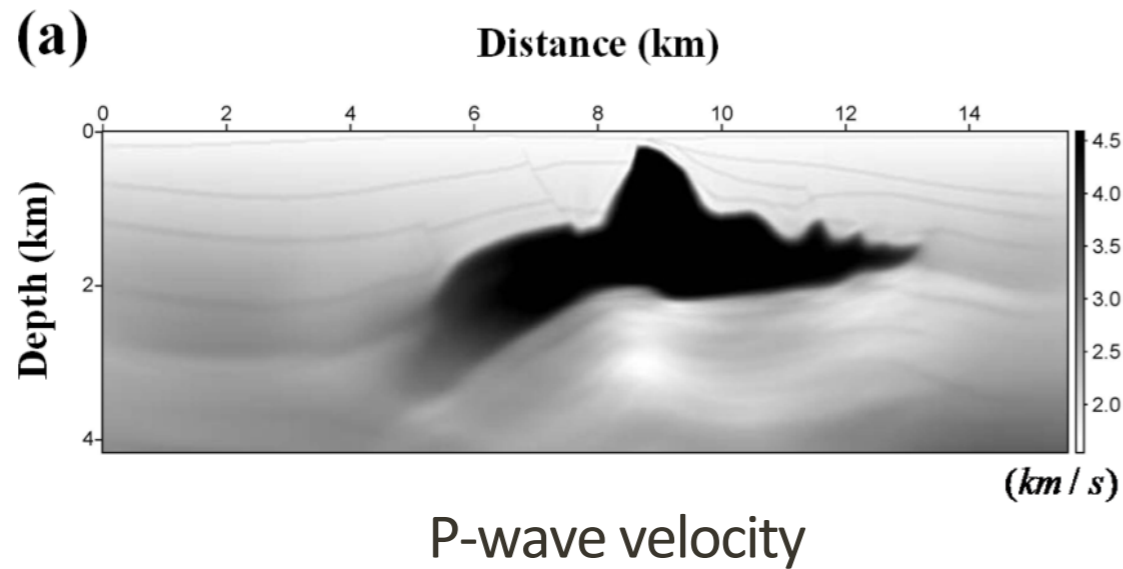
Conventional inversion results

- Depth profiles - Conventional method I



New inversion results

