

# Random objective surface-wave waveform inversion

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**Related work has been accepted for publication in GEOPHYSICS:**

Pan, Y. and Gao, L., accepted, Radom-objective full waveform inversion of surface waves: *Geophysics*, doi:10.1190/geo-2019-0613.1



## ■ Conventional objective function in FWI

$$\min \Phi(\mathbf{m}) = \sum_{i=1}^{N_s} \mathcal{L}[\mathbf{d}_i^{obs}, \mathbf{d}_i^{syn}(\mathbf{m})],$$

$\mathcal{L}$  is a desired measure function,  $N_s$  is the total number of shots.

## ■ Common choices of $\mathcal{L}$ in surface-wave FWI (Pan et al., 2020)

L2 misfit  $\Phi_{l_2}(\mathbf{m}) = \frac{1}{2} \sum_{i=1}^{N_s} \|\mathbf{d}_i^{syn}(\mathbf{m}) - \mathbf{d}_i^{obs}\|_2^2,$

Envelope misfit  $\Phi_{Env}(\mathbf{m}) = \frac{1}{2} \sum_{i=1}^{N_s} \|\mathbf{e}_i^{syn}(\mathbf{m}) - \mathbf{e}_i^{obs}\|_2^2, \quad \mathbf{e} = \sqrt{\mathbf{d}^2 + \mathcal{H}^2(\mathbf{d})}.$

FK-spectra misfit  $\Phi_{FK}(\mathbf{m}) = \frac{1}{2} \sum_{i=1}^{N_s} \|\mathbf{D}_i^{syn}(\mathbf{m}) - \mathbf{D}_i^{obs}\|_2^2, \quad \mathbf{D} = |\mathcal{F}_{2D}(\mathbf{d})|$



## ■ Random objective waveform inversion (ROWI)

$$\underset{\mathbf{m}}{\operatorname{argmin}} \frac{1}{2} \left\{ \begin{array}{l} \| \mathbf{d}_1^{syn}(\mathbf{m}) - \mathbf{d}_1^{obs} \|_2^2, \quad \dots, \quad \| \mathbf{d}_{N_s}^{syn}(\mathbf{m}) - \mathbf{d}_{N_s}^{obs} \|_2^2 \\ \| \mathbf{e}_1^{syn}(\mathbf{m}) - \mathbf{e}_1^{obs} \|_2^2, \quad \dots, \quad \| \mathbf{e}_{N_s}^{syn}(\mathbf{m}) - \mathbf{e}_{N_s}^{obs} \|_2^2 \\ \| \mathbf{D}_1^{syn}(\mathbf{m}) - \mathbf{D}_1^{obs} \|_2^2, \quad \dots, \quad \| \mathbf{D}_{N_s}^{syn}(\mathbf{m}) - \mathbf{D}_{N_s}^{obs} \|_2^2 \end{array} \right\}$$

- Optimization: stochastic preconditioned steepest descent

*Randomly* choose one of the  $3 \cdot N_s$  objective functions at each iteration and optimize it with (Shigapov, 2019)

$$\mathbf{m}_{k+1} = \mathbf{m}_k - \alpha H_a^{-1} \delta \mathbf{m}_k$$

**Comp. cost: 1 iteration FWI =  $N_s$  iterations ROWI**



## ■ Why ROWI

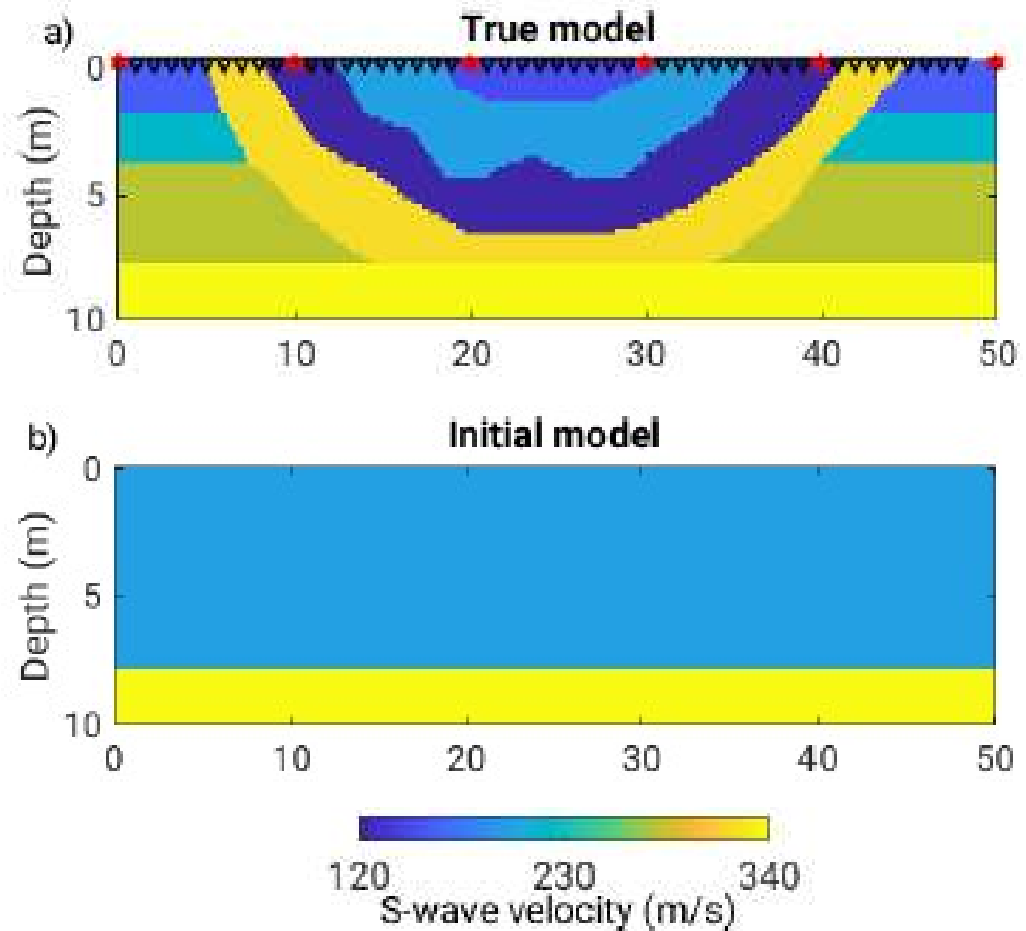
$$\operatorname{argmin}_{\mathbf{m}} \frac{1}{2} \left\{ \begin{array}{l} \|\mathbf{d}_1^{syn}(\mathbf{m}) - \mathbf{d}_1^{obs}\|_2^2, \quad \dots, \quad \|\mathbf{d}_{N_s}^{syn}(\mathbf{m}) - \mathbf{d}_{N_s}^{obs}\|_2^2 \\ \|\mathbf{e}_1^{syn}(\mathbf{m}) - \mathbf{e}_1^{obs}\|_2^2, \quad \dots, \quad \|\mathbf{e}_{N_s}^{syn}(\mathbf{m}) - \mathbf{e}_{N_s}^{obs}\|_2^2 \\ \|\mathbf{D}_1^{syn}(\mathbf{m}) - \mathbf{D}_1^{obs}\|_2^2, \quad \dots, \quad \|\mathbf{D}_{N_s}^{syn}(\mathbf{m}) - \mathbf{D}_{N_s}^{obs}\|_2^2 \end{array} \right\}$$

