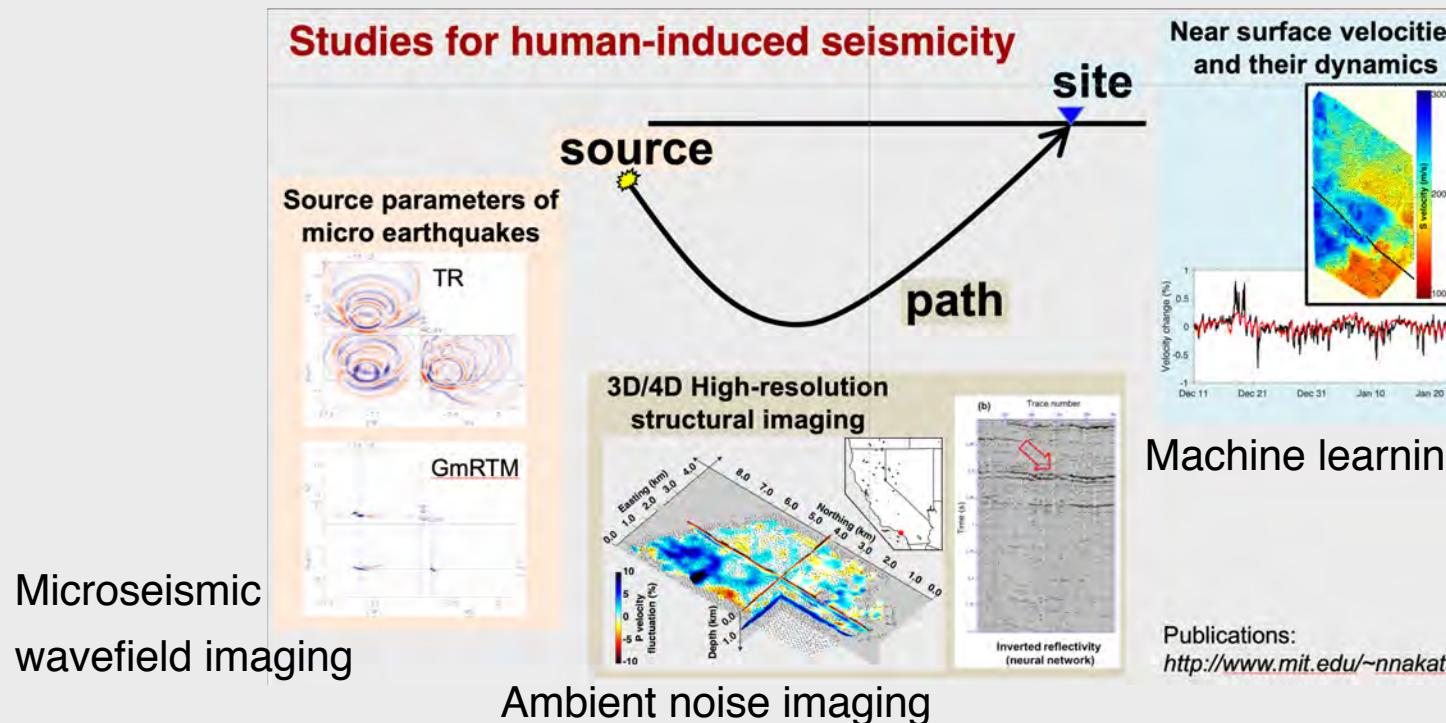


Towards high-resolution seismic imaging and monitoring



Nori Nakata (MIT)

Towards high-resolution seismic imaging and monitoring using ambient noise, machine learning, and microseismic events

Nori Nakata (MIT)

Yuji Kim, Jianhang Yin, Bin Lyu (Univ. Oklahoma)

Yuwei Wang (Southwest Petroleum Univ)

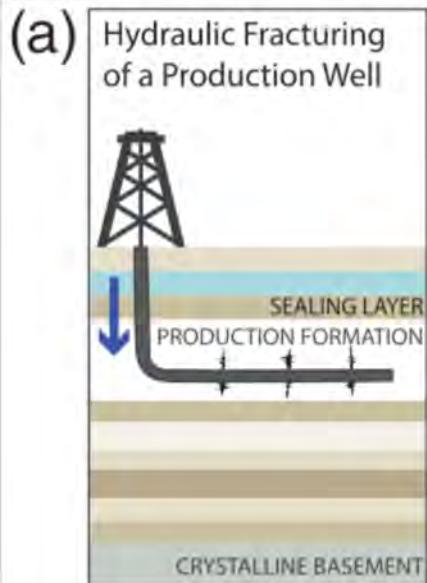
Rie Nakata Kamei, Zack Spica (Univ. Tokyo)

Greg Beroza (Stanford)

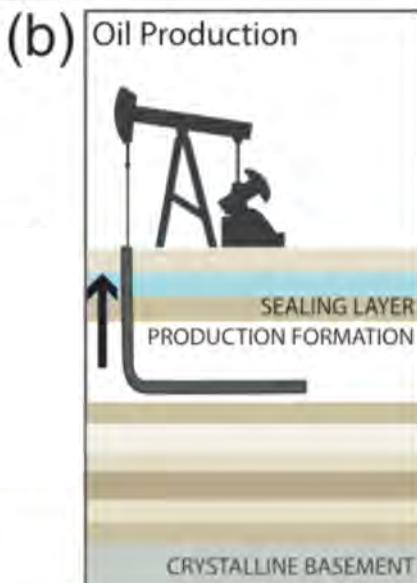
David Shelly (USGS)

Human-Induced Seismicity

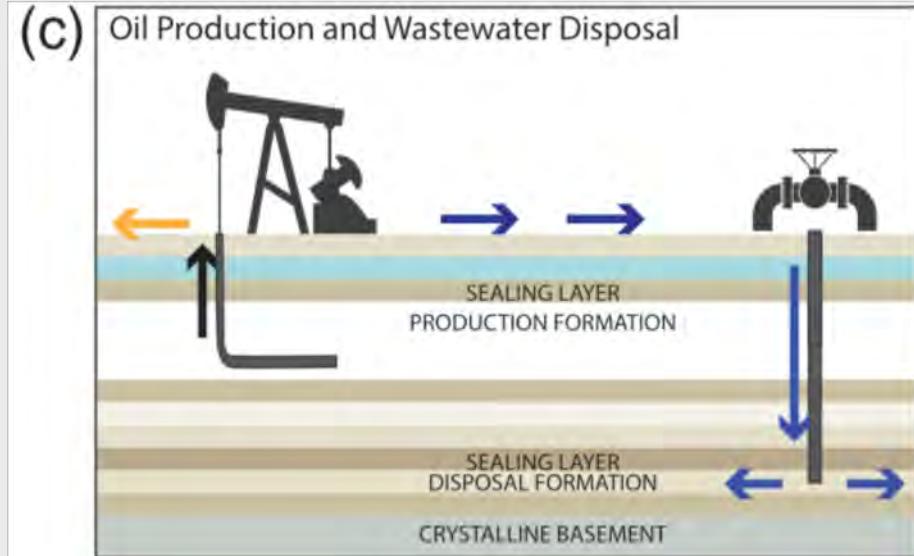
Fracking



Production



Wastewater disposal



- EOR
- Geothermal
- Mining

Human-Induced Seismicity

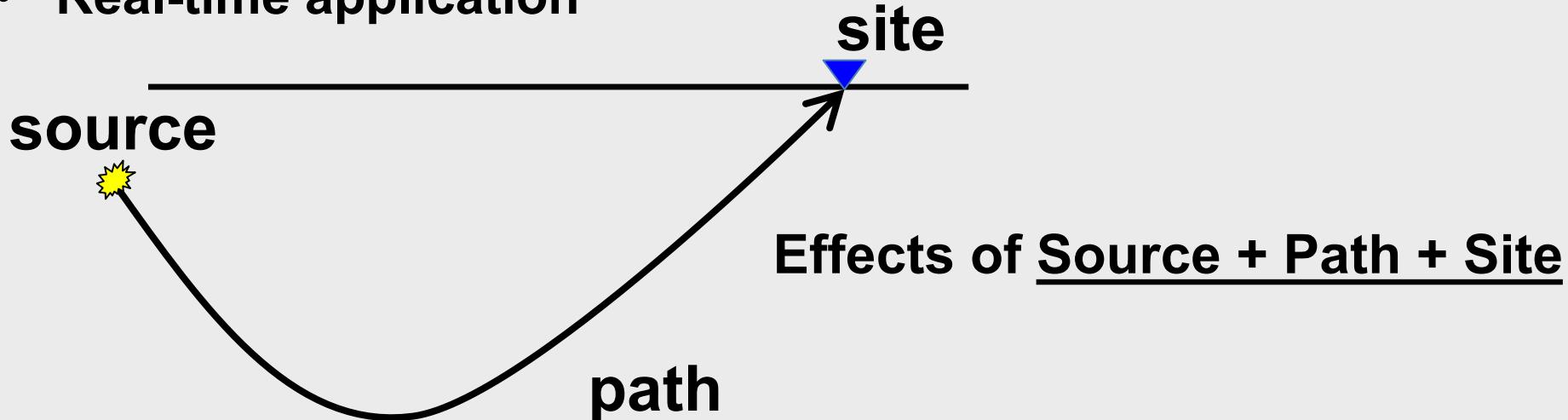
Tasks

- **Assessing seismic hazard**
- **Understanding seismic sources**
- **Understanding physics of seismicity**
- **Effectively enhancing production**
- **Real-time application**
- ...

Human-Induced Seismicity

Tasks

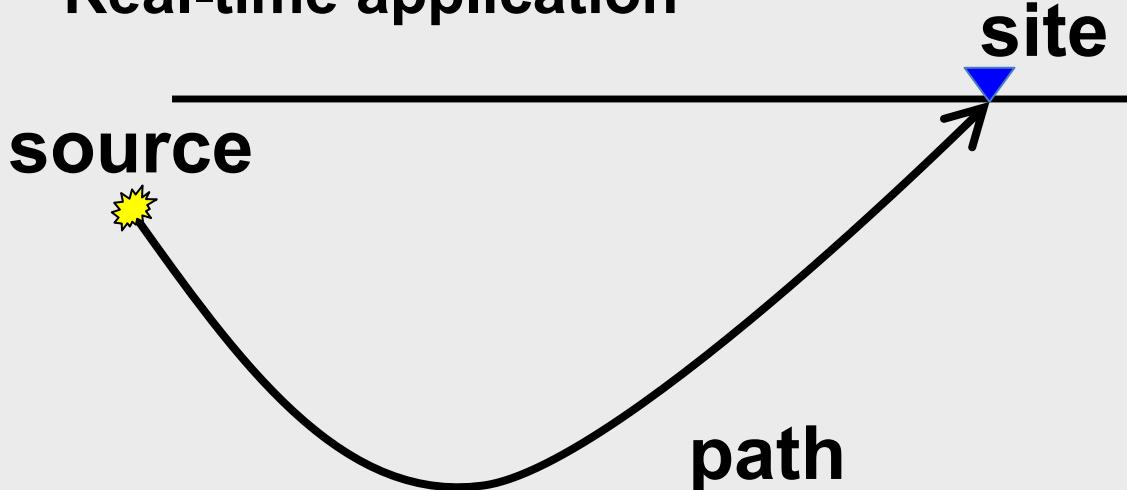
- Assessing seismic hazard
- Understanding seismic sources
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- Effectively enhancing production
- Real-time application



Human-Induced Seismicity

Tasks

- Assessing seismic hazard
- Understanding seismic sources
- Understanding physics of seismicity
- Effectively enhancing production
- Real-time application



My approach

Unique datasets

- Ambient noise
- Microseismic events
- Repeated active surveys
- Tube waves
- ...

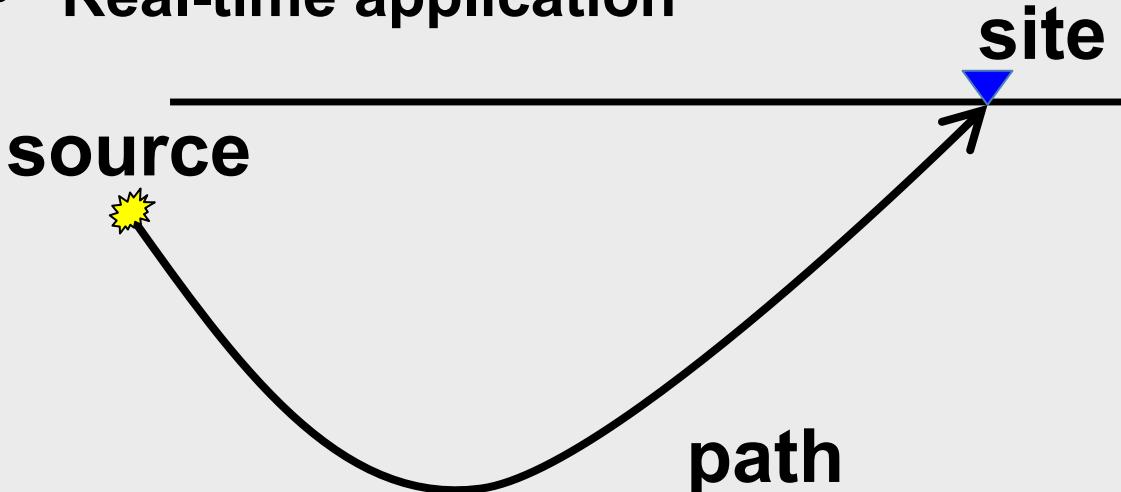
Advanced techniques

- Time-lapse FWI
- Machine learning
- Stochastic modeling
- Wavefield migration
- ...

Human-Induced Seismicity

Tasks

- Assessing seismic hazard
- Understanding seismic sources
- Understanding physics of seismicity
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My approach

Unique datasets

- Ambient noise
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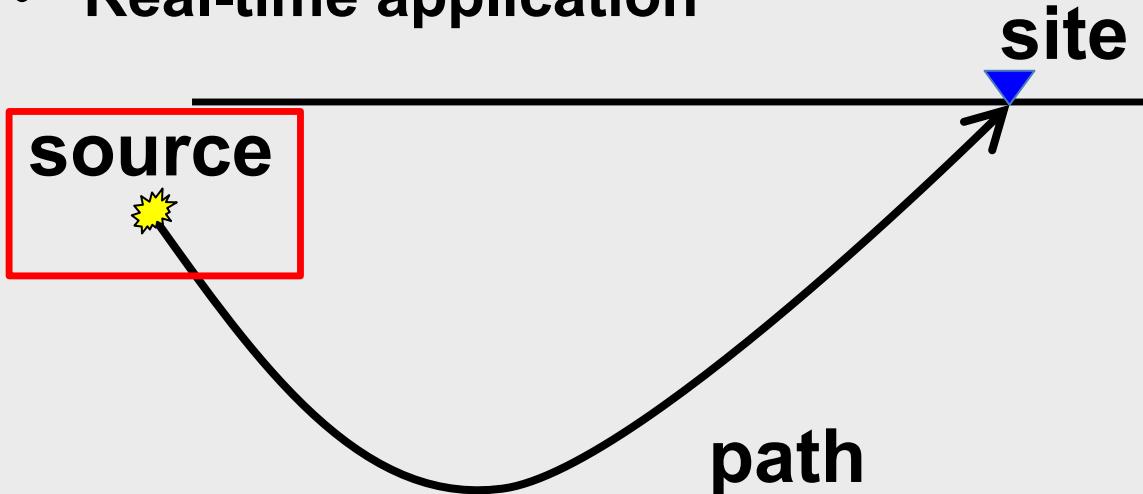
Advanced techniques

- Time-lapse FWI
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- ...

Human-Induced Seismicity

Tasks

- Assessing seismic hazard
- Understanding seismic sources
- Understanding physics of seismicity
- Effectively enhancing production
- Real-time application



My approach

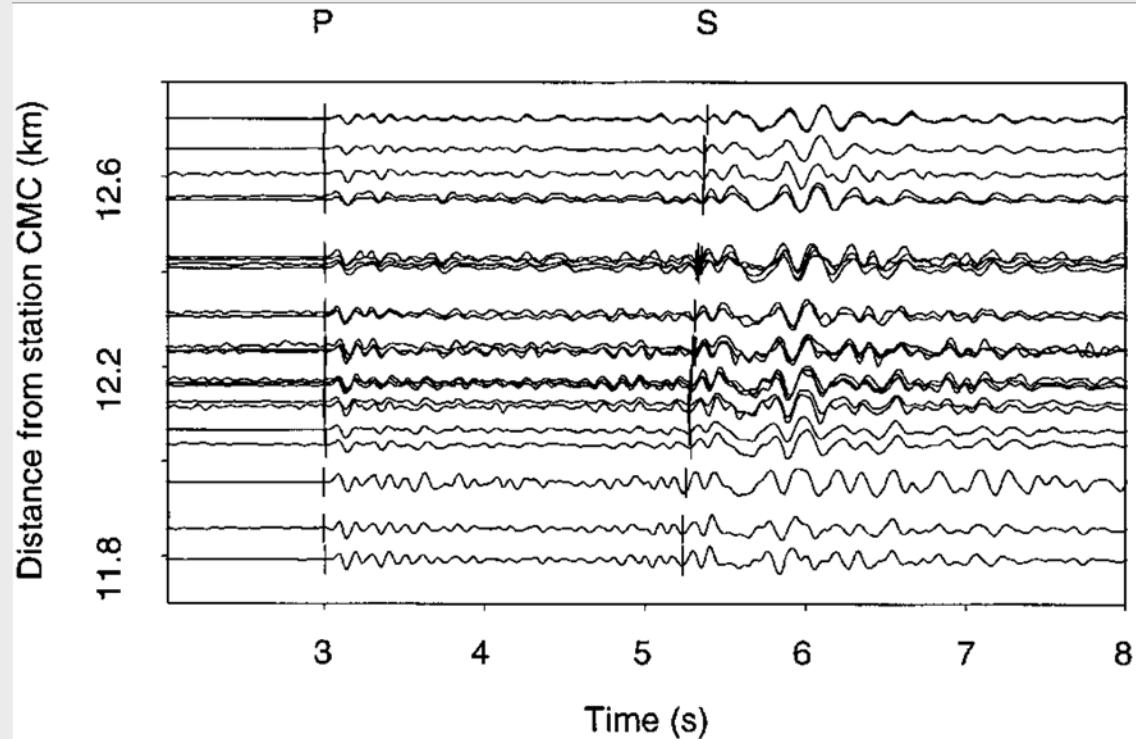
Unique datasets

- Ambient noise
- **Microseismic events**
- Repeated active surveys
- Tube waves
- ...