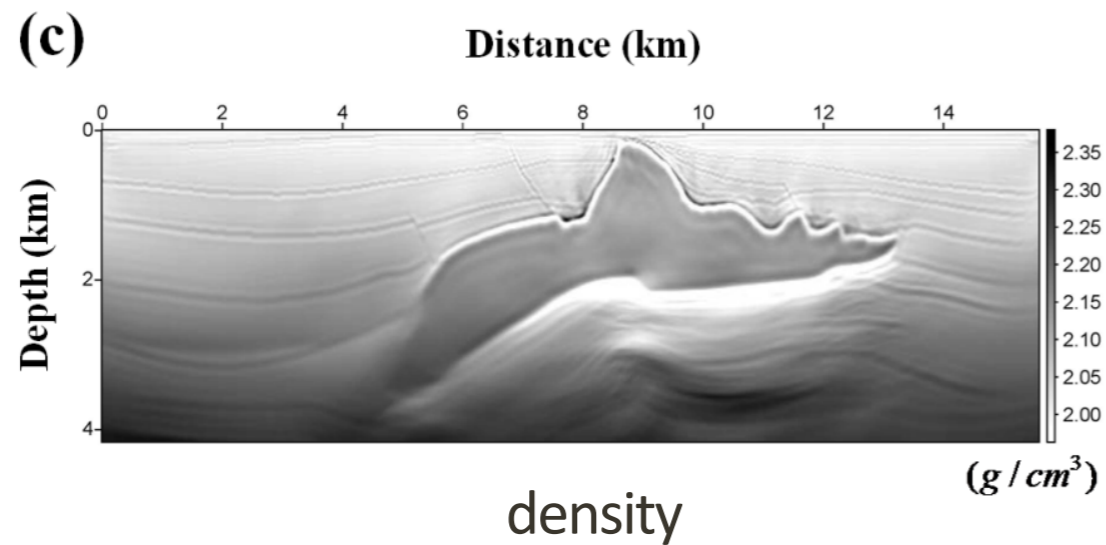
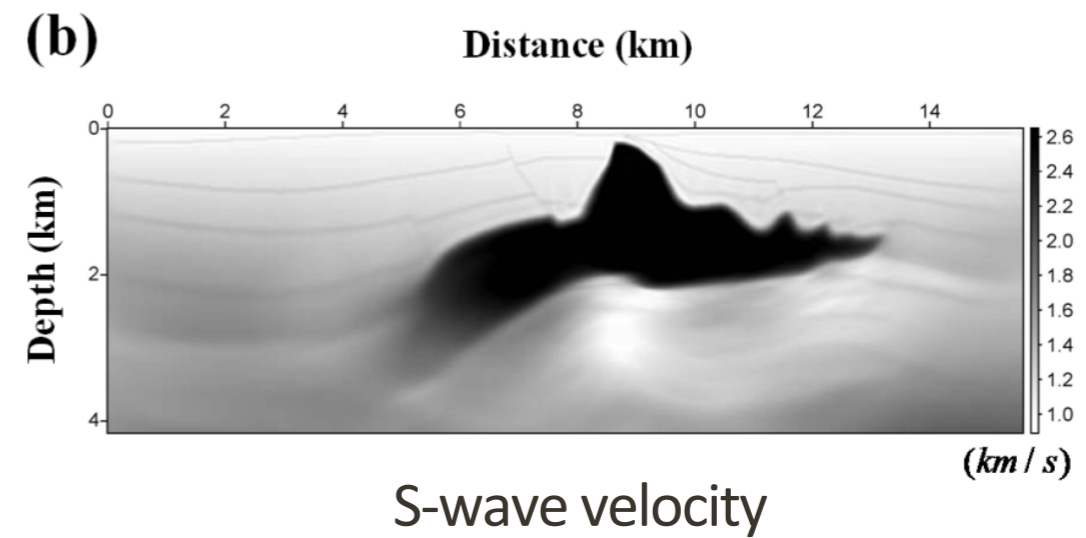
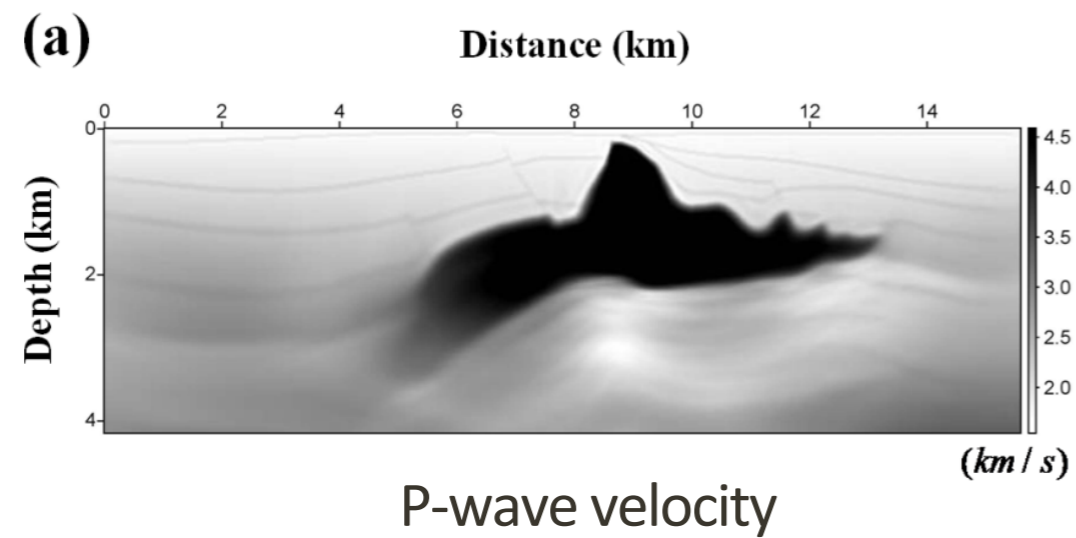
- Inverted velocity models - In the first stage