- Gain the **advantages of all measure functions**;
- Using one shot per iteration: **avoids redundant information**;
- Envelope misfit and the stochastic nature: **robust solution paths**.



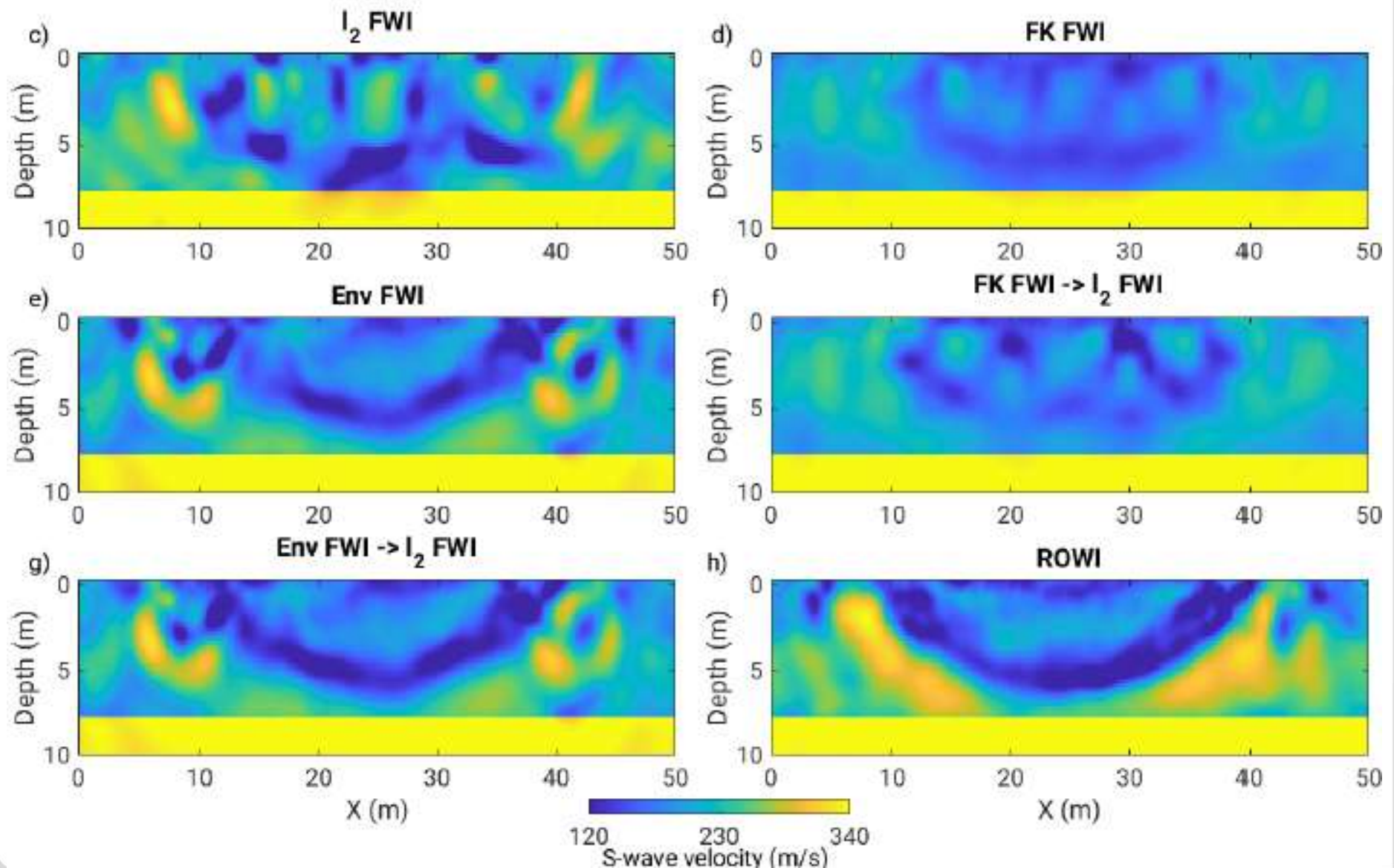
# Synthetic example

- 6 vertical-force sources
- First-order Gaussian wavelet of 25 Hz
- 48 receivers
- Homogeneous  $V_p$ ,  $Q_s$ ,  $Q_p$ , and density models
- Rayleigh wave (P-Sv)