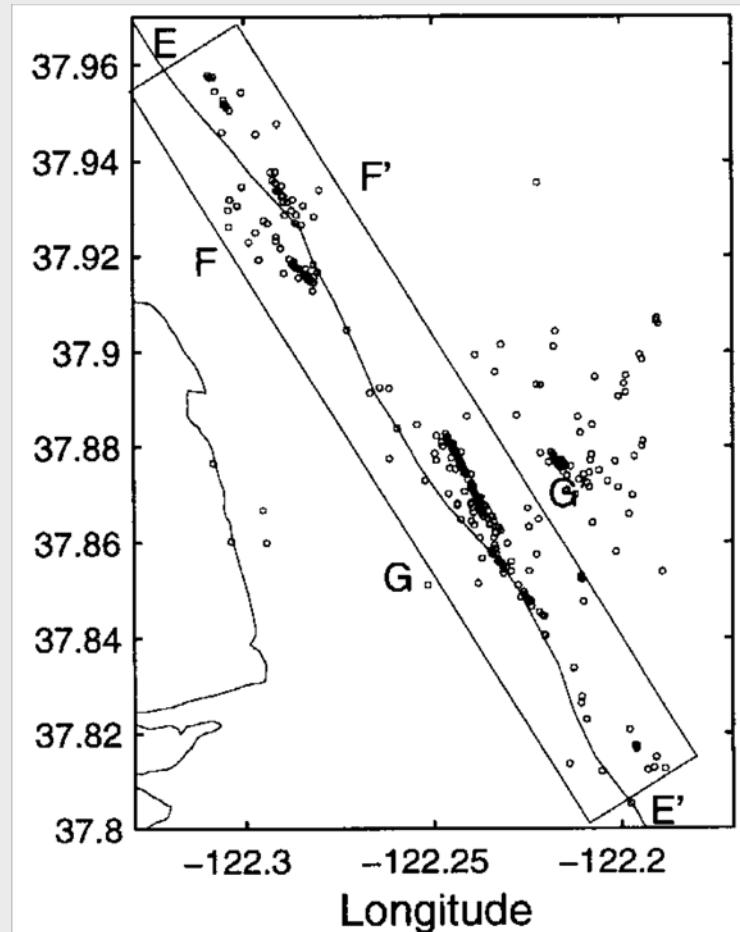
Advanced techniques

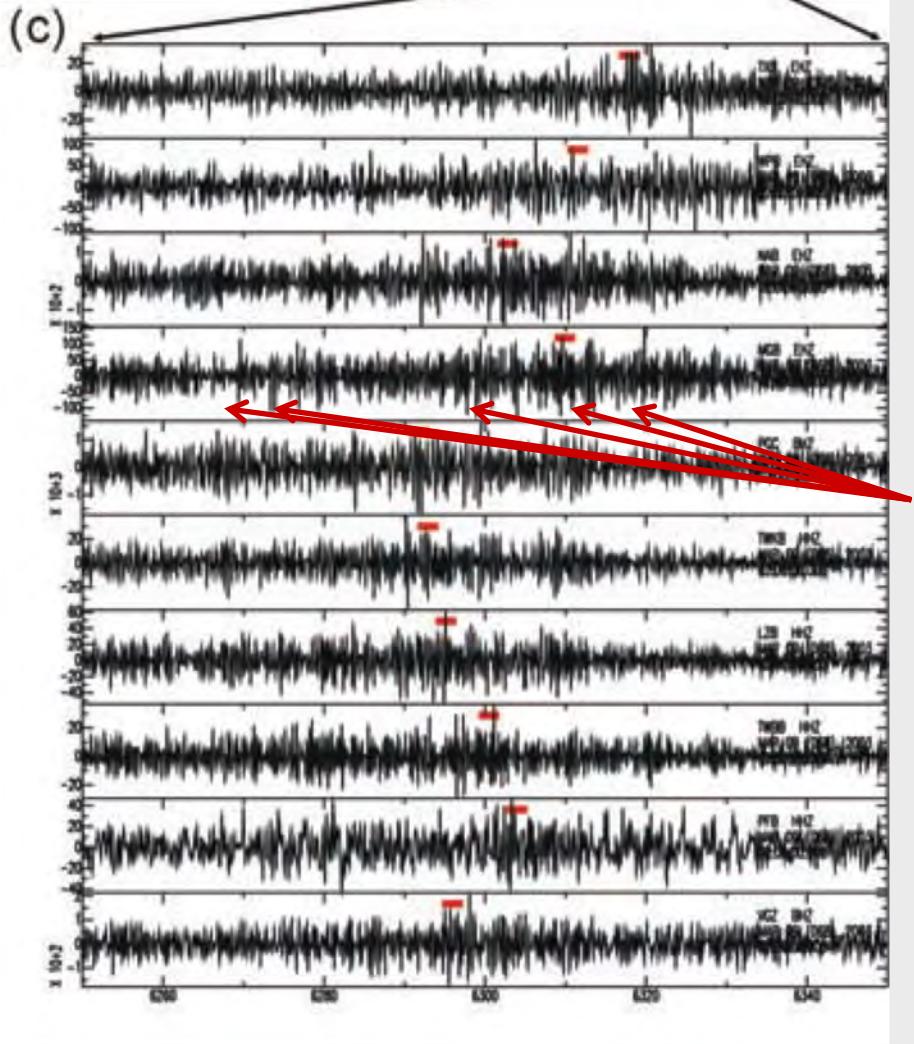
- Time-lapse FWI
- Machine learning
- Stochastic modeling
- **Wavefield migration**
- ...

Arrival time is useful if clearly identified



Arrival time picks are used to locate source.





Small, weak, and extended events are hard to detect/locate

Wave arrivals?