The density is fixed as 2 g/cm^3

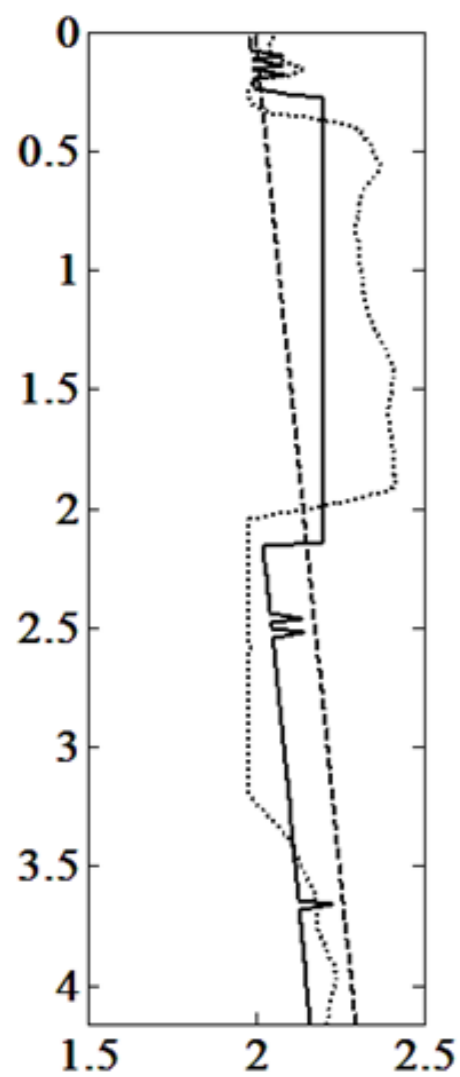
New inversion results

- Inversion results - In the second stage

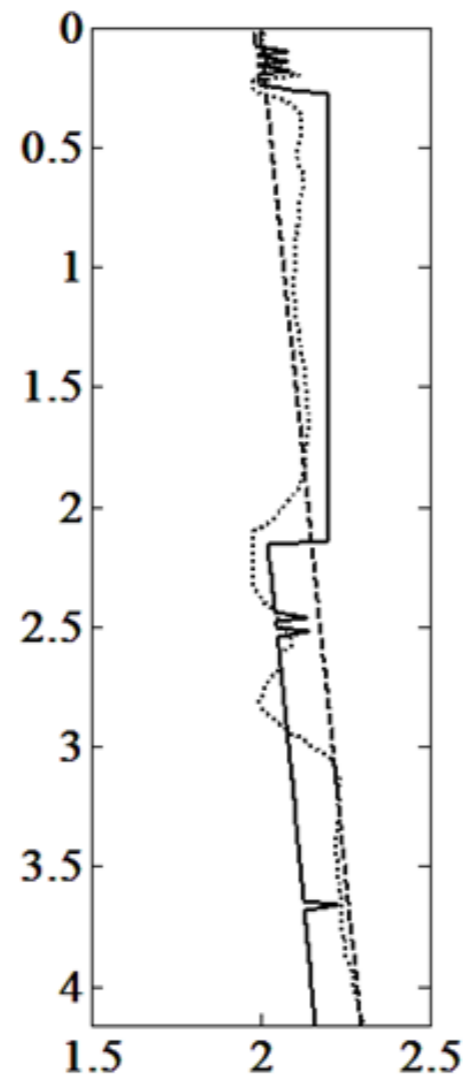


New inversion results

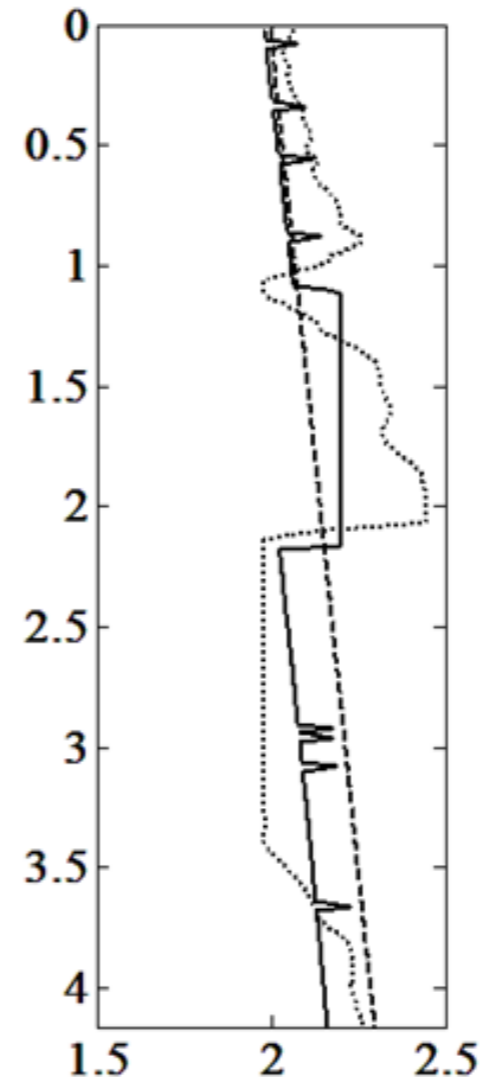
- Density profiles - Conventional method I & new method



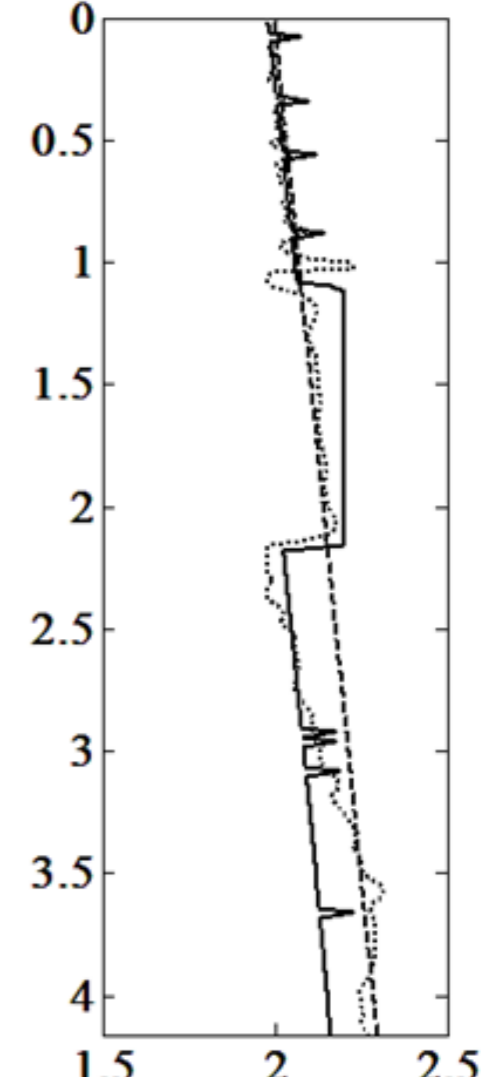
conventional



new



conventional



new

Contents

- FWI algorithm
- Conventional FWI & examples
- Parameter-selection strategy & examples
- **Conclusions**

Conclusions

- **Parameter selection strategy**
 - in the first stage, invert Lamé constants with density fixed as a constant
 - in the second stage, invert all parameters simultaneously
- **Numerical examples show that**
 - the new inversion strategy gives more reliable density models

Conclusions

- To enhance the accuracy of inverted density model
 - accurate velocity models are necessary, which are obtained with density fixed in the first stage of the new strategy

Future plans

- Collaboration with SLIM group
 - combine elastic waveform inversion for isotropic and VTI media with the Curvelet transform, simultaneous sources inversion, stochastic inversion, etc.
 - develop inversion techniques to properly recover salt models.

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