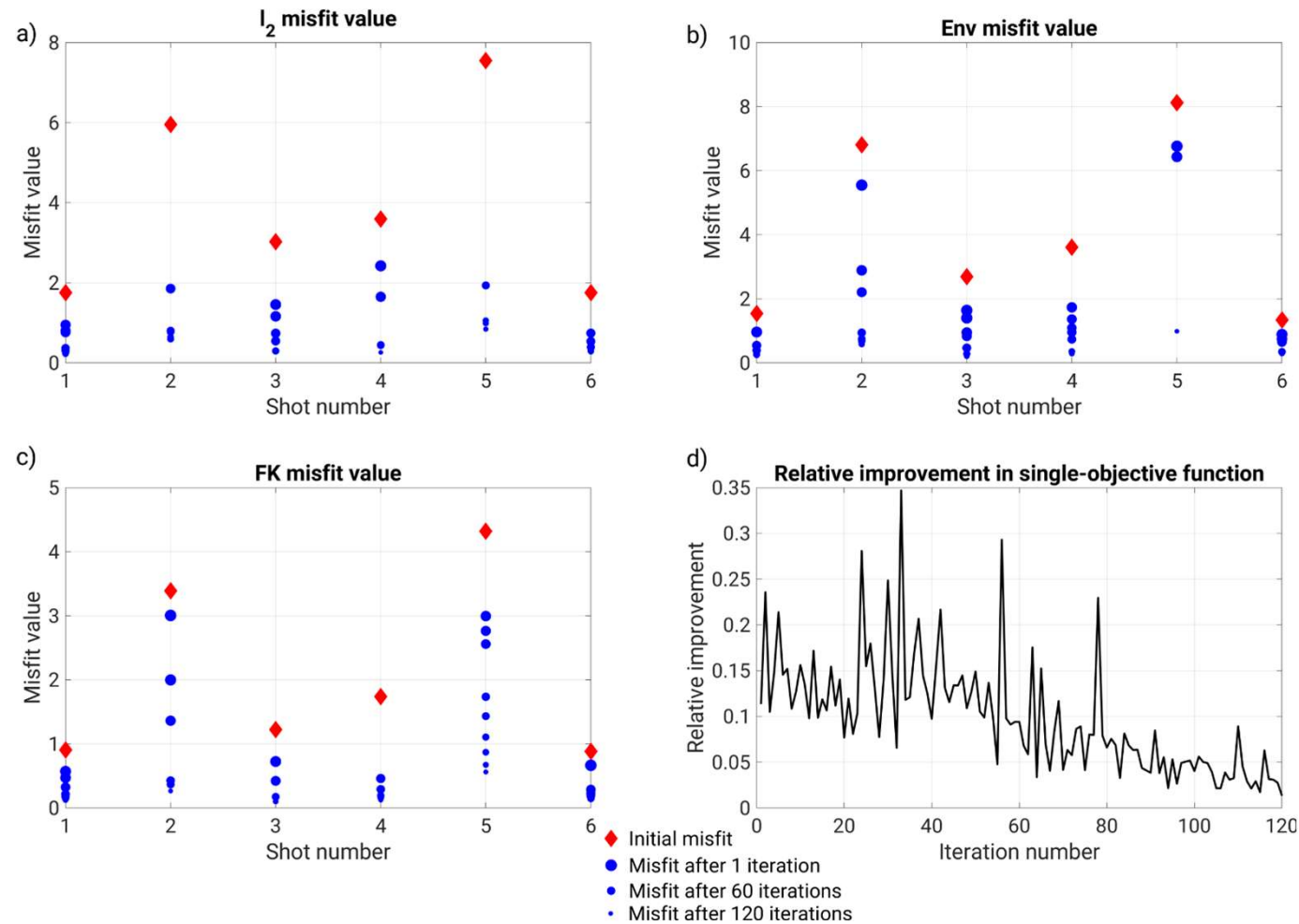
# Synthetic example





# Synthetic example

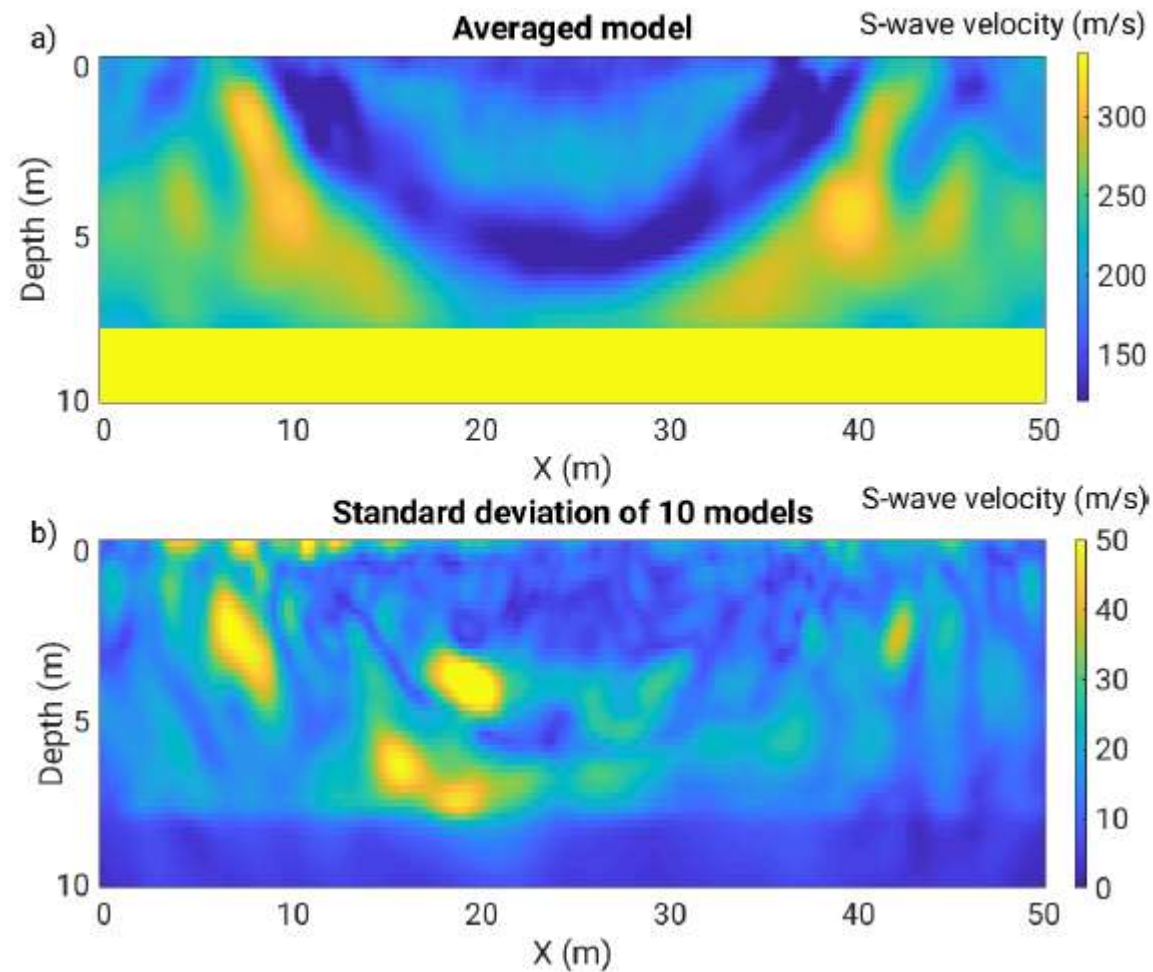