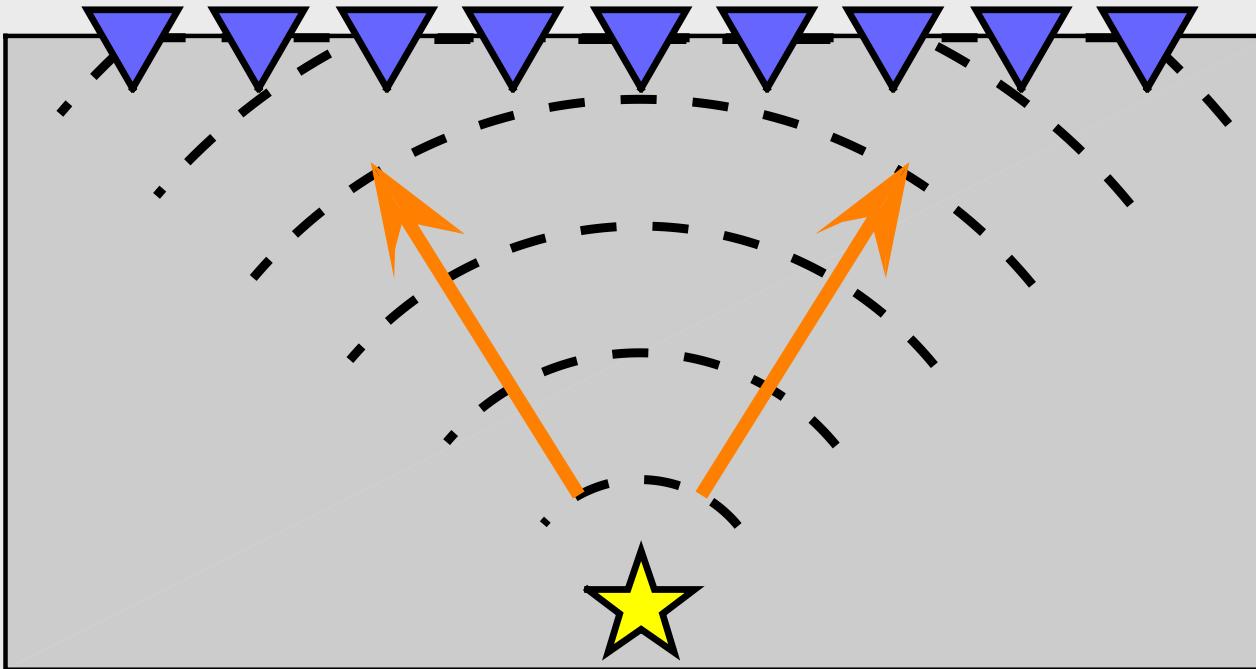


We can't pick arrival times accurately

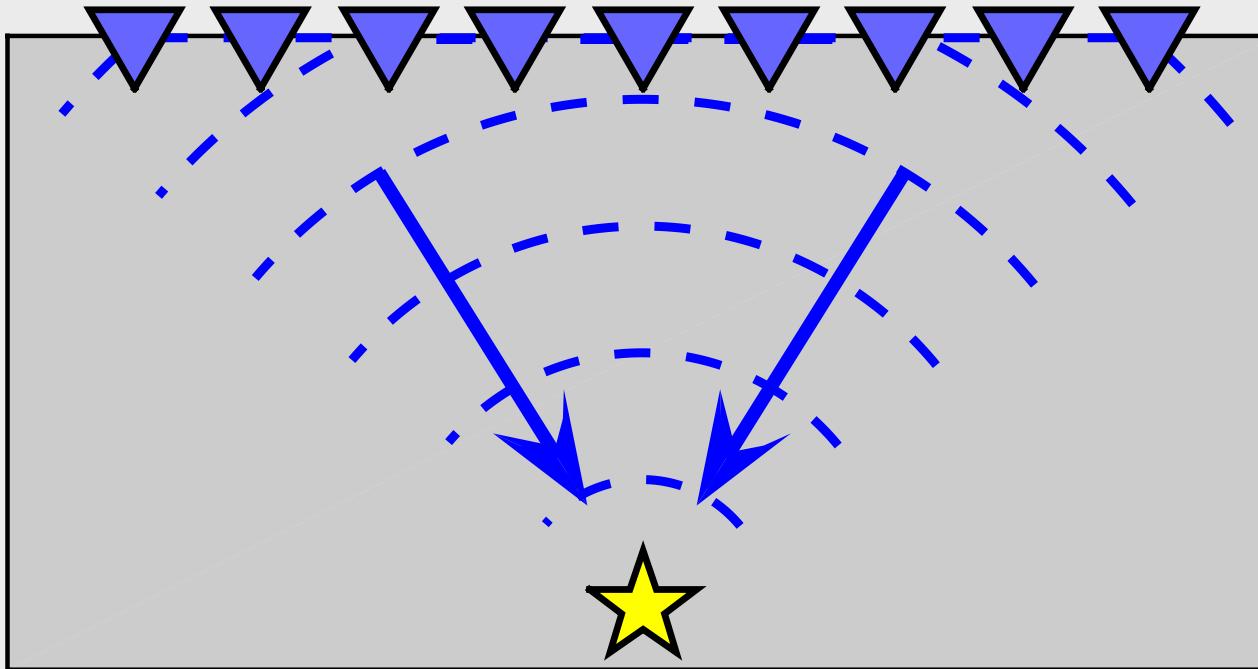


Migration-based processing

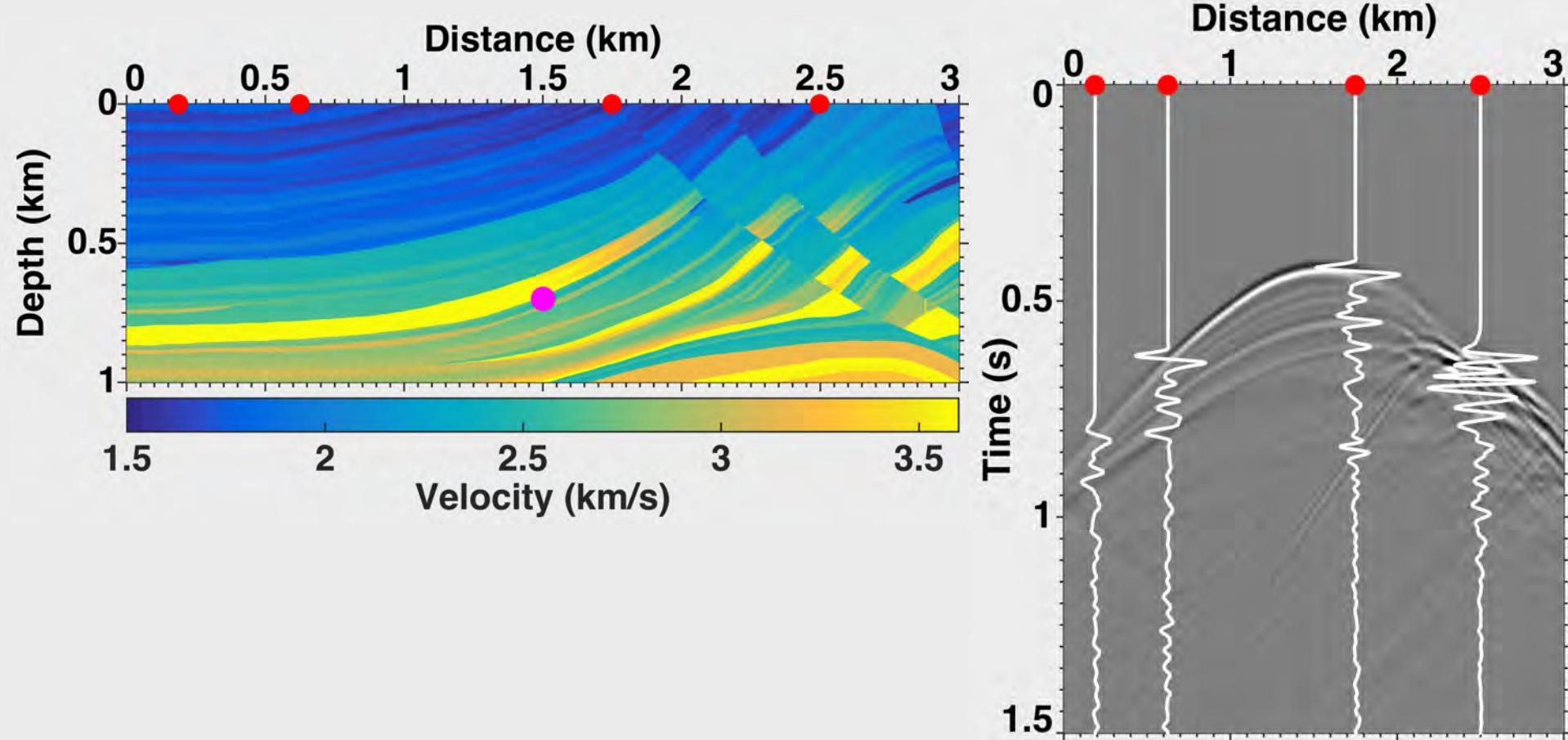
Recording microseismic data



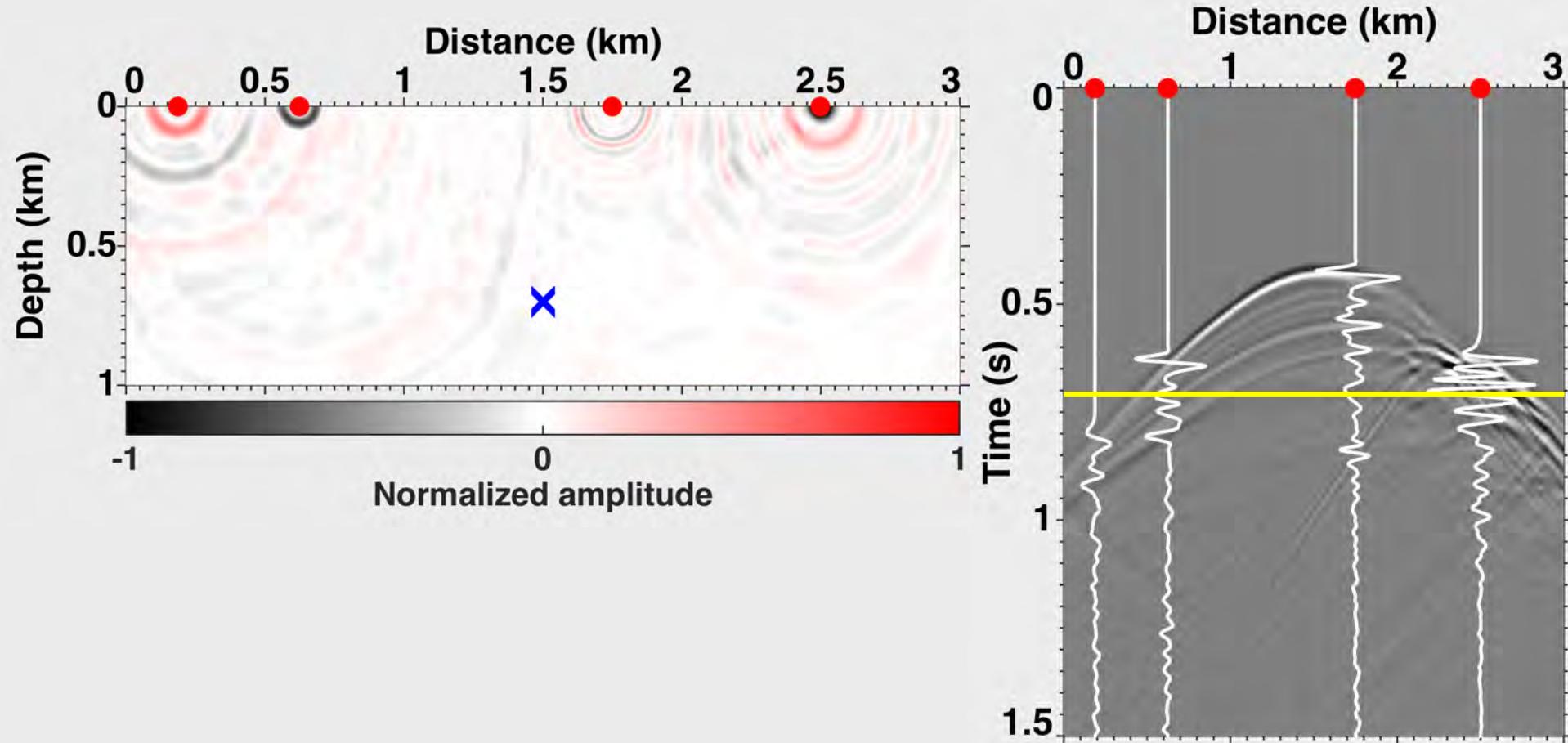
Time-reversal imaging



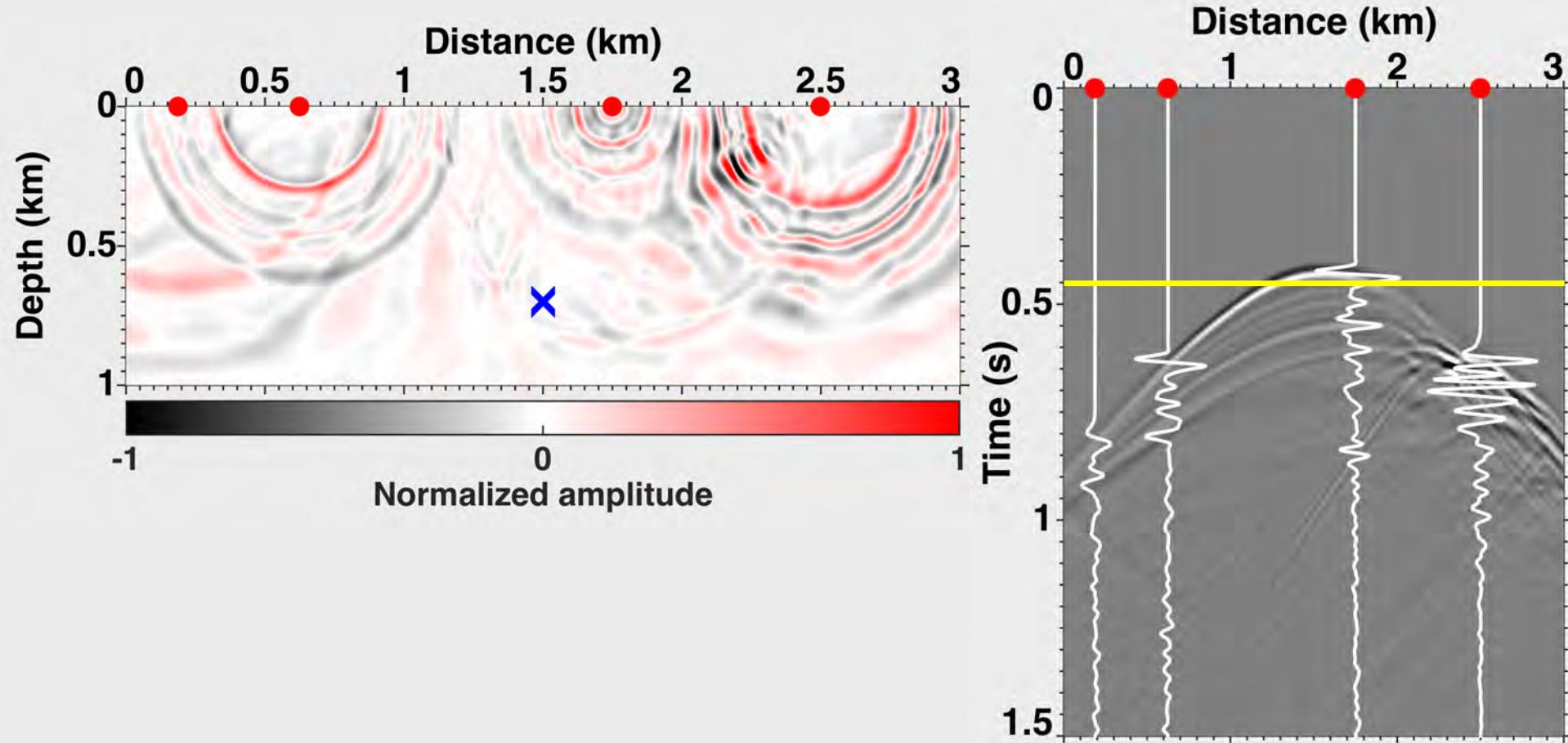
Numerical example



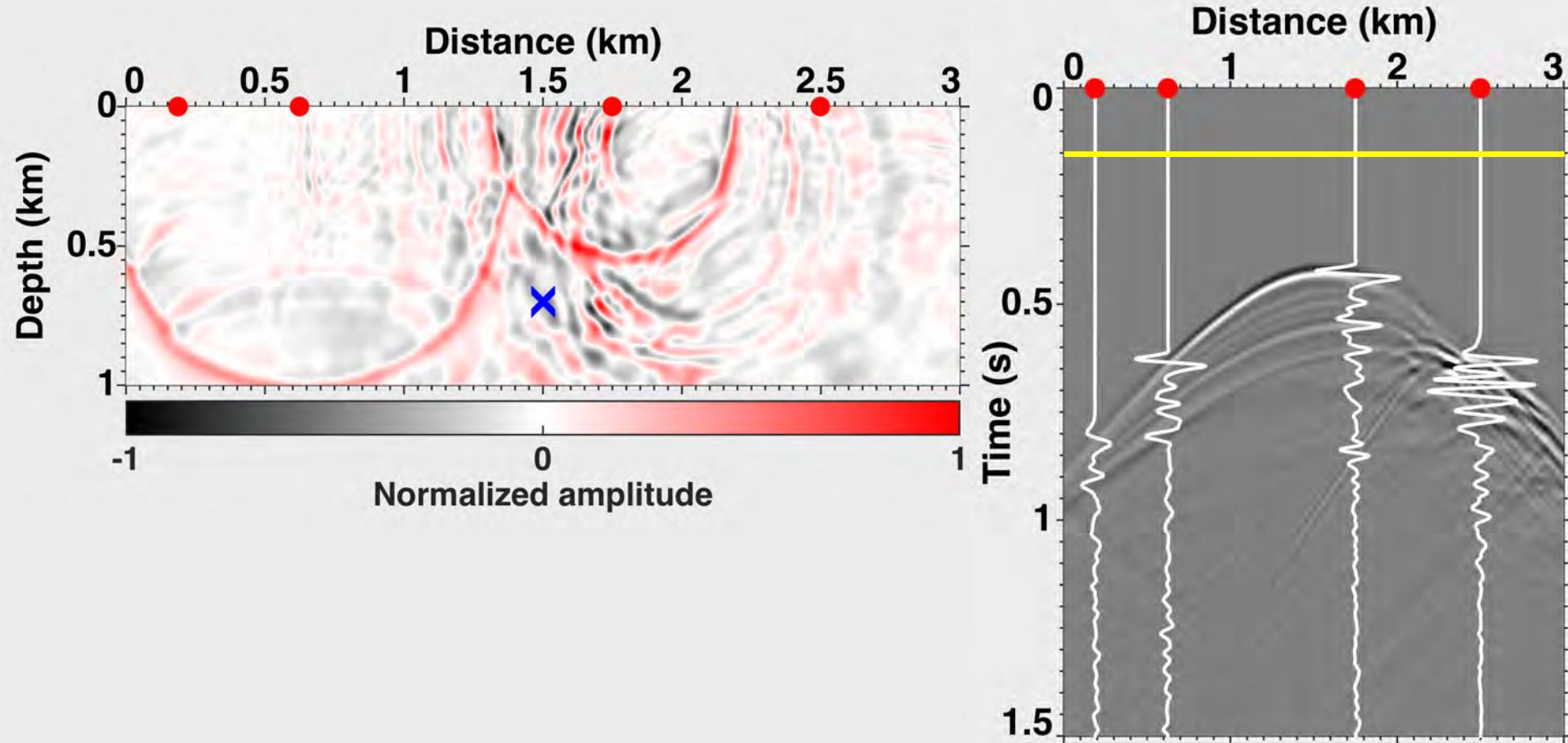
Time-reversal imaging



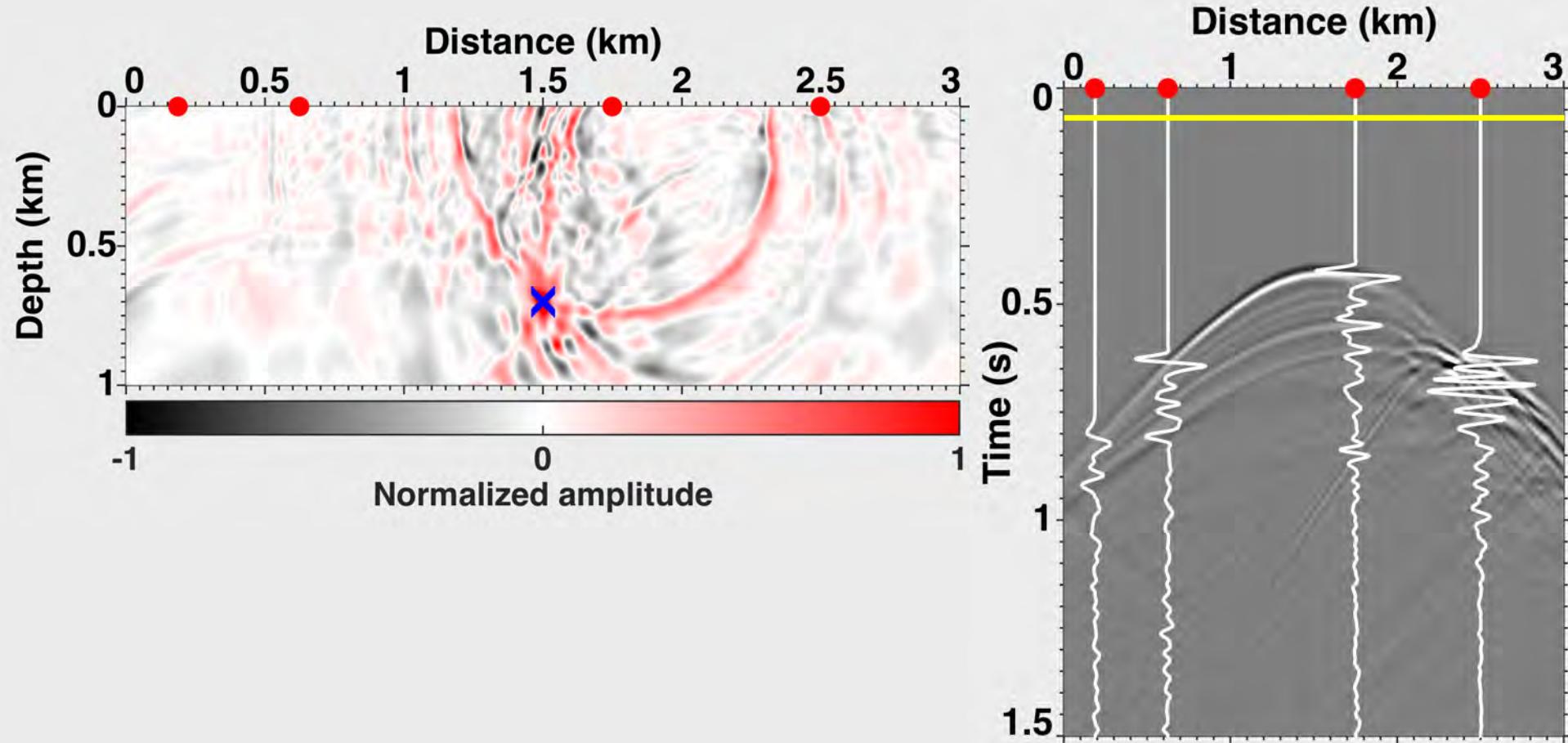
Time-reversal imaging



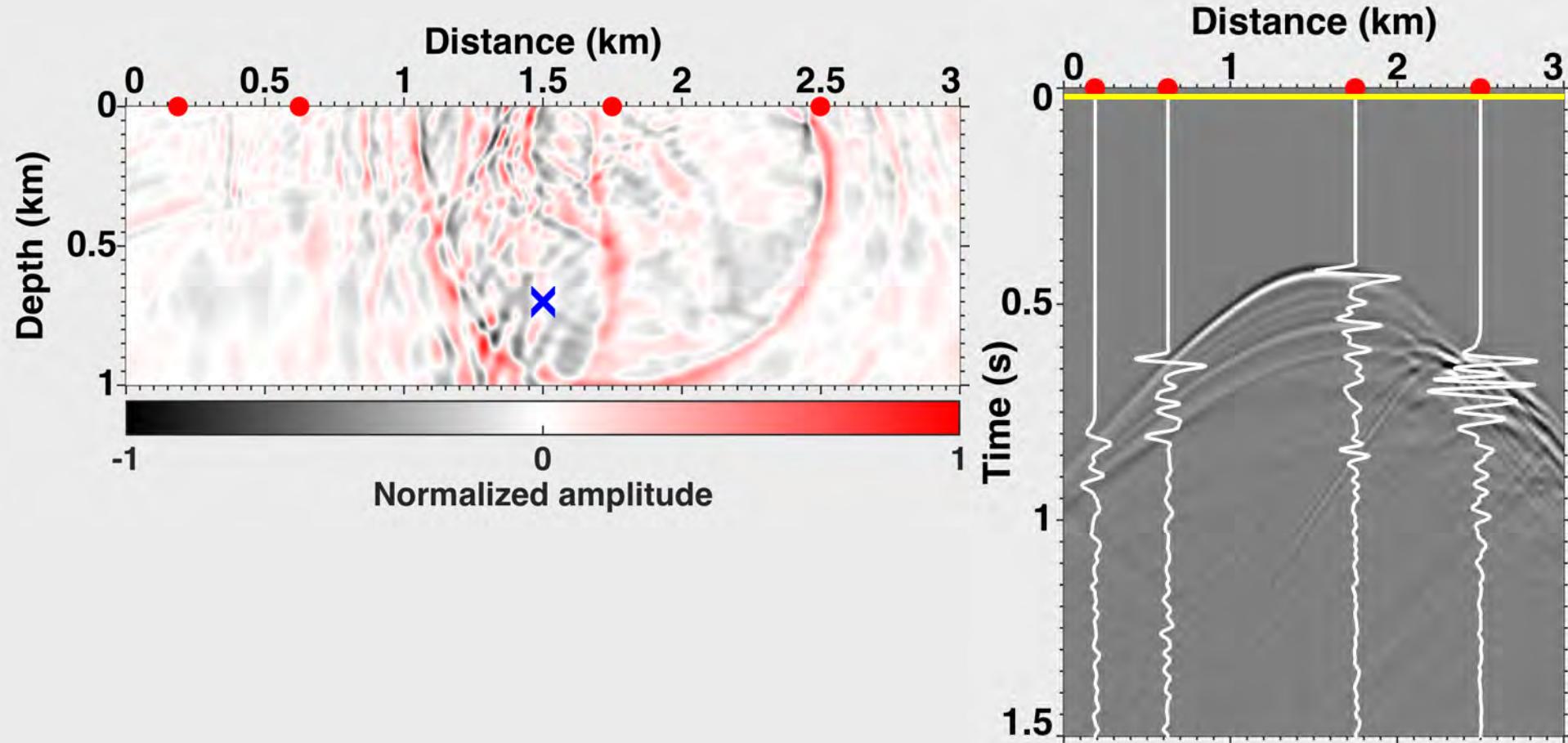
Time-reversal imaging