## ■ Data misfit of ROWI





# Synthetic example

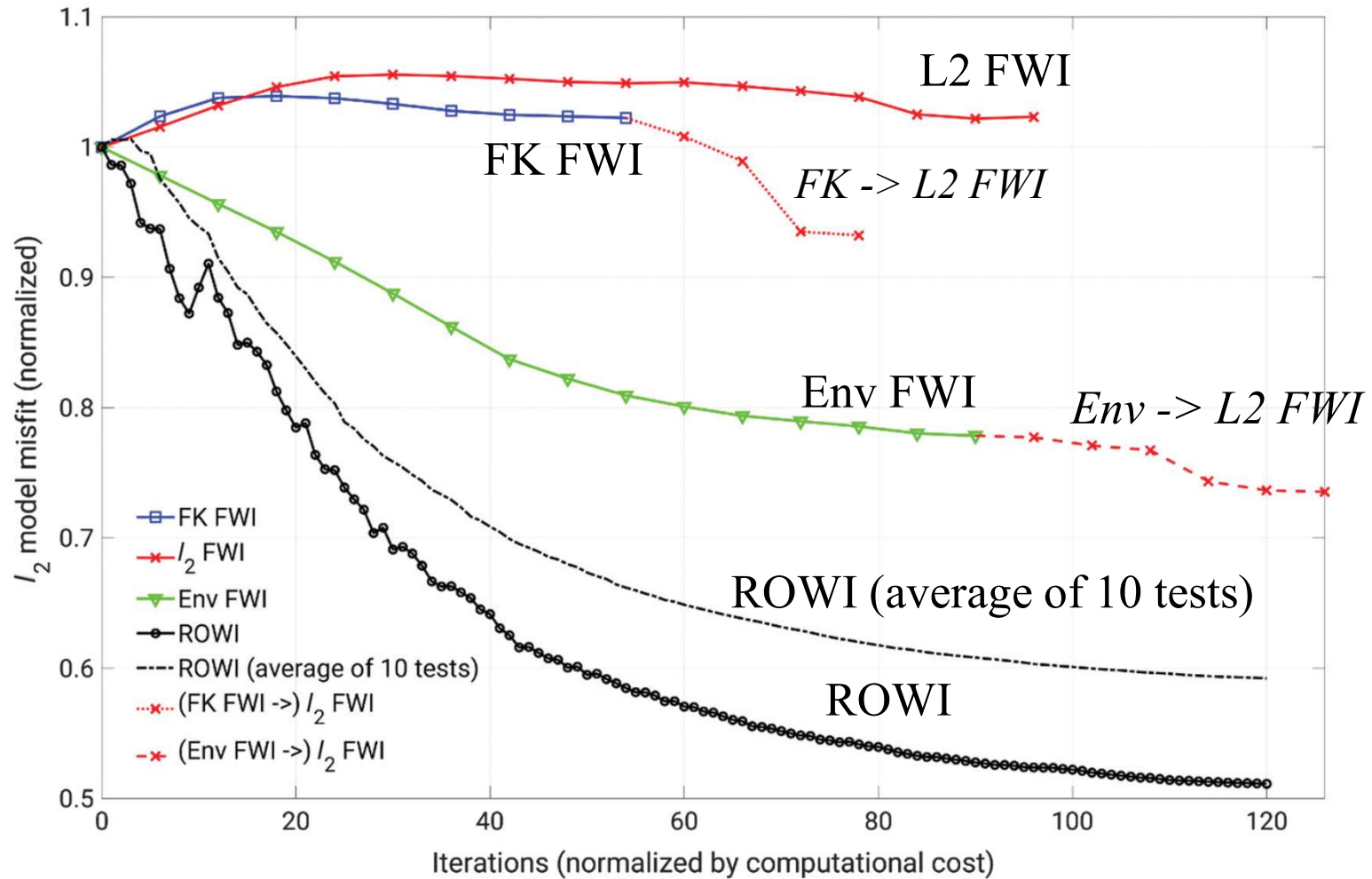
- Averaged result of 10 ROWIs with different random sequences