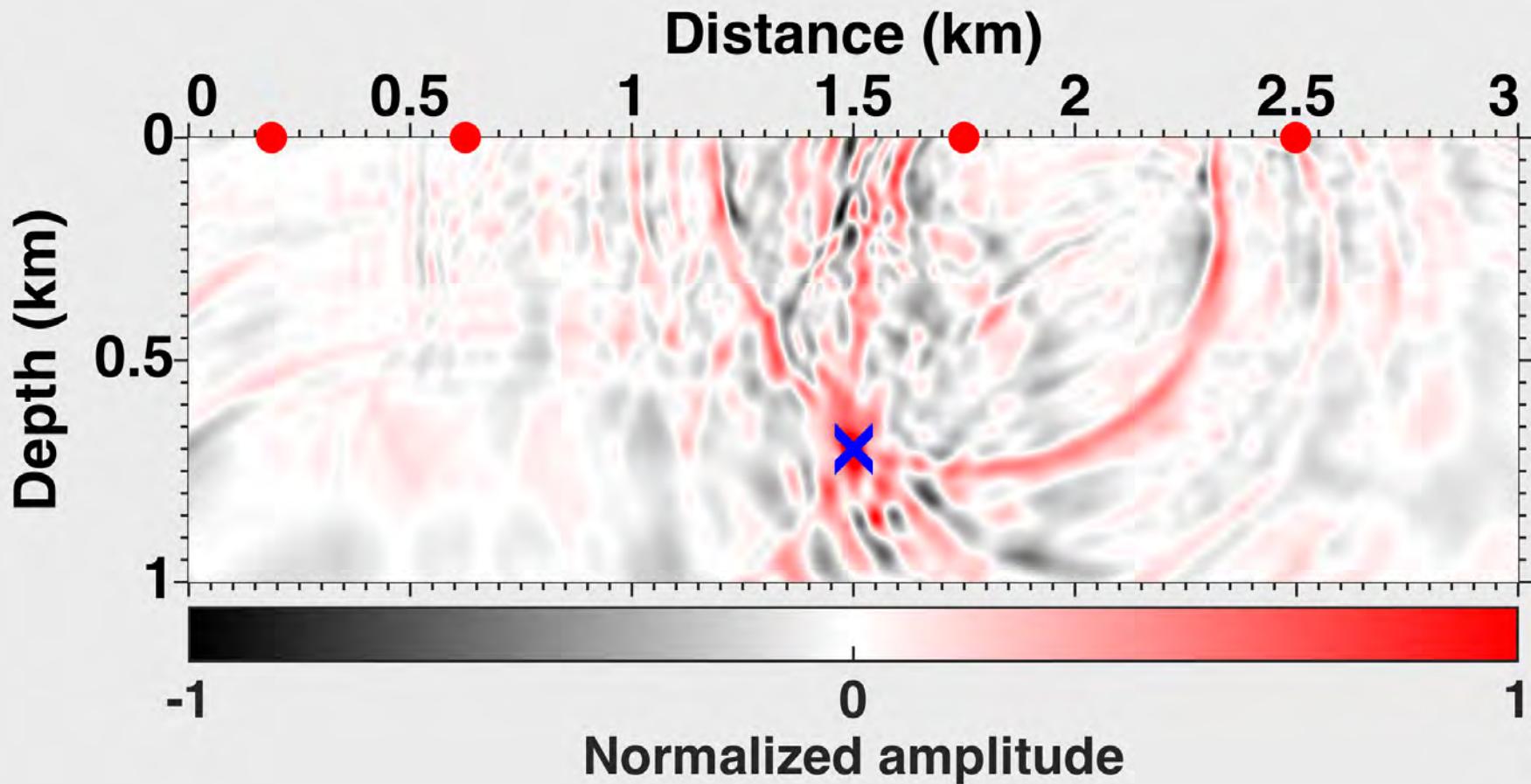
Time-reversal imaging



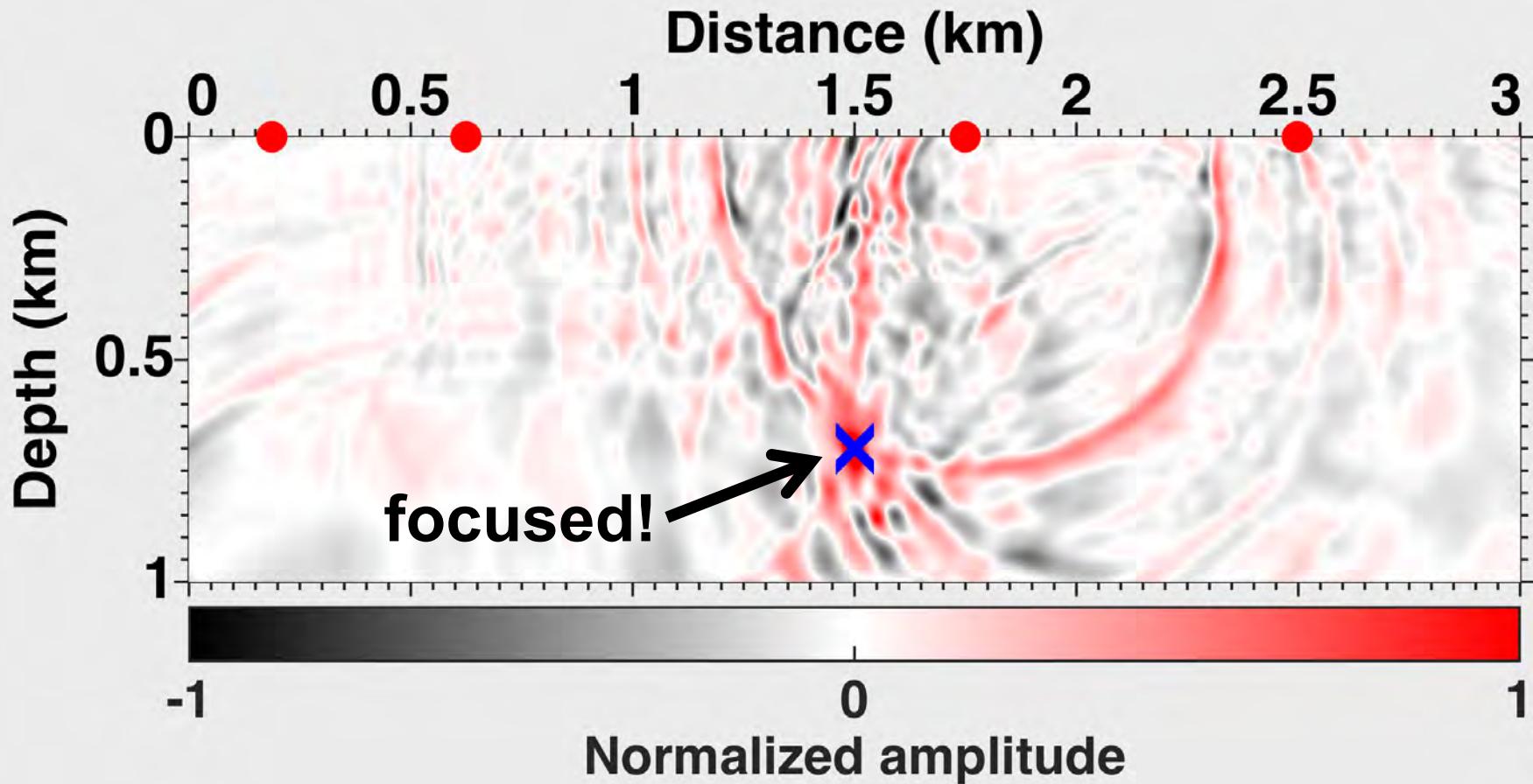
Time-reversal imaging



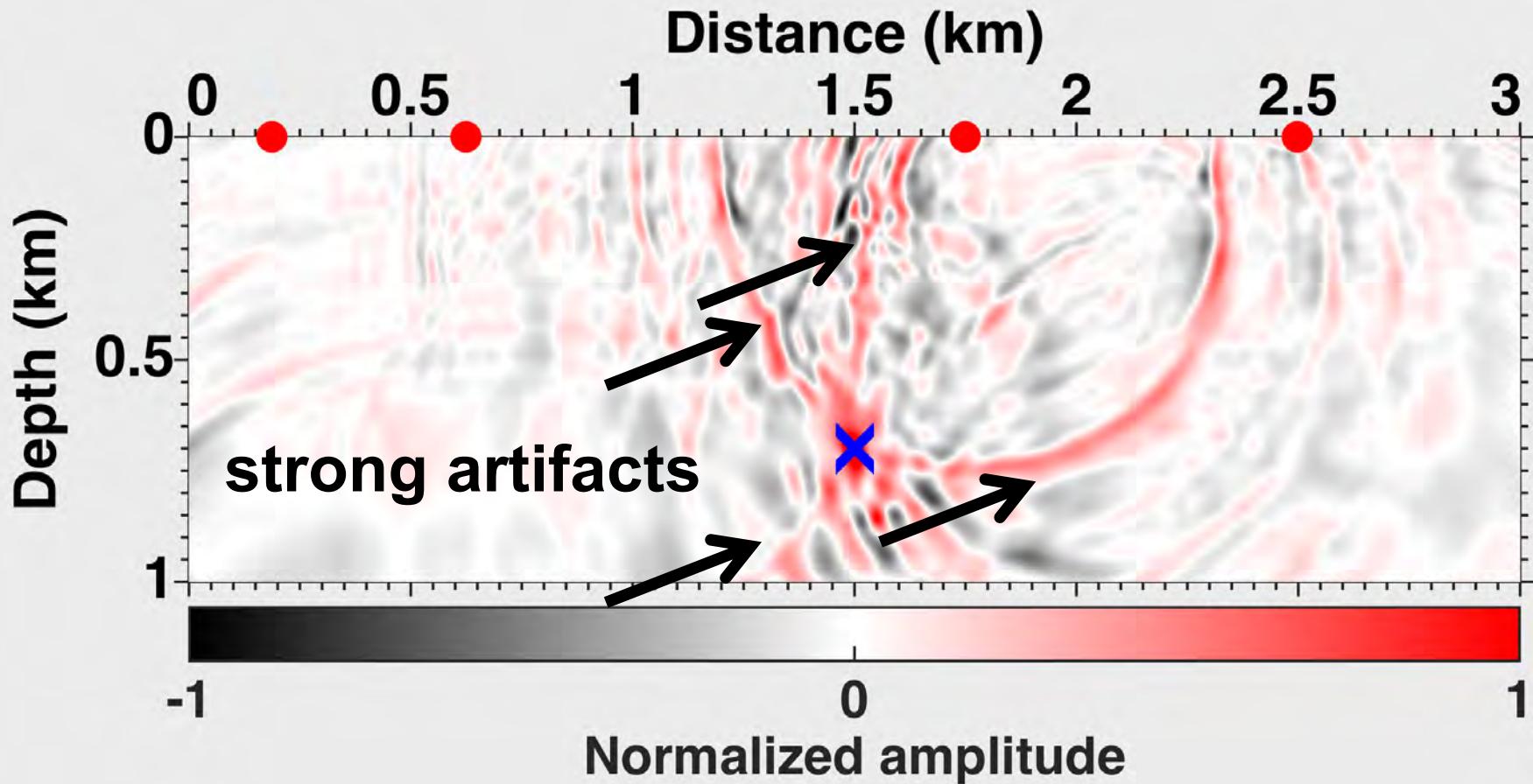
Time-reversal imaging



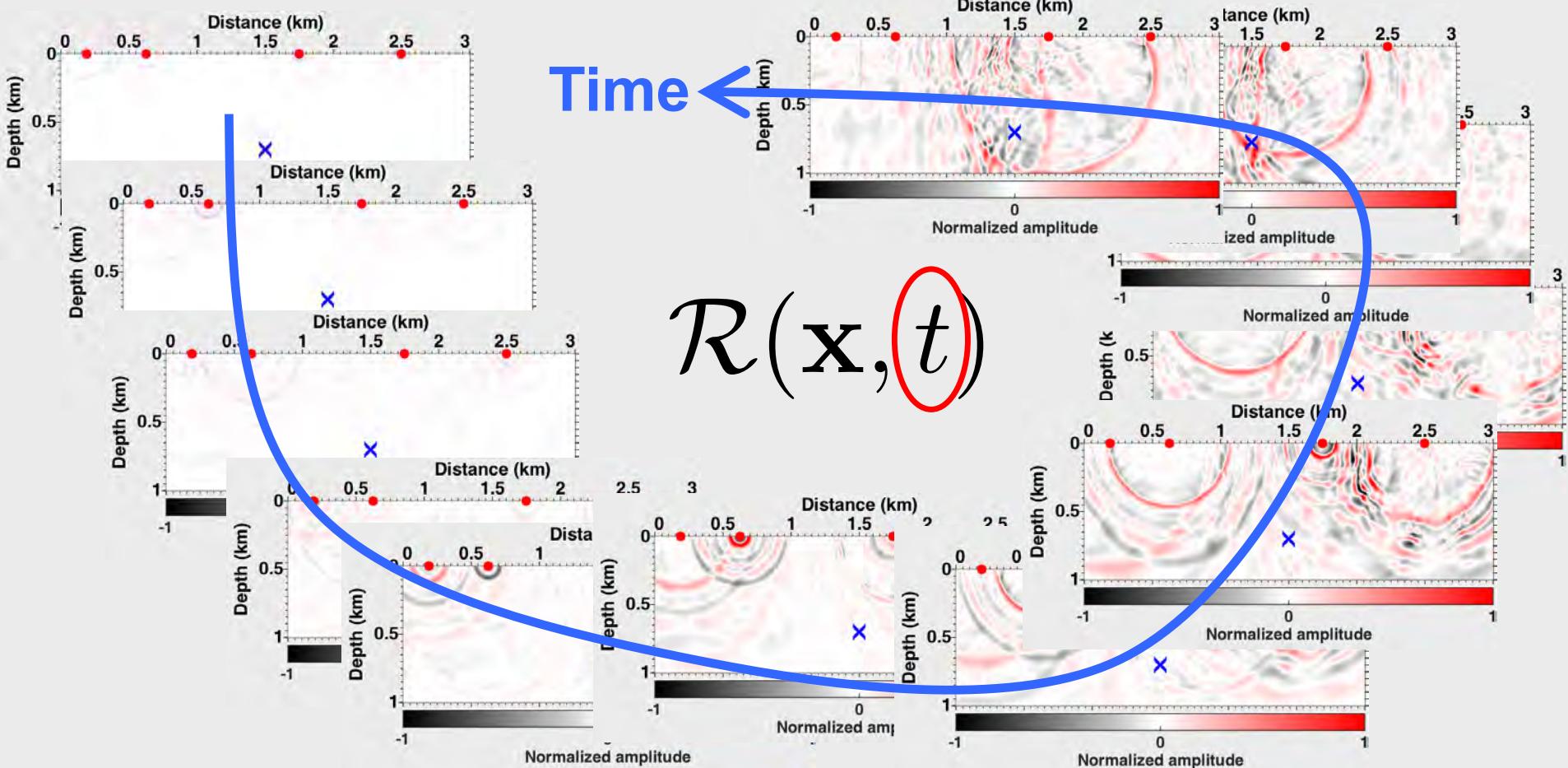
Time-reversal imaging



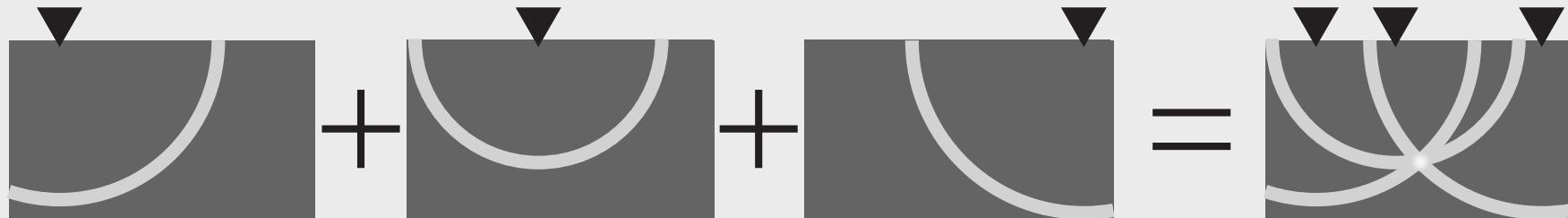
Time-reversal imaging



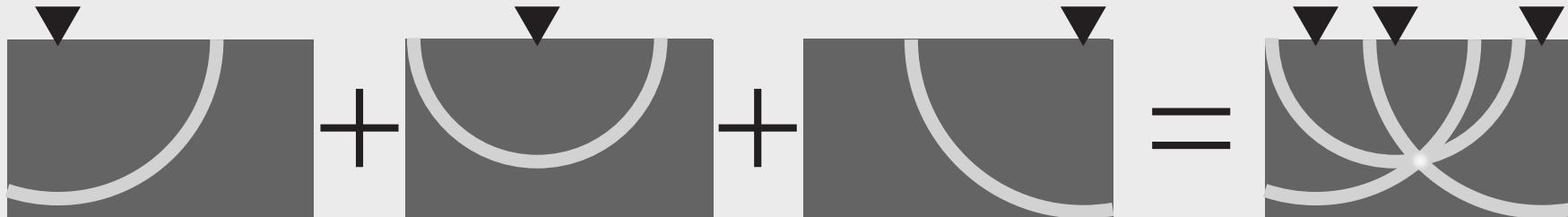
Time-reversal imaging



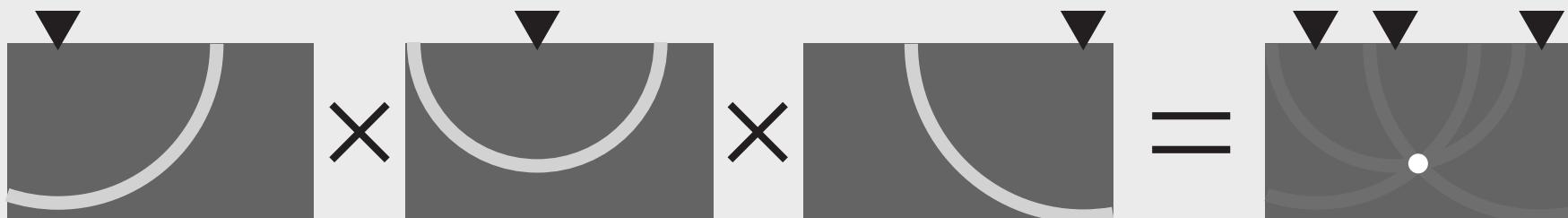
Time-reversal image (conventional method)



Time-reversal image (conventional method)



Geometric-mean RTM (new approach)