# Synthetic example

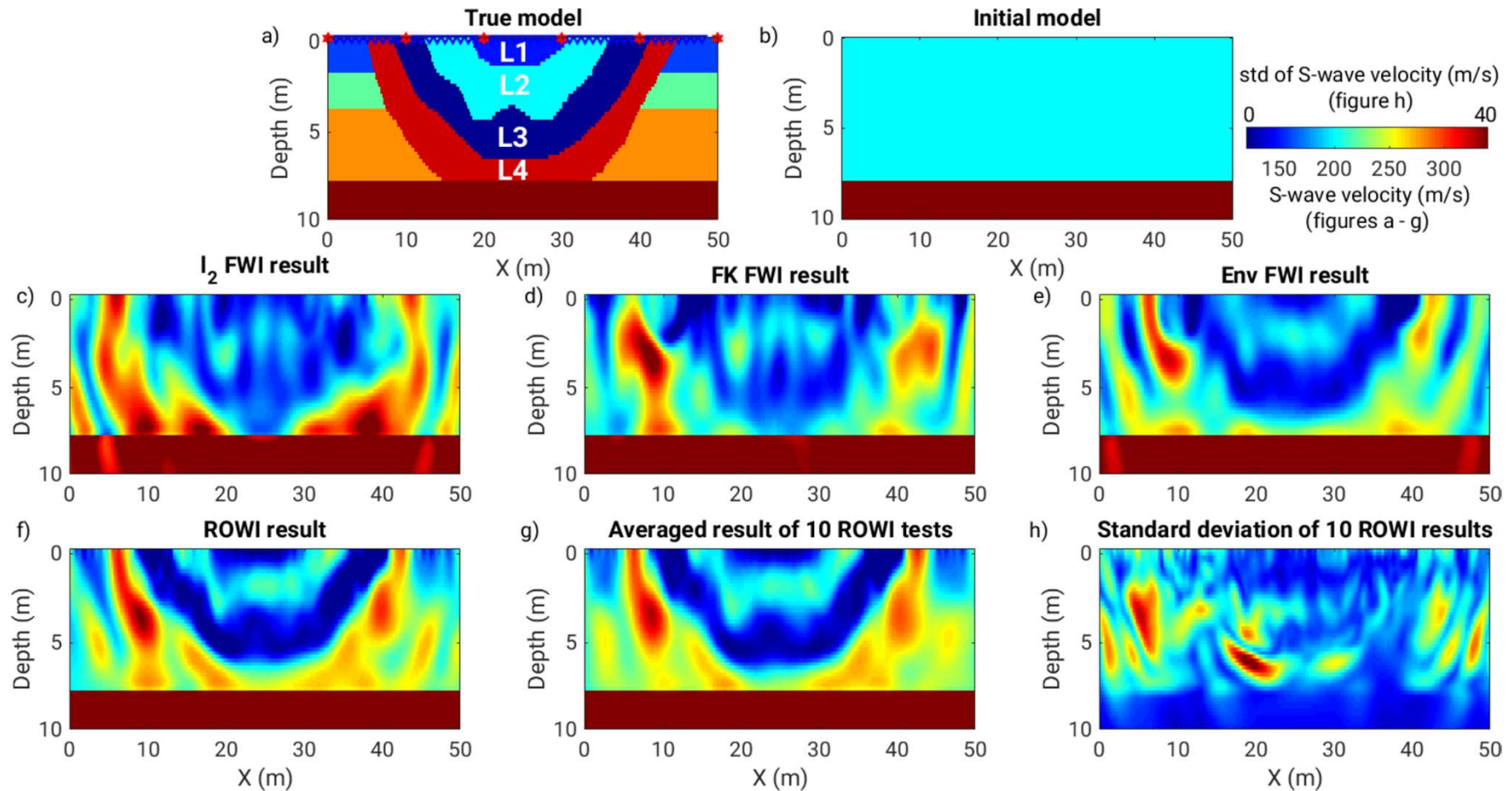
## Model misfit





# Synthetic example

## ■ Result on Love wave (SH wave)



(Pan et al., EAGE 2020)



# Field example

**Location: Rheinstetten, Karlsruhe, Germany**

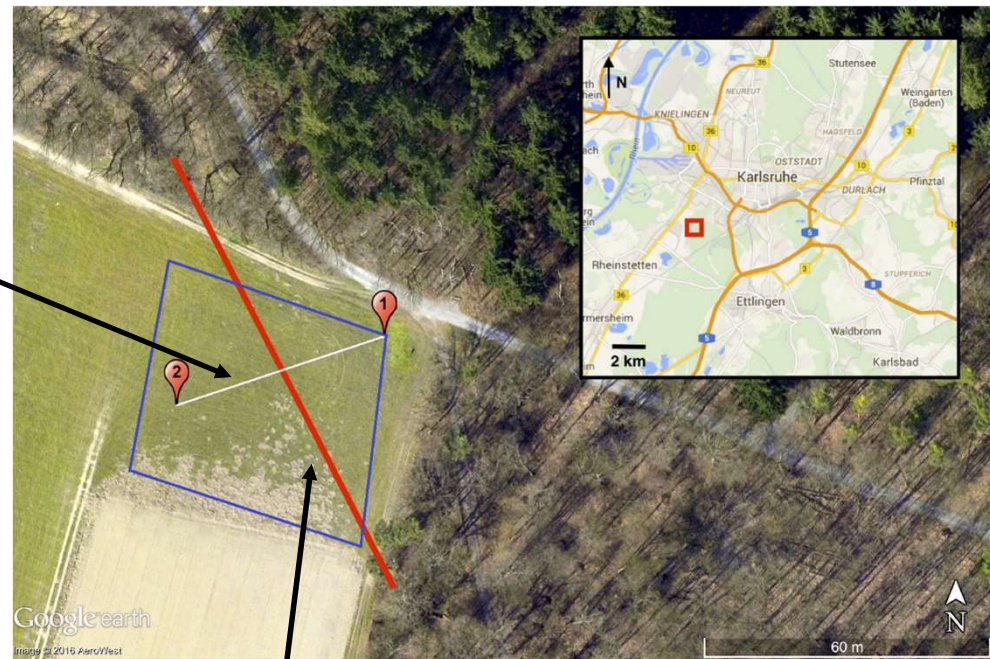
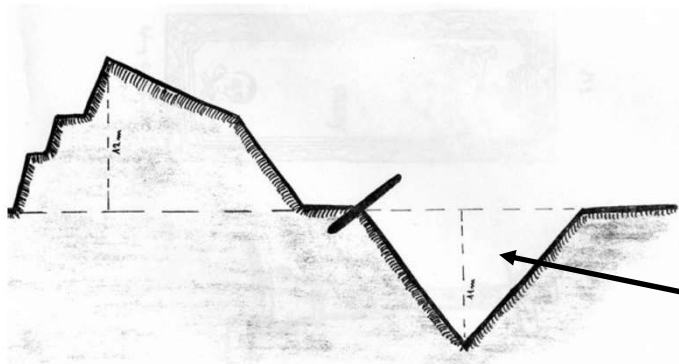
*Vertical hammer source*

*Vertical receivers (4.5 Hz)*

*48 traces with 1 m spacing*

*Source interval: 4 m*

*Number of shots: 12*



**This 'V'-shaped trench is refilled and is invisible from ground (red line)**



# Field example

## ■ ROWI + multiscale technique

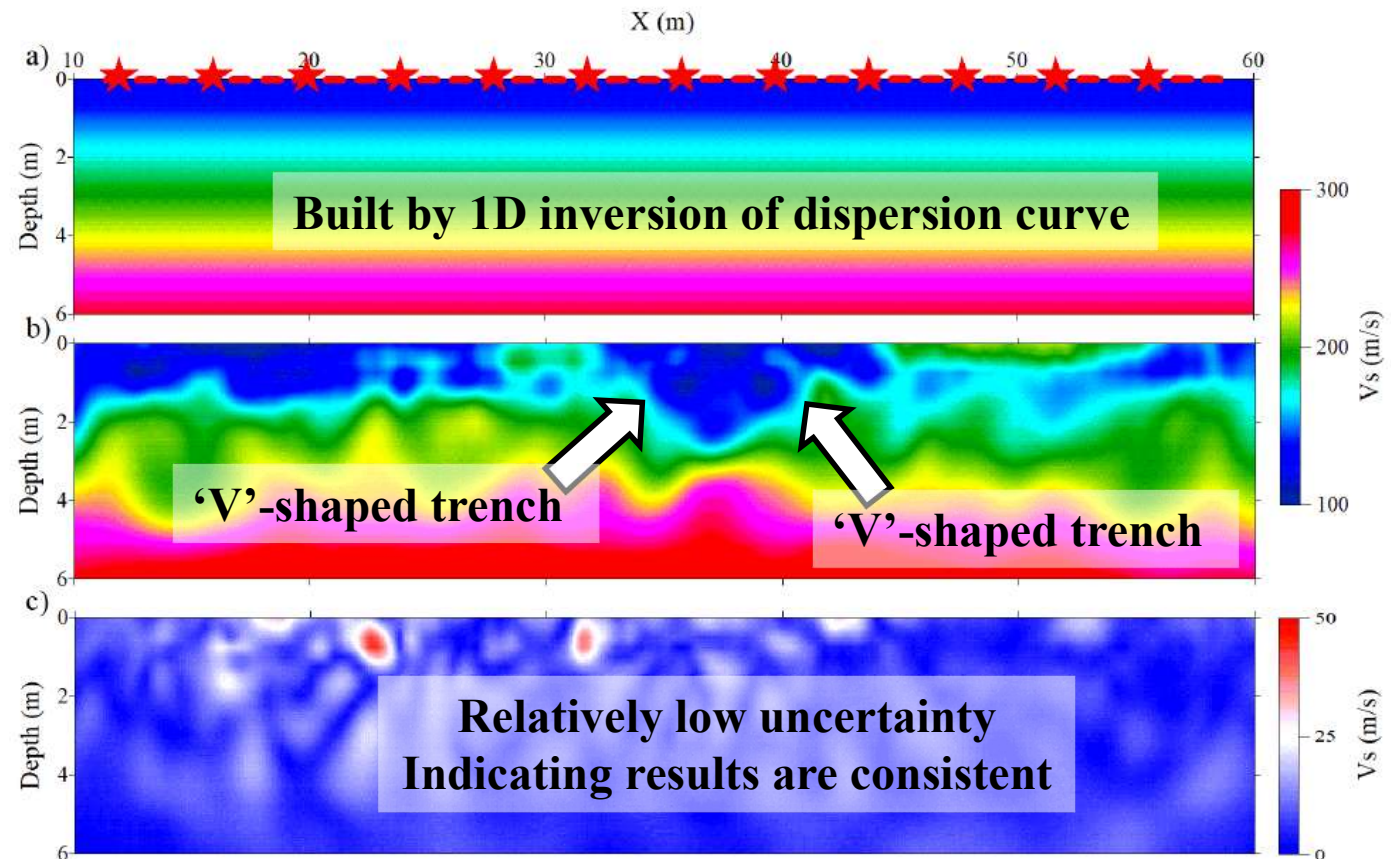
- 6 stages with a corner freq. of [10, 20, 30, 40, 50, 60] Hz
- 60 iterations at each stage ( $\sim 5$  iterations of conventional FWI)
- 4 times of ROWI with different random sequences
- Multiparameter viscoelastic inversion (Gao et al., 2020)