At time-lag = 0...,

Time-reversal imaging

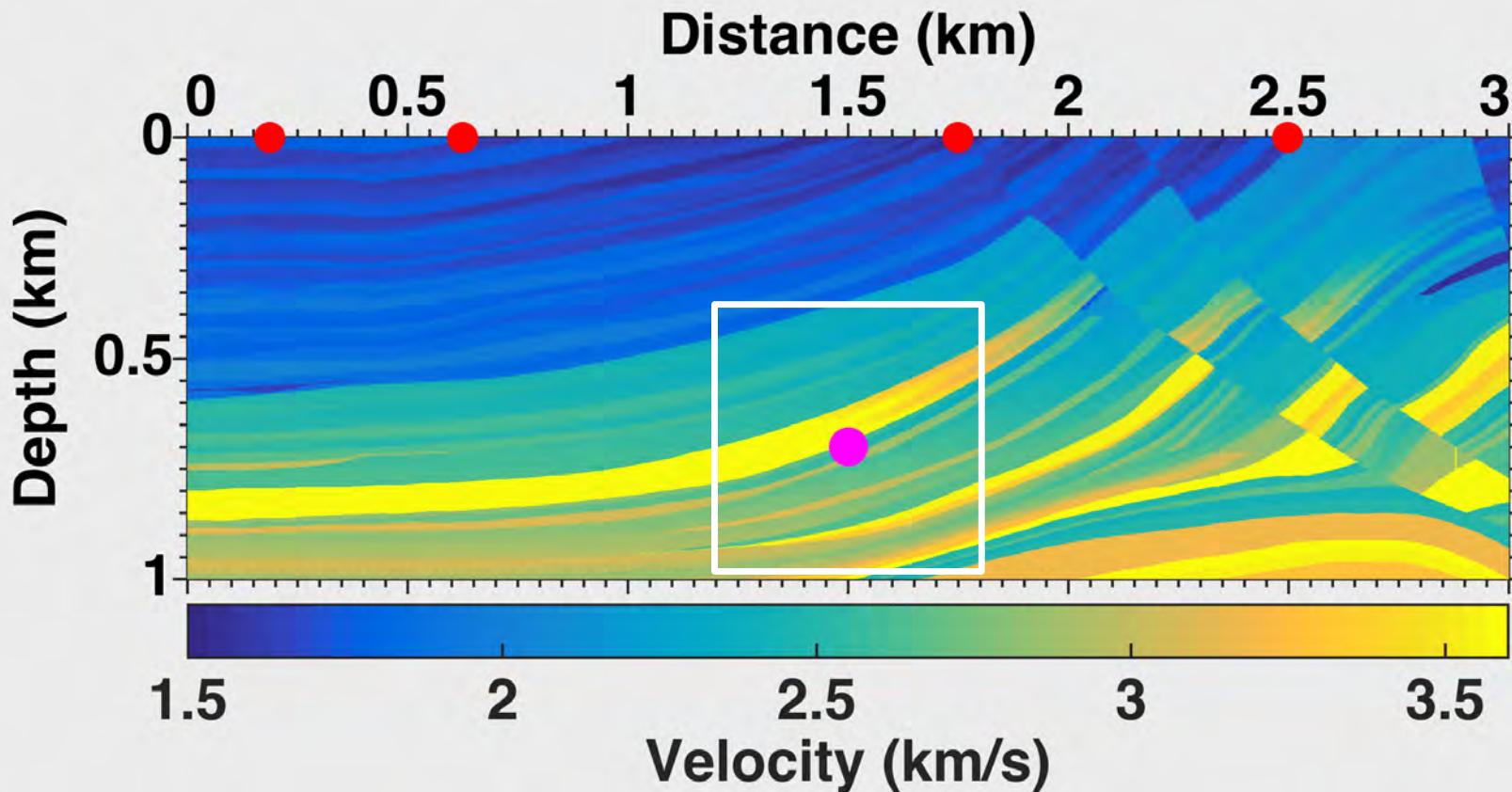
= Arithmetic-mean reverse-time imaging

$$R(x, t) = W_1(x, t) + W_2(x, t) + W_3(x, t) + W_4(x, t)$$

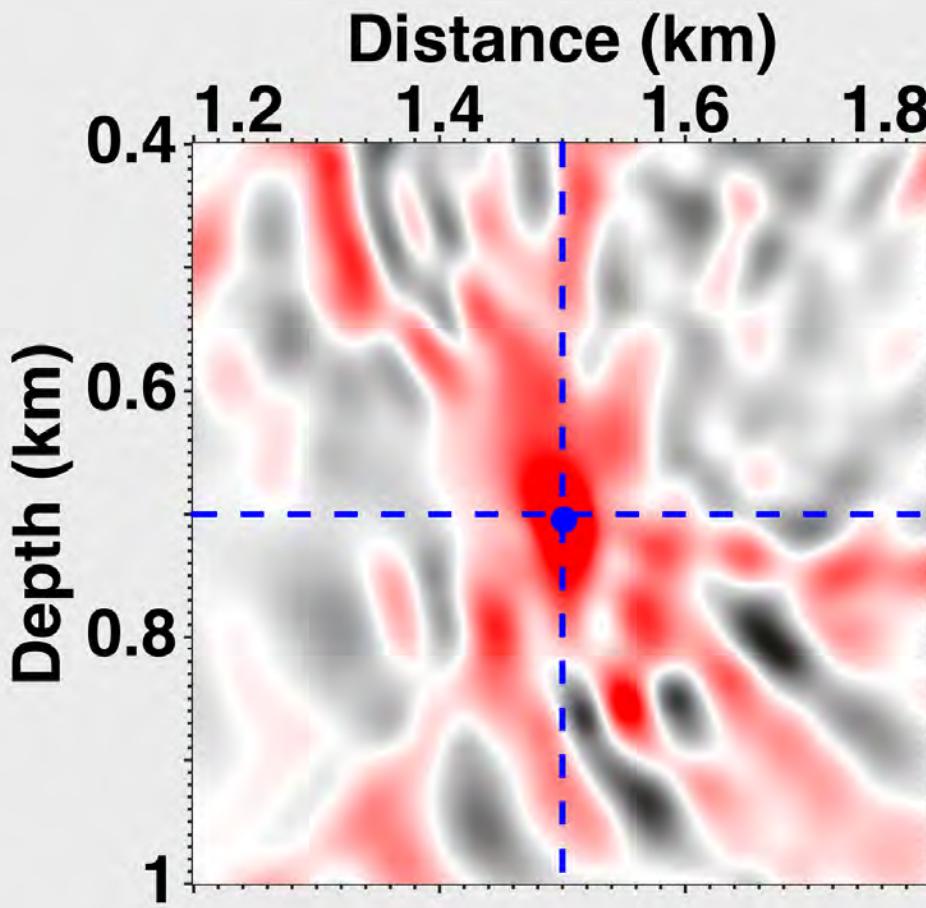
Geometric-mean reverse-time migration

$$R(x) = \sum_t W_1(x, t) \times W_2(x, t) \times W_3(x, t) \times W_4(x, t)$$

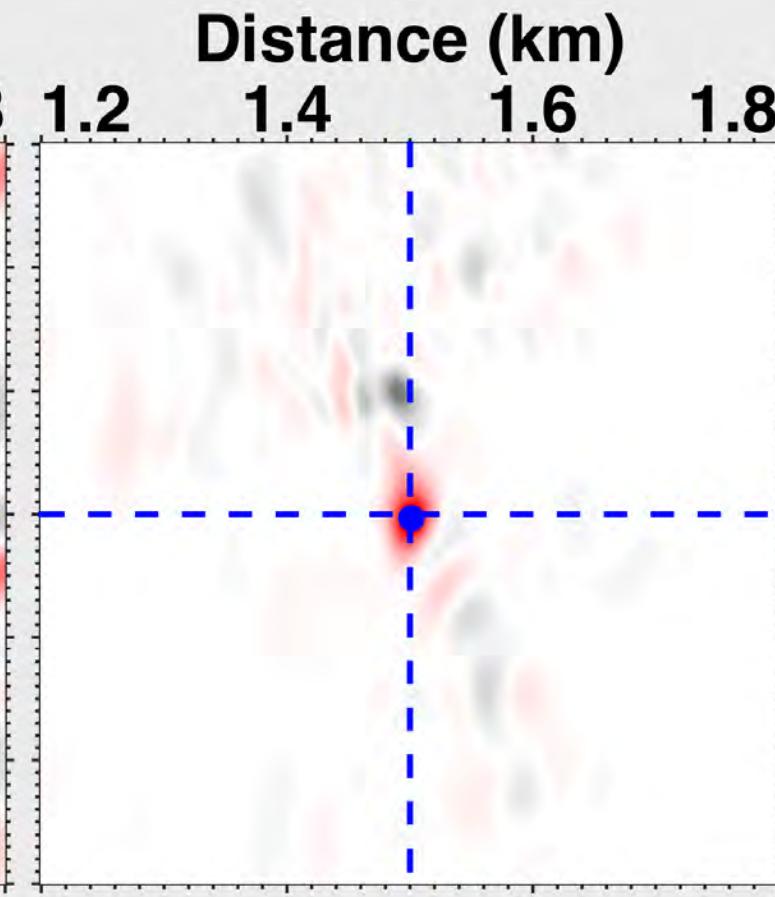
Comparison



Time reversal



GmRTM

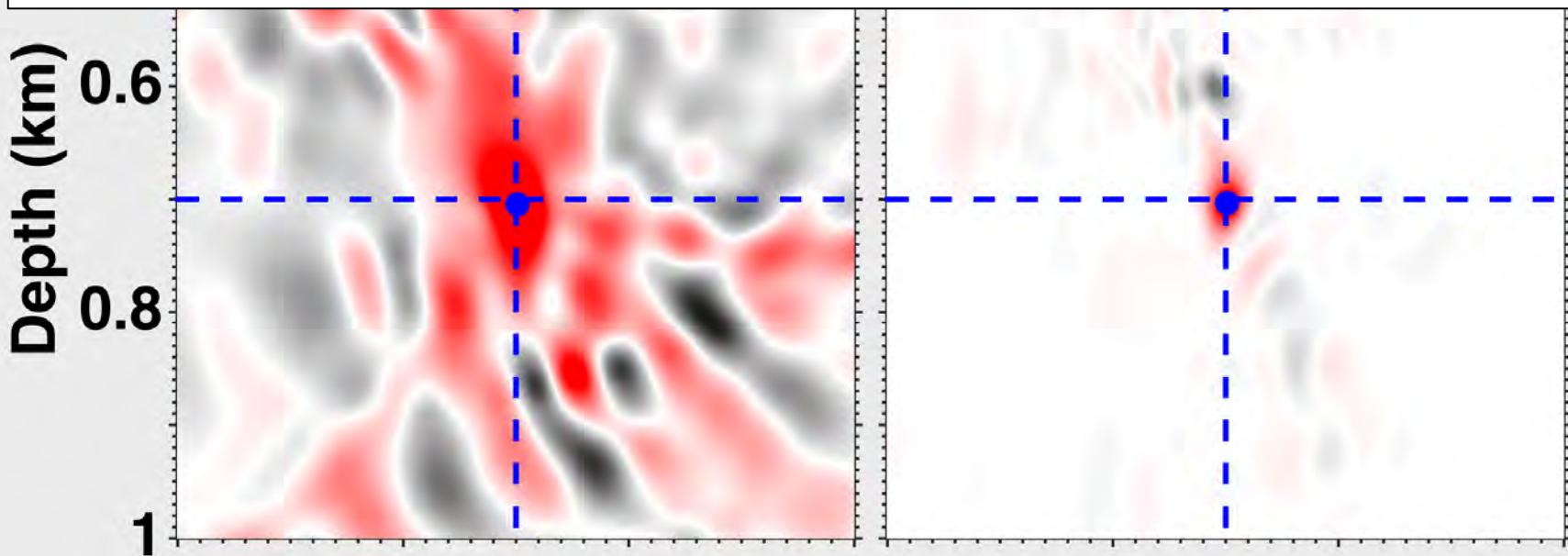


Time reversal

GmRTM

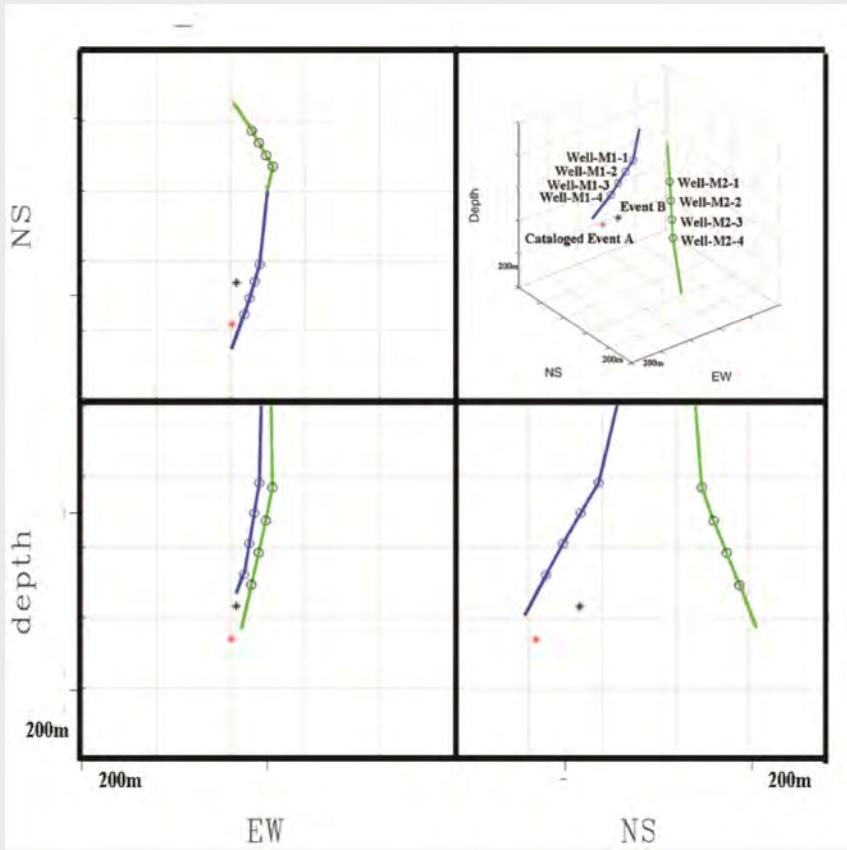
GmRTM is efficient for :

- locating with higher spatial resolution
- using long-duration signals (fracking, tremors...)
- enhancing weak signals (i.e., noisy signals)



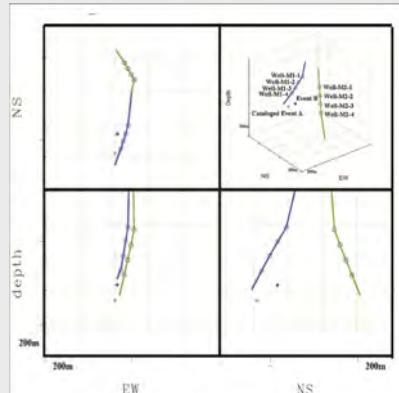
Field data example (Fracking)

An oil field in Japan

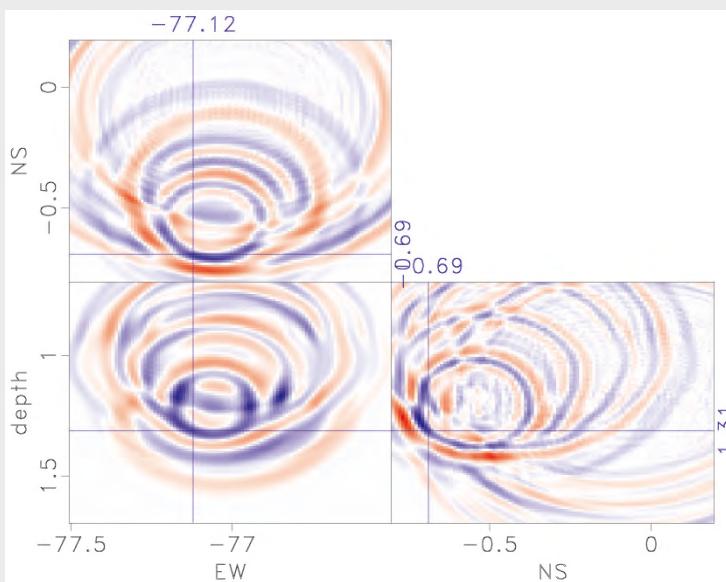


- Only 8 receivers available
- Challenging geometry

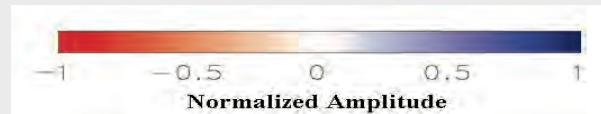
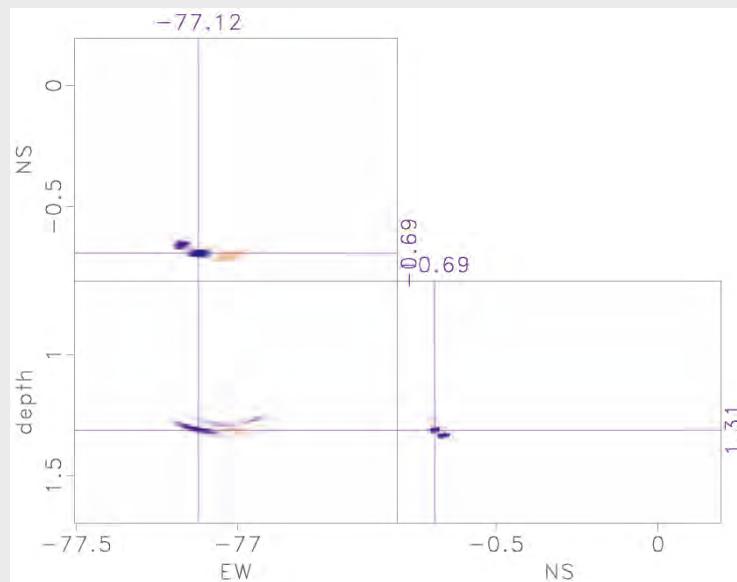
Field data example (Fracking)