# Field example

## ■ Good initial model

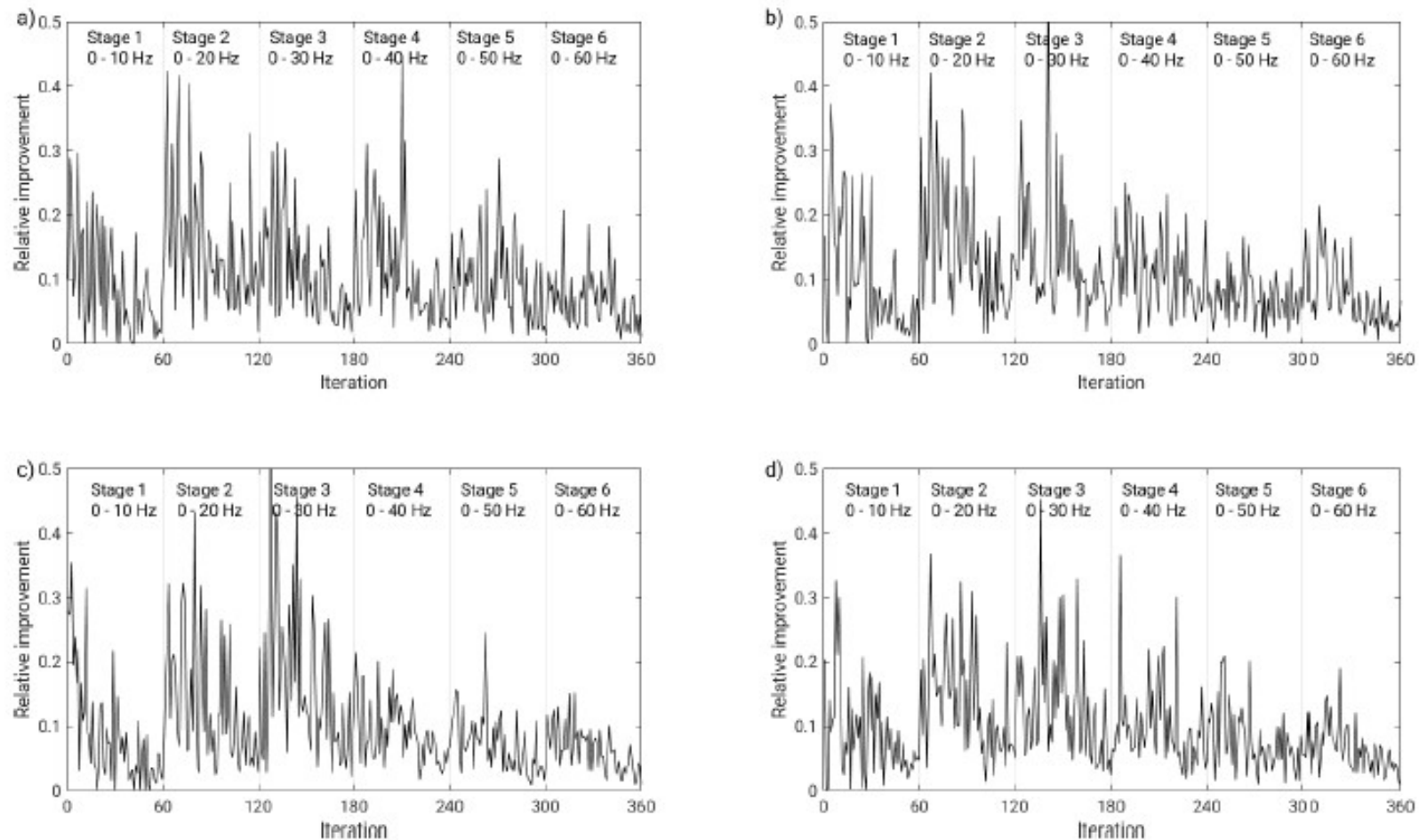
Initial model





# Field example

## ■ Data misfit (relative improvement in the misfit of 4 ROWIs)

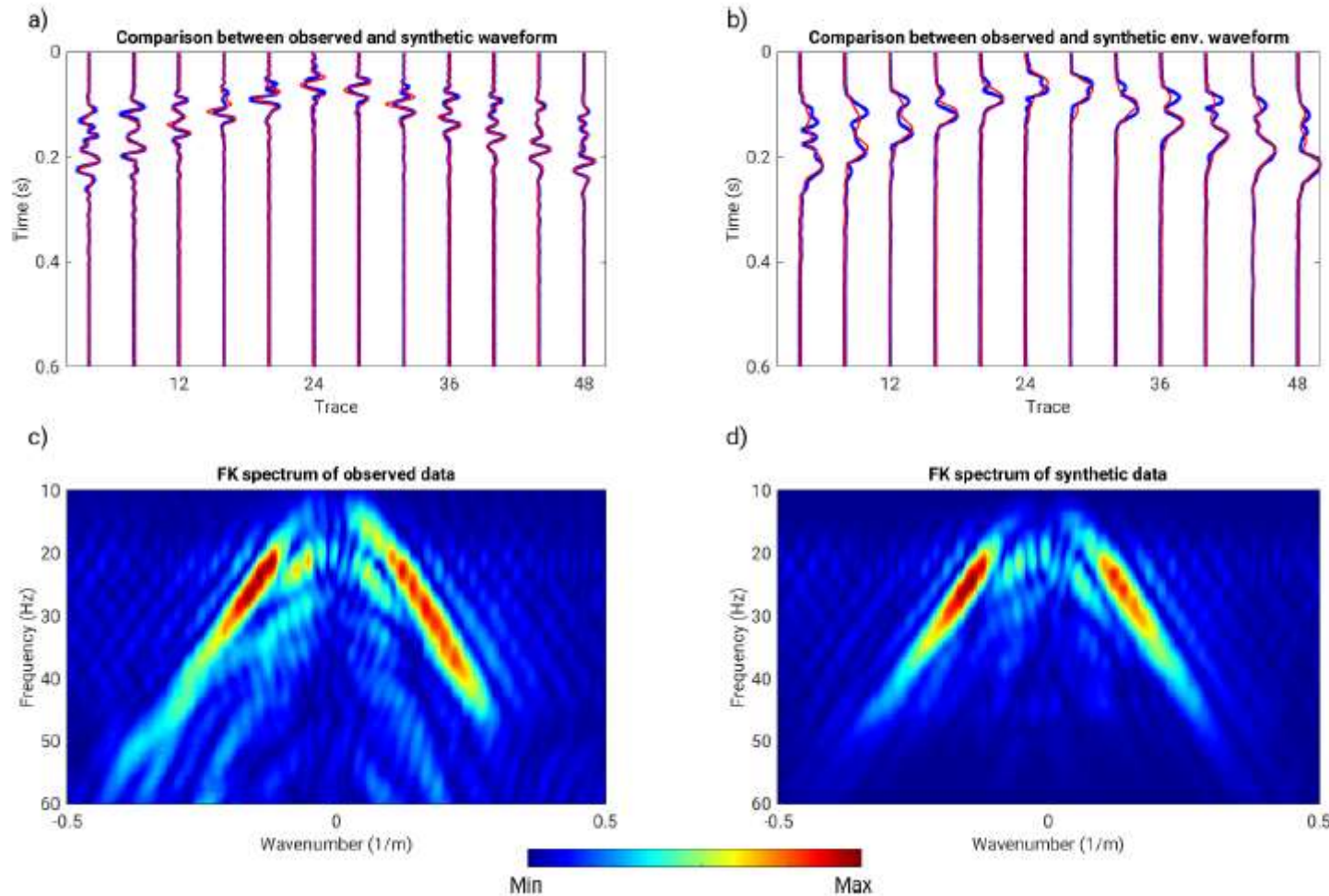




# Field example

## ■ Data fitting

central shot

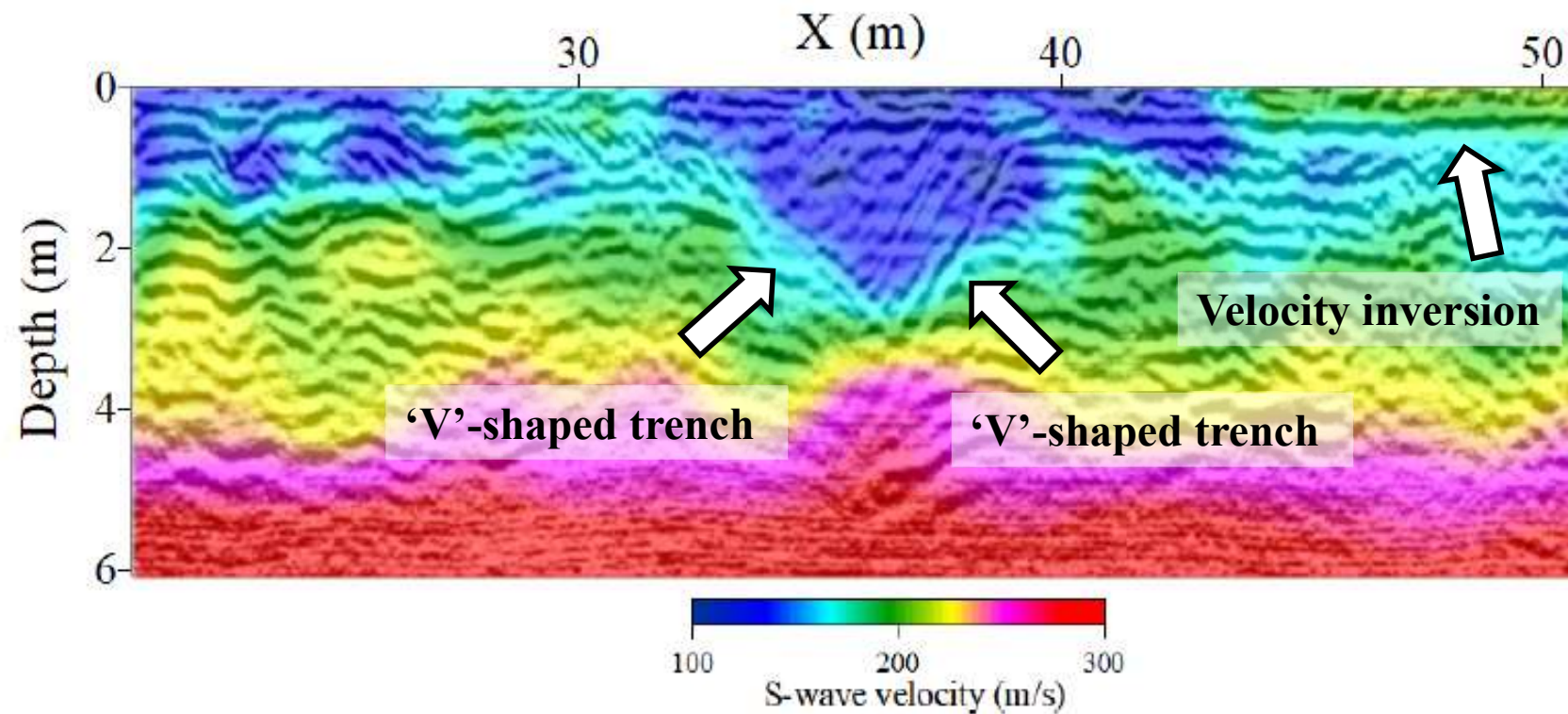




# Field example

## ■ Model evaluation

### Comparison with a migrated GPR profile





# Field example

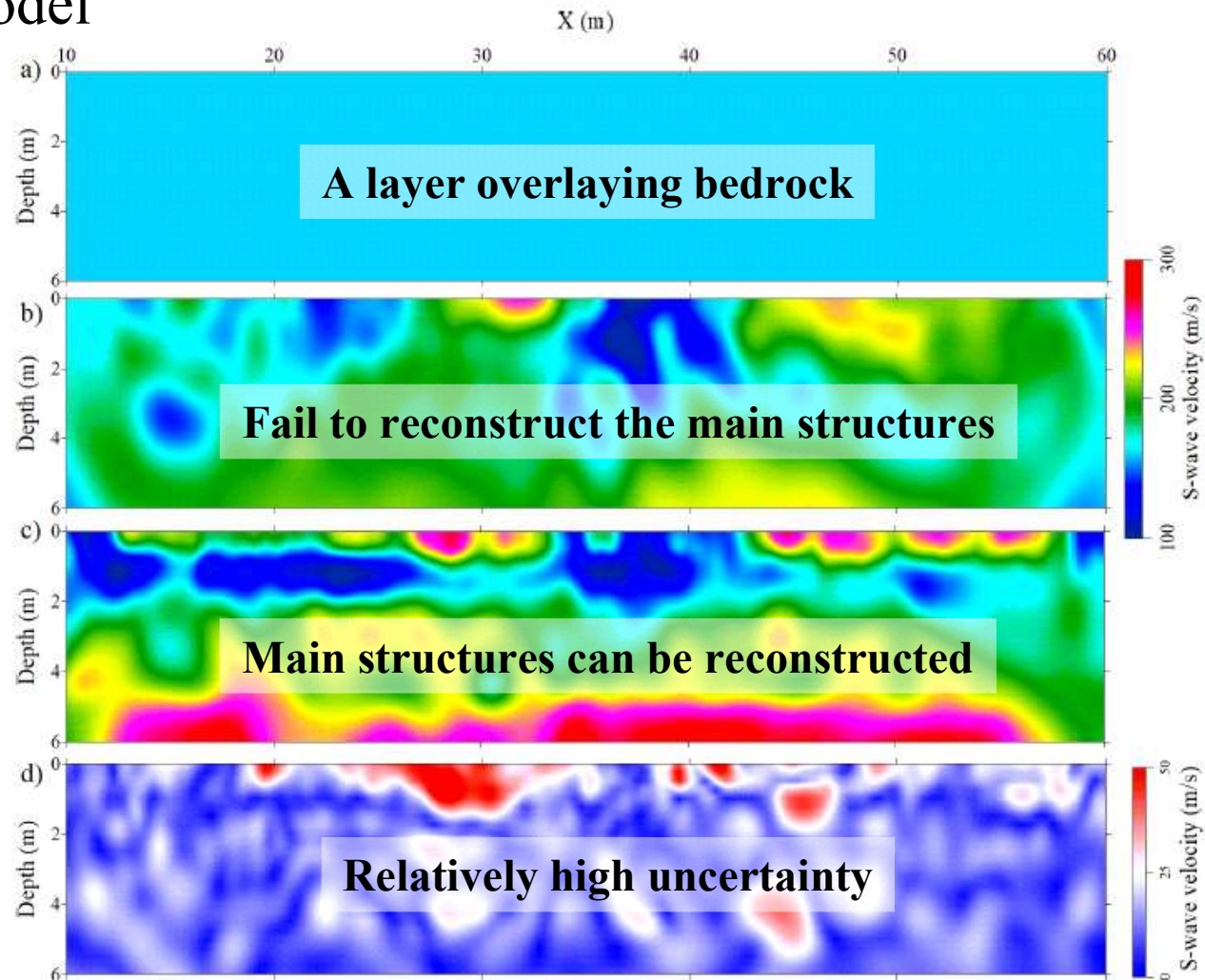
## ■ Poor initial model

Initial model

FWI  
(conventional  $l_2$ )

ROWI  
(average of 4 results)

Uncertainty  
(std of 4 results)





## Conclusions

- We proposed ROWI method under a multi-objective framework.
- Synthetic and field examples proved that ROWI outperforms conventional single-objective FWI.
- Standard deviation of the final models (weak Pareto solutions) provided useful information for uncertainty evaluation.

***Thank you!***



# References

1. Gao, L., Pan, Y., and Bohlen, T., 2020, 2D multi-parameter viscoelastic shallow-seismic full waveform inversion: reconstruction tests and first field-data application: *Geophysical Journal International*, doi:10.1093/gji/ggaa198
2. Pan, Y., Gao, L., and Shigapov, R., 2020, Multi-objective full waveform inversion of shallow-seismic wavefields: *Geophysical Journal International*, 220(3), 1619-1631.
3. Pan, Y. and Gao, L., accepted, Radom-objective full waveform inversion of surface waves: *Geophysics*, doi:10.1190/geo-2019-0613.1
4. Pan, Y., Gao, L., and Bohlen, T., accepted, Random-objective waveform inversion of shallow-seismic SH and Love waves, *82nd EAGE annual meeting*
5. Shigapov, R., 2019, Probabilistic waveform inversion: Quest for the law: Ph.D. thesis, Karlsruher Institut für Technologie (KIT).