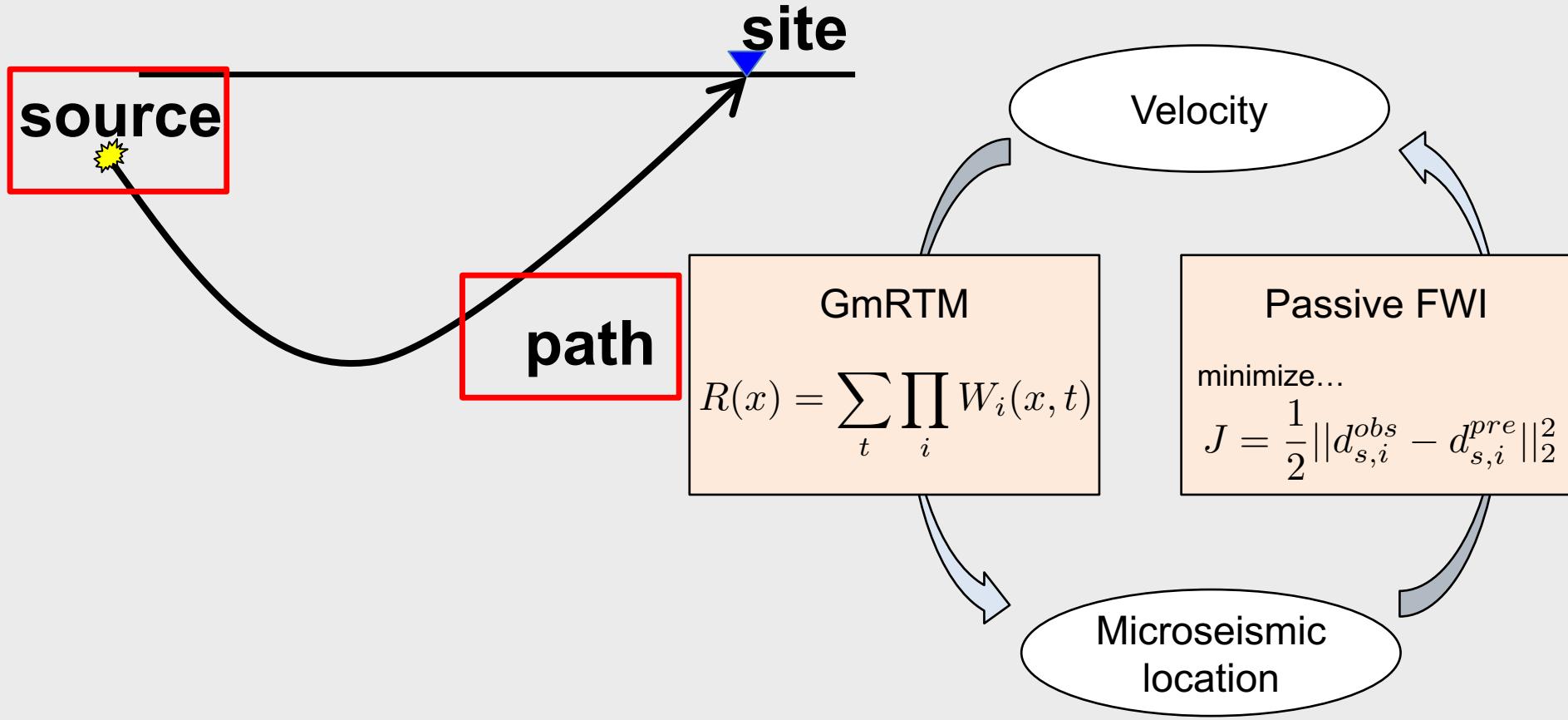
Time reversal



GmRTM

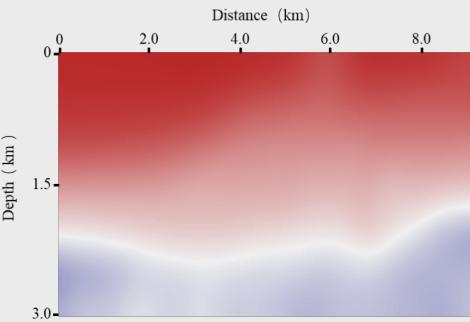


Velocity update with microseismic data



Velocity update with microseismic data

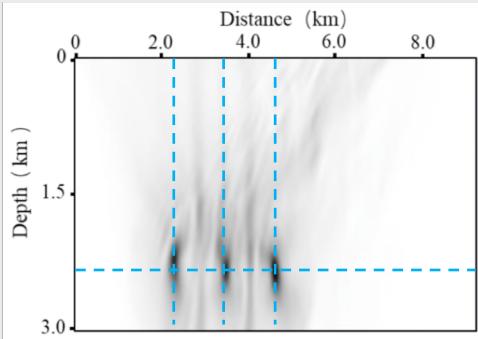
Initial velocity model



GmRTM

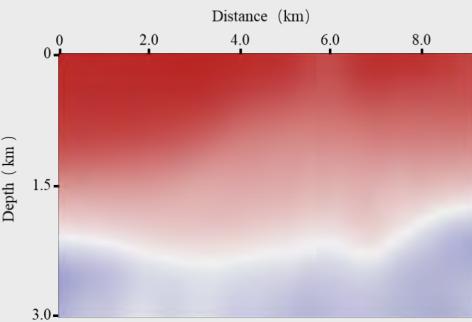
FWI

Initial source locations



Velocity update with microseismic data

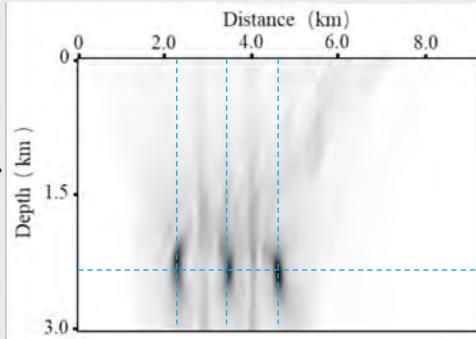
Initial velocity model



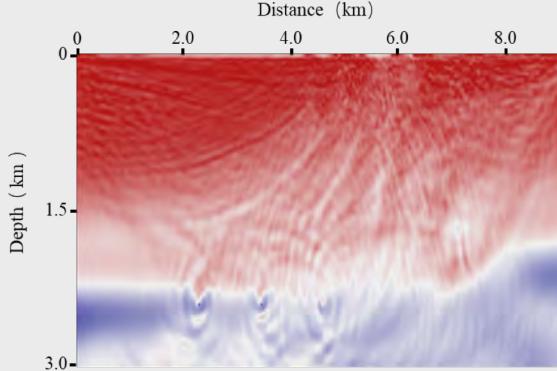
GmRTM

FWI

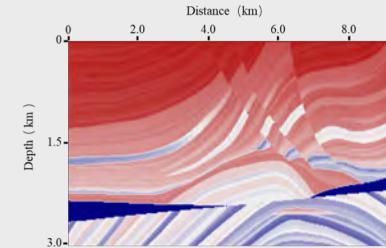
Initial source locations



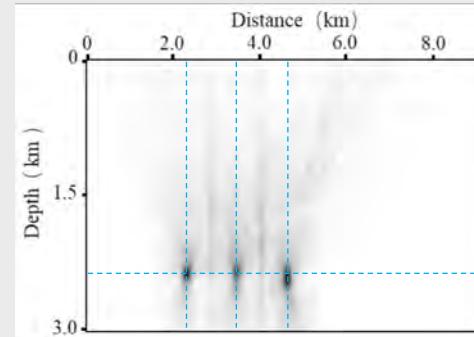
Inverted velocity model



True velocity model



Inverted source locations

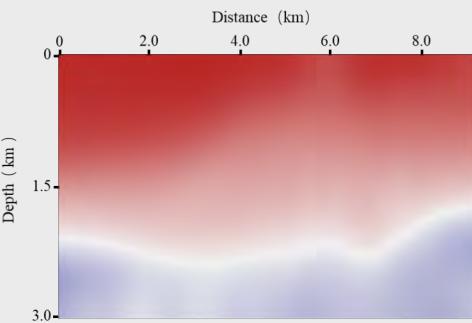


km/s

A vertical color bar indicating velocity in km/s. The scale ranges from 2.0 (dark red) to 5.0 (dark blue), with intermediate ticks at 3.0 and 4.0.

Velocity update with microseismic data

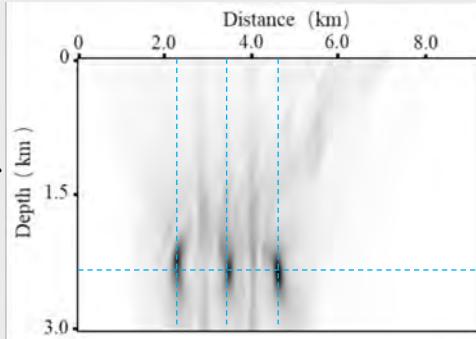
Initial velocity model



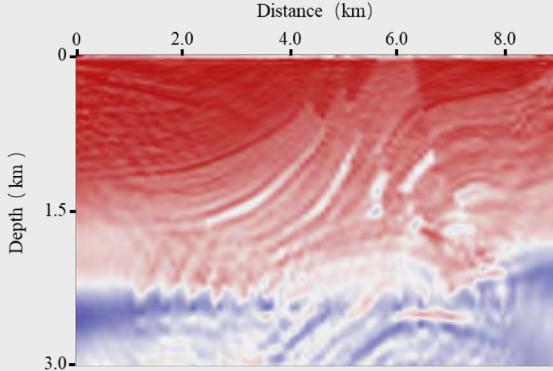
GmRTM

FWI

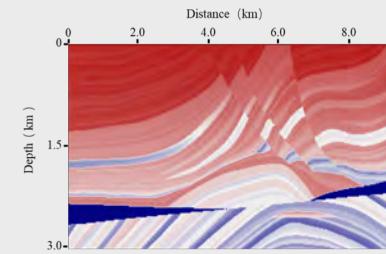
Initial source locations



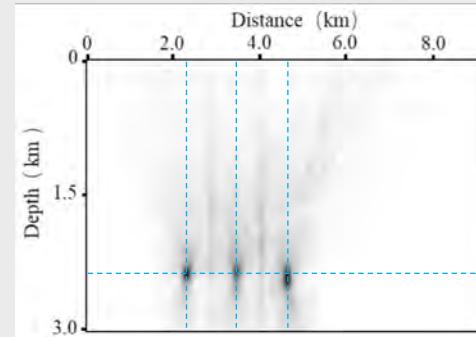
Inverted velocity model



True velocity model



Inverted source locations

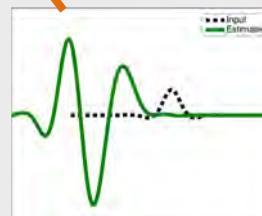
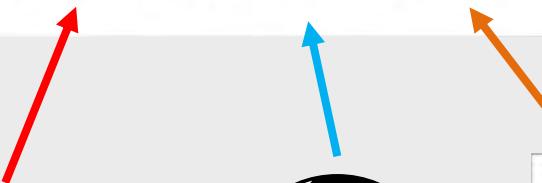


A vertical color bar indicating velocity in km/s. The scale ranges from 2.0 (dark blue) to 5.0 (dark red), with intermediate ticks at 3.0 and 4.0.

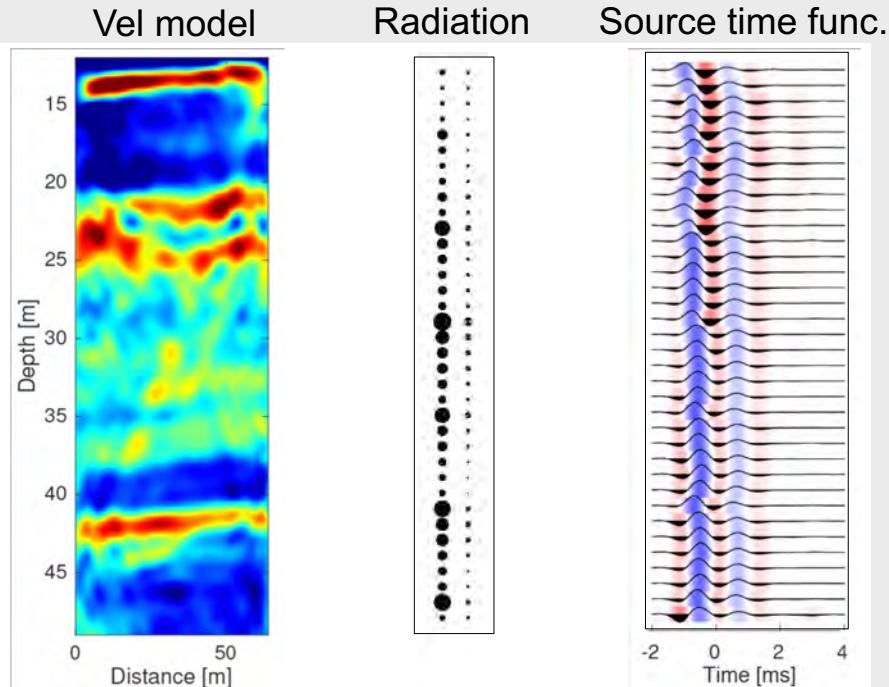
FWI for source parameter estimation

Cost function

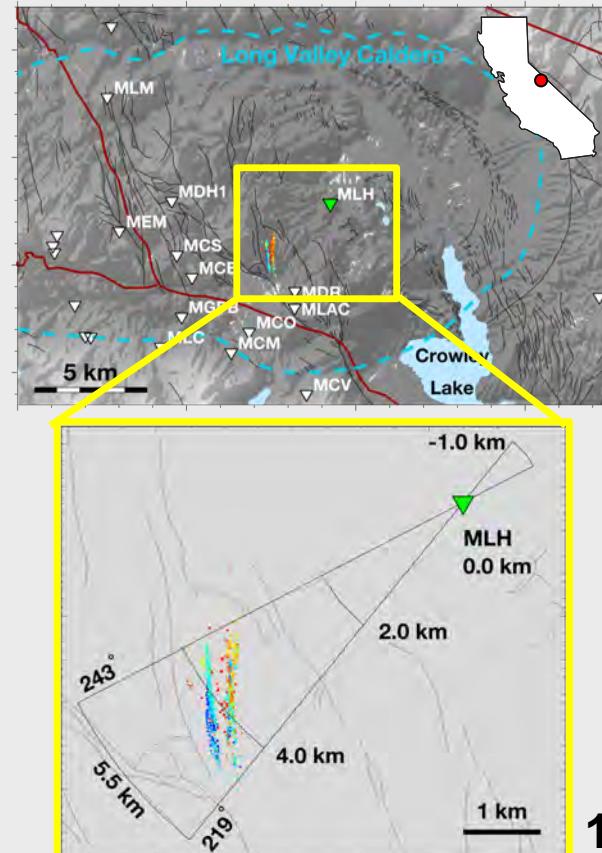
$$E(\mathbf{v}_p, \mathbf{v}_s, \sigma_{xx}, \sigma_{zz}, \sigma_{xz}, w) = \sum_{i=1}^{N_s} \sum_{j=1}^{N_r} \sum_{k=1}^{N_t} (d_{mod}(\mathbf{v}_p, \mathbf{v}_s, \sigma_{xx}, \sigma_{zz}, \sigma_{xz}, w) - d_{obs})^2$$



Field data example (crosswell active survey)

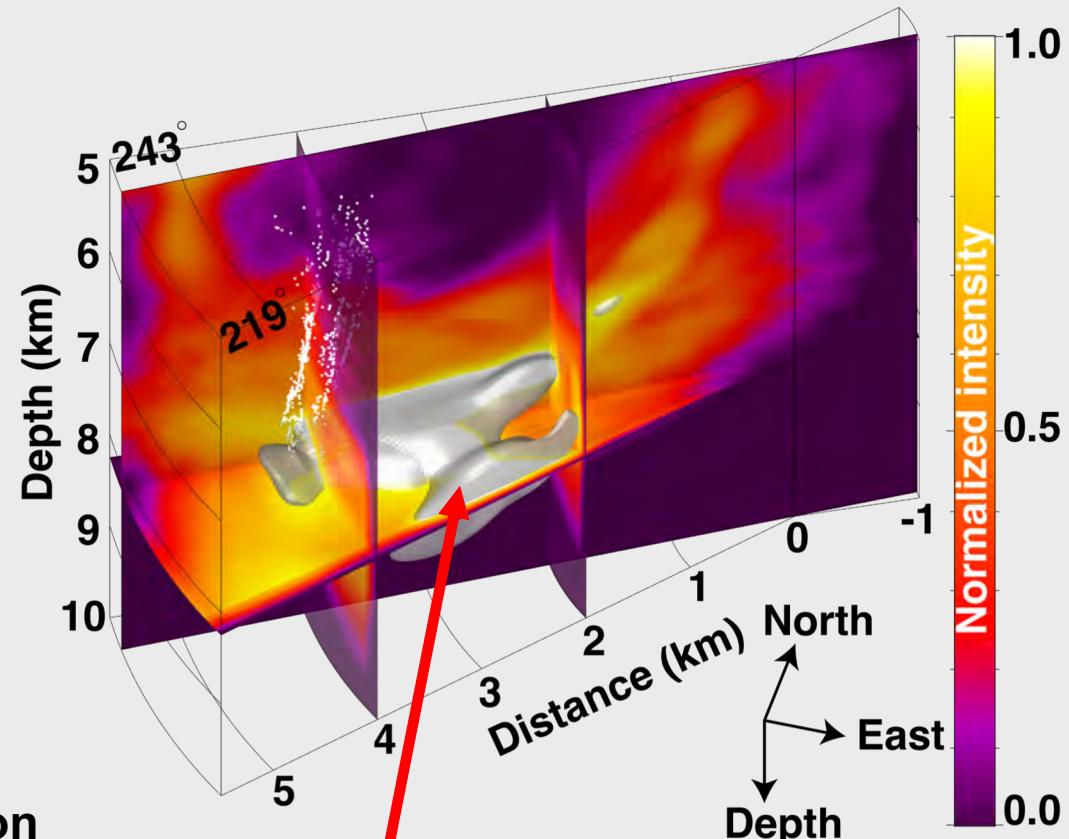


Structure imaging with small earthquakes

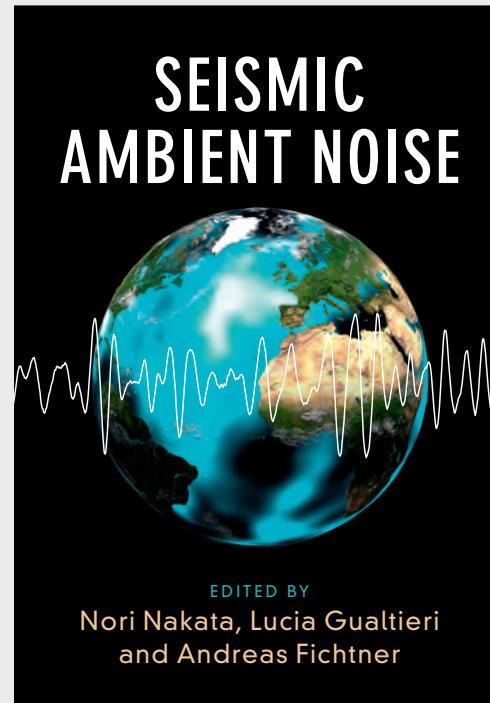
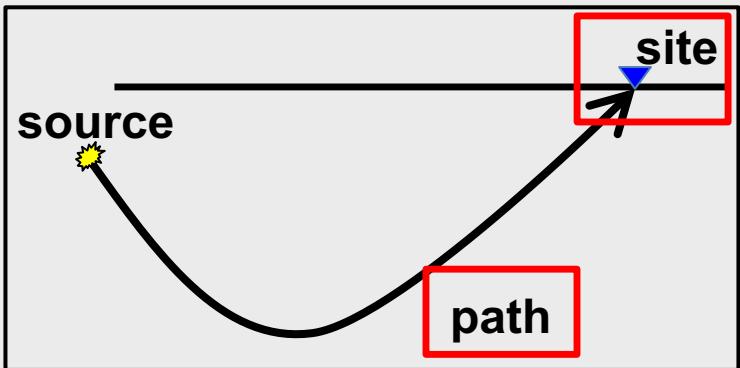


1 station
800 earthquakes

Magma roof



Ambient noise



A book about
Seismic Ambient Noise

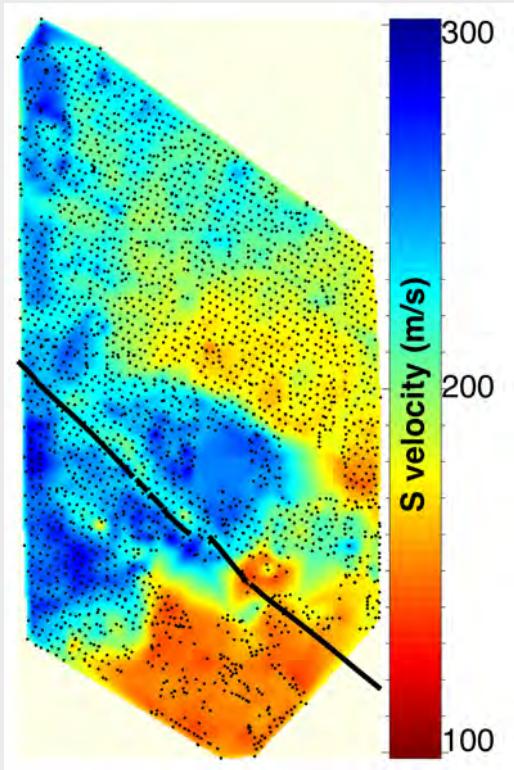
Cambridge University Press

370 pp

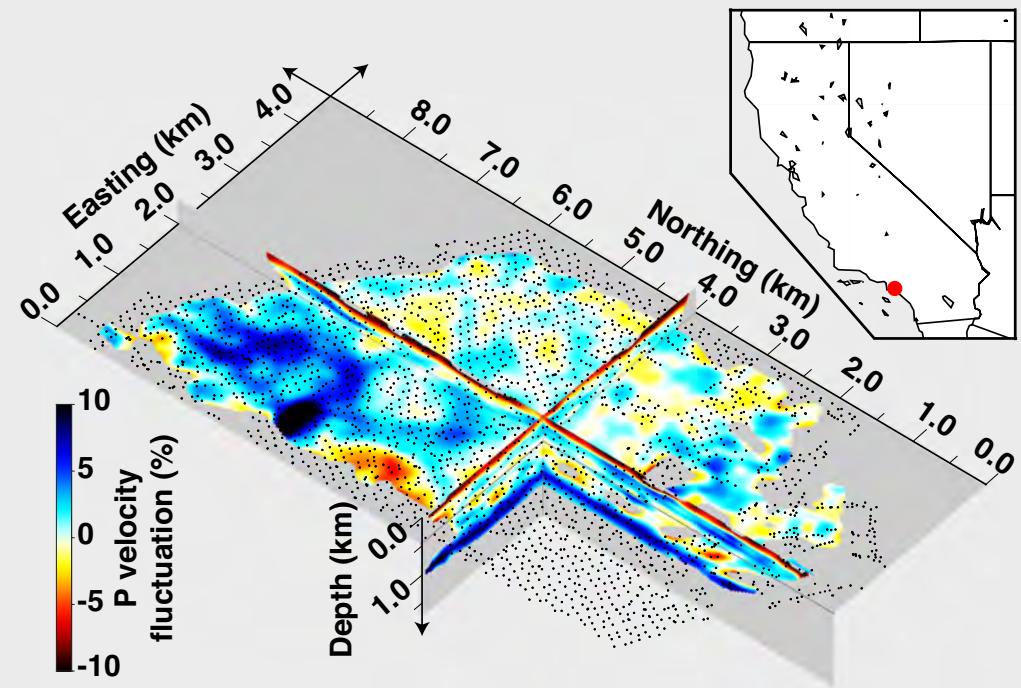
Published on May 2, 2019

Velocity estimation

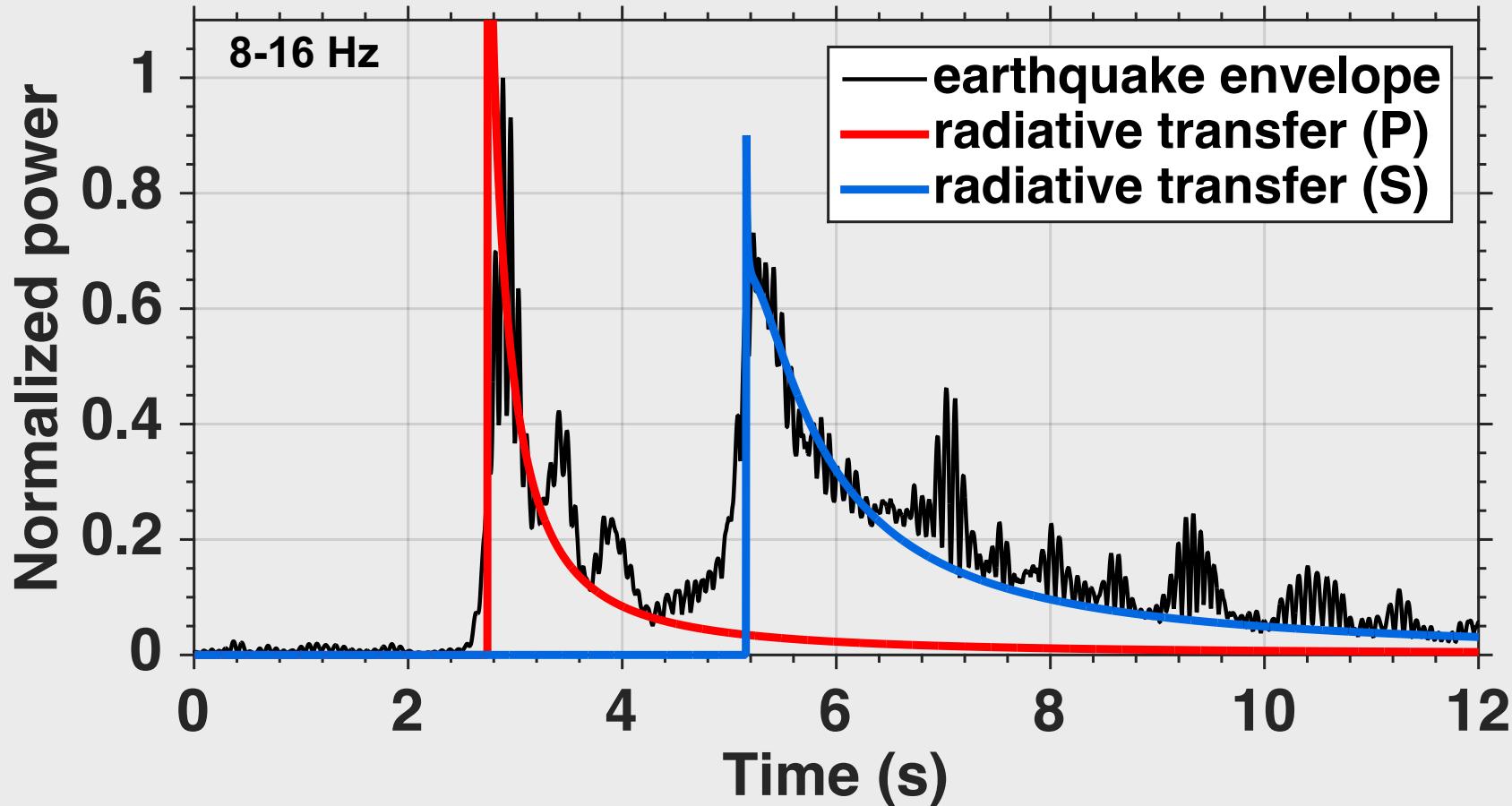
Near-surface
S velocity



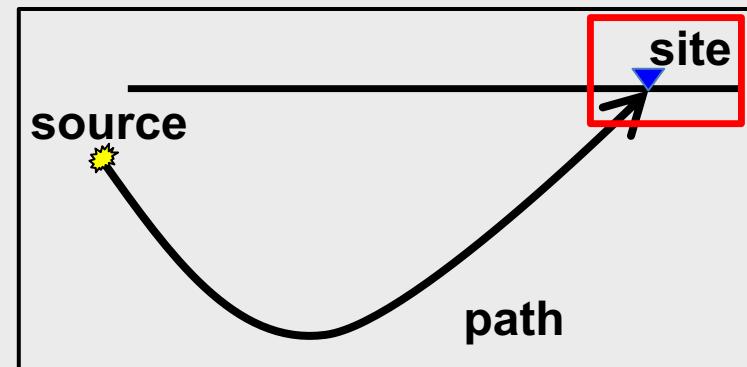
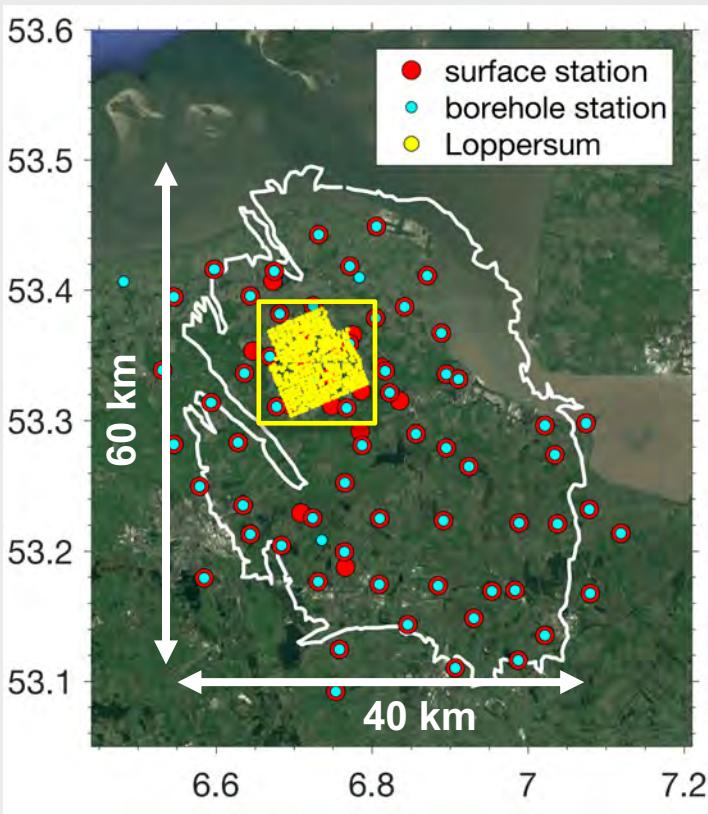
3D P velocity



High-frequency ground motion prediction

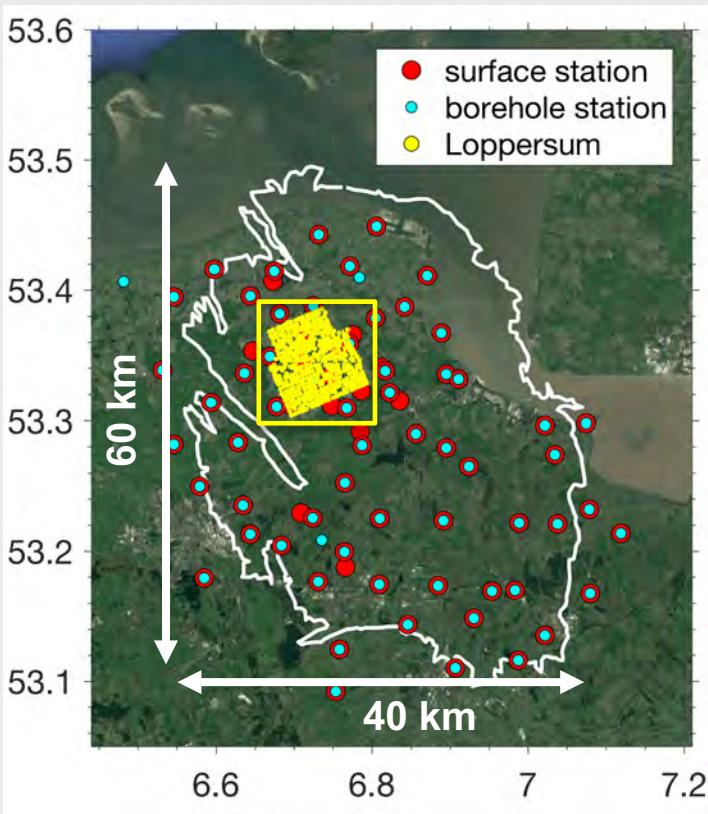


Groningen Gas Field



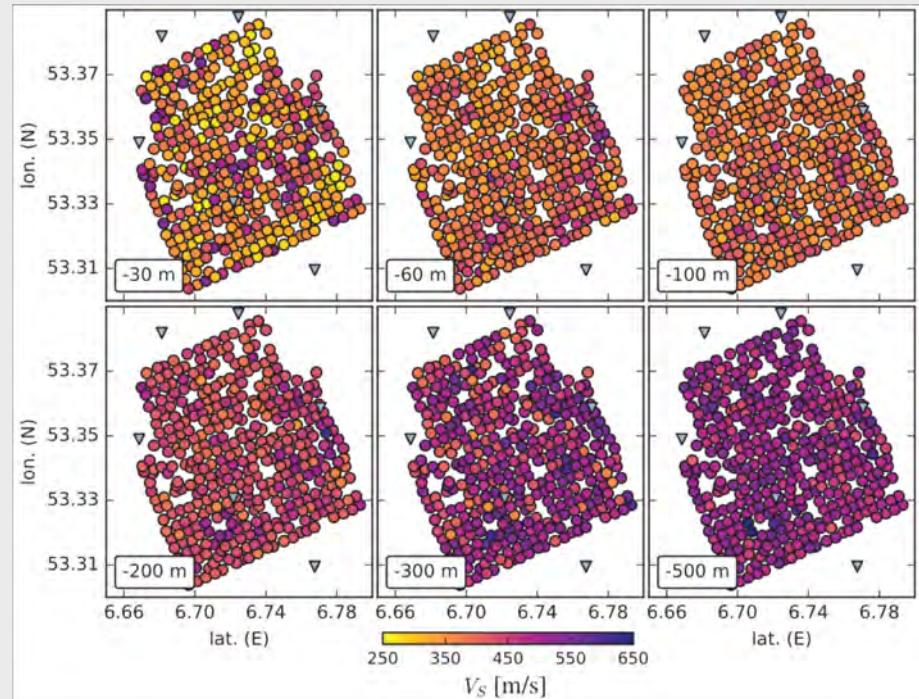
- **415 stations**
- **3 components**
- **Receiver spacing: 400 m**
- **One-month continuous record**

Groningen Gas Field

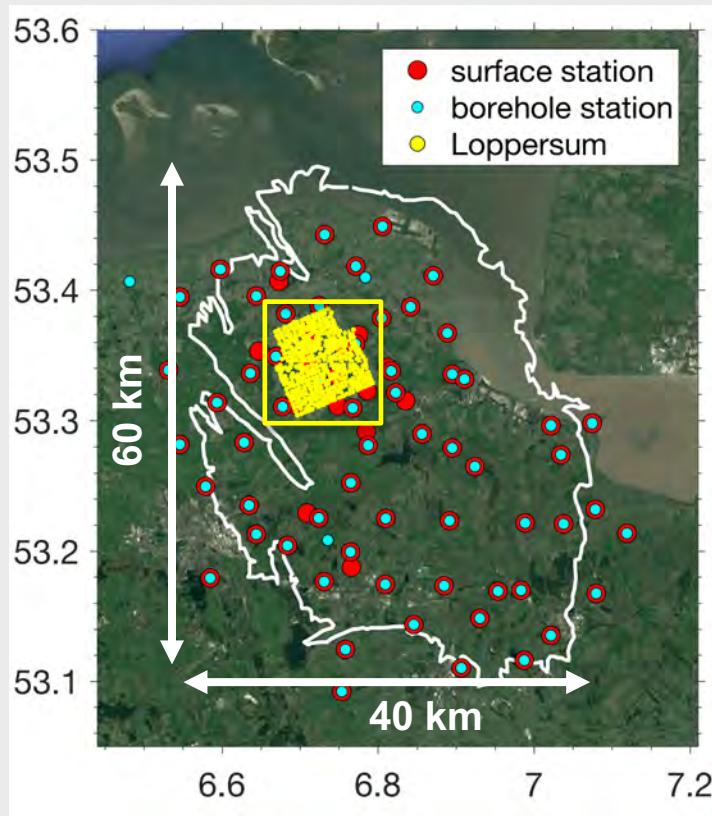


Near-surface velocity model estimated from

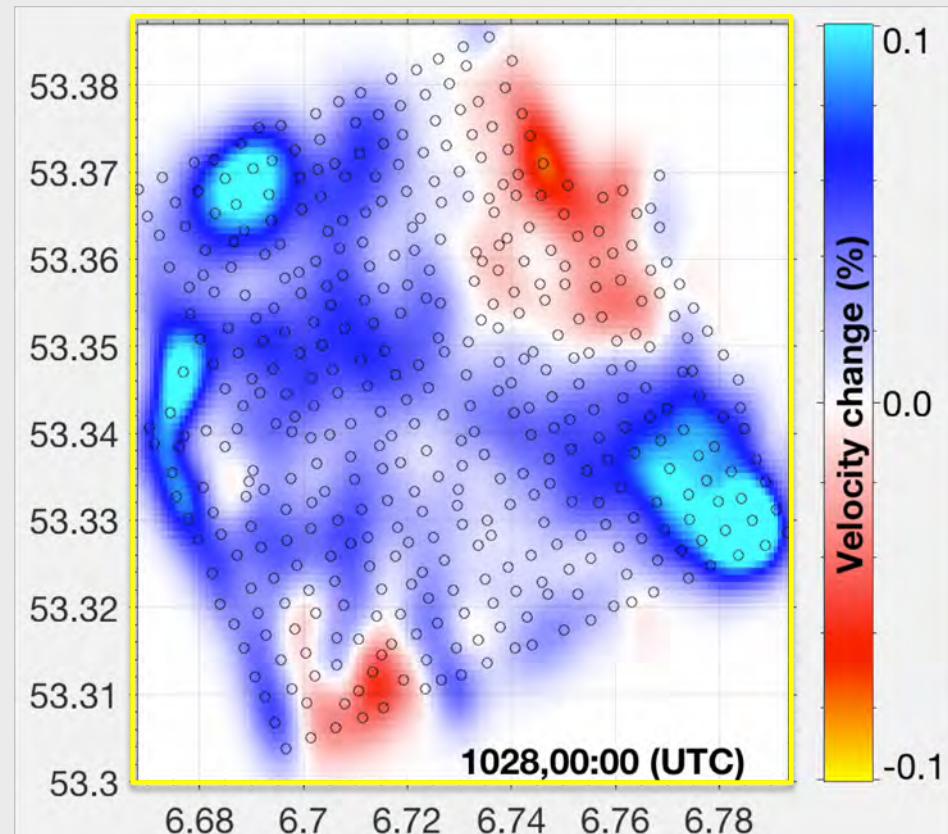
- H/V
- Rayleigh and Love waves



Groningen Gas Field

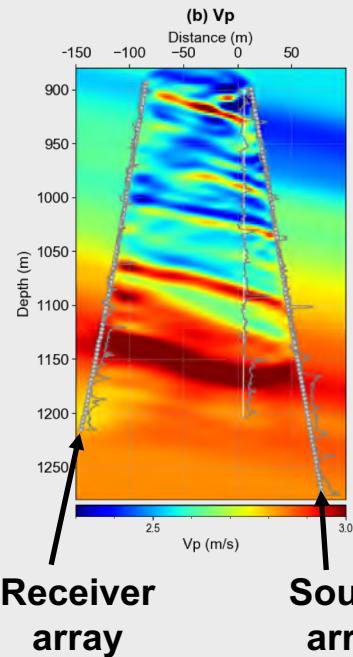


Hourly seismic velocity change

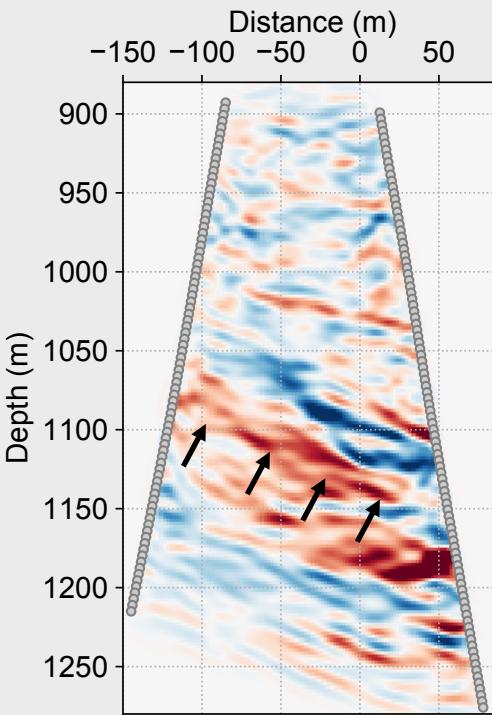


Time-lapse FWI (CO_2 monitoring)

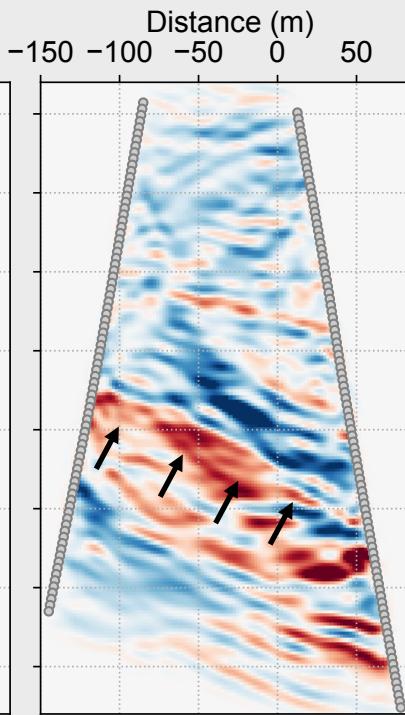
Baseline



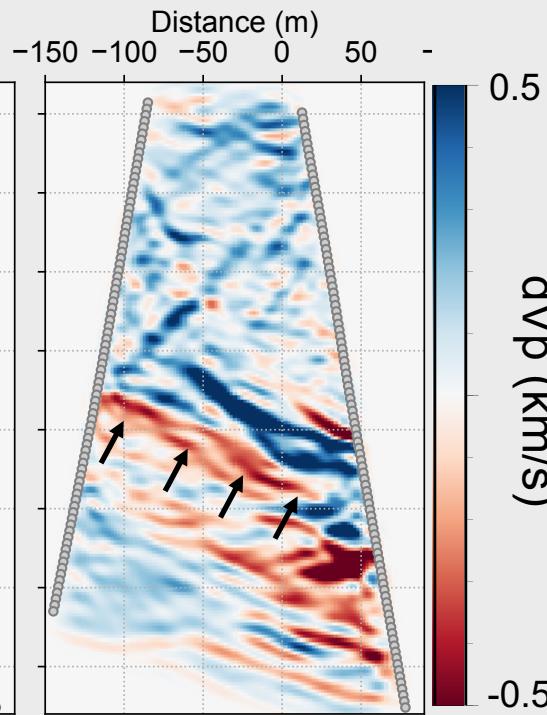
**MN1 – Baseline
(3000 t injected)**



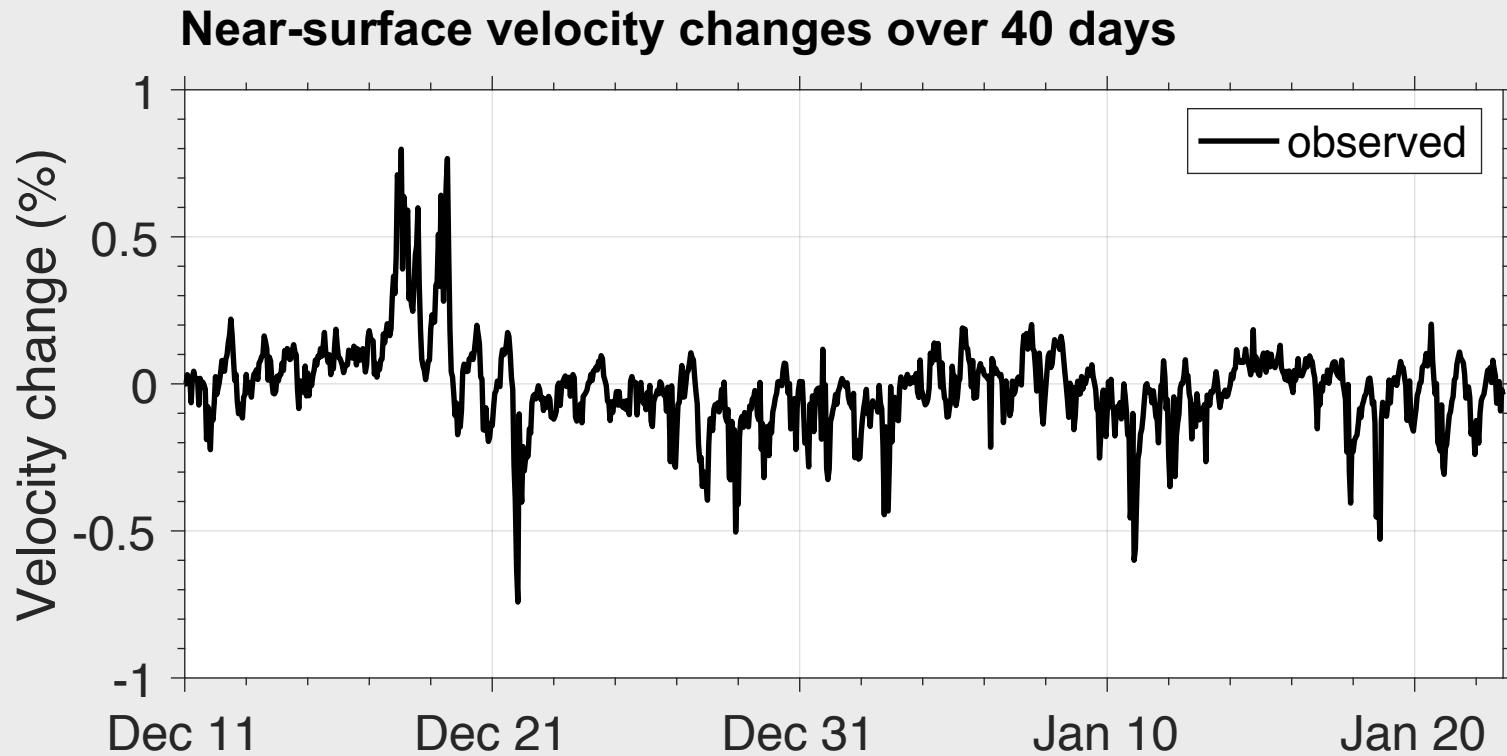
**MN2 – Baseline
(6000 t injected)**



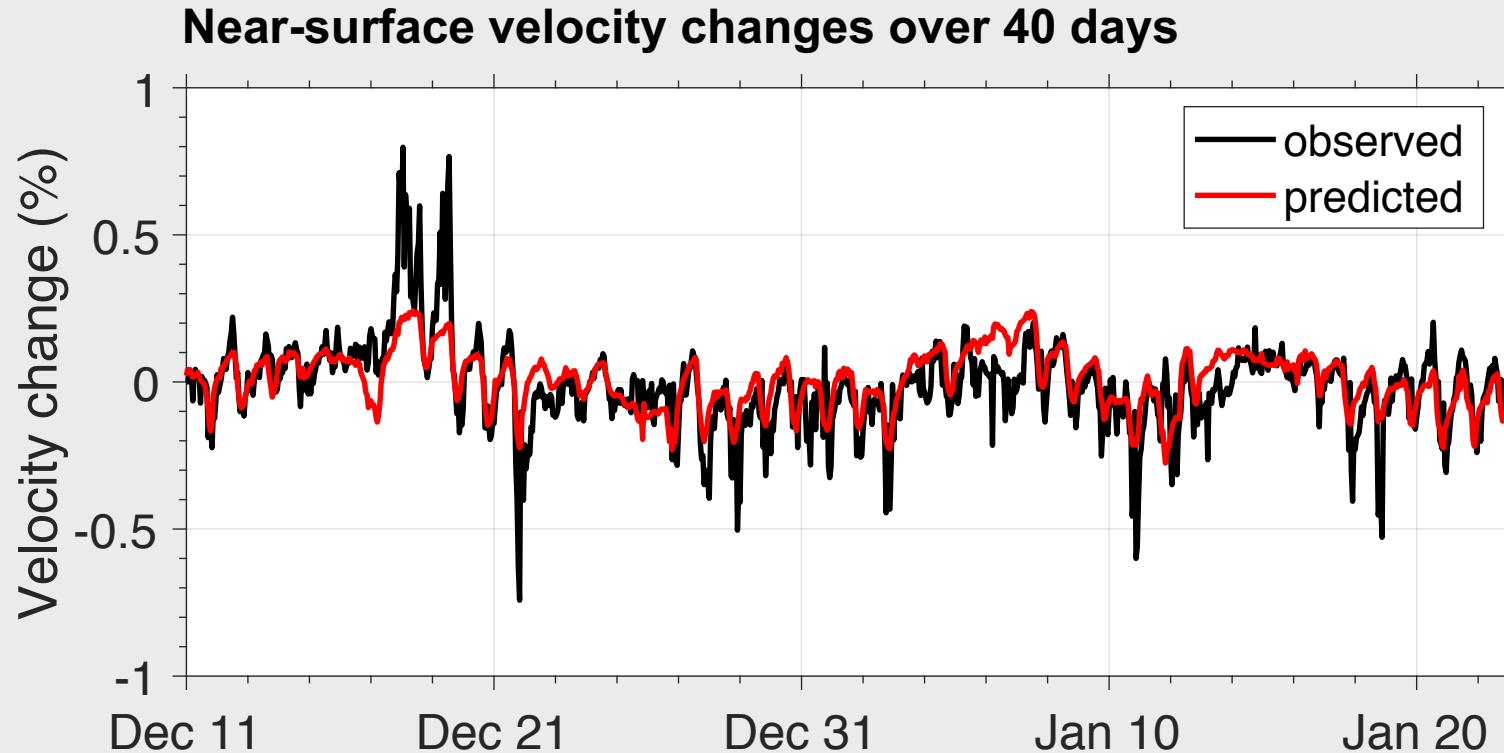
**MN3 – Baseline
(9000 t injected)**



Velocity change prediction by Machine learning

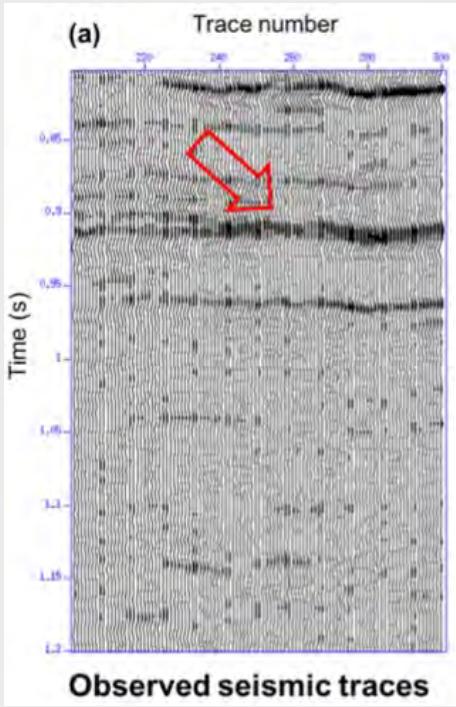


Velocity change prediction by Machine learning



Prediction using machine-learning regression analysis
Input data: first 20 days & environmental parameters

Reflectivity inversion



$$s(t) = h(t) * r(t)$$

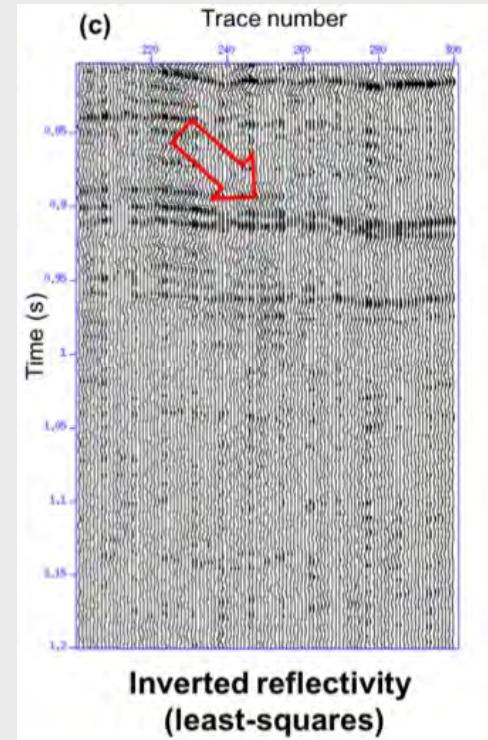
↑ ↑ ↙

source wavelet reflectivity

observed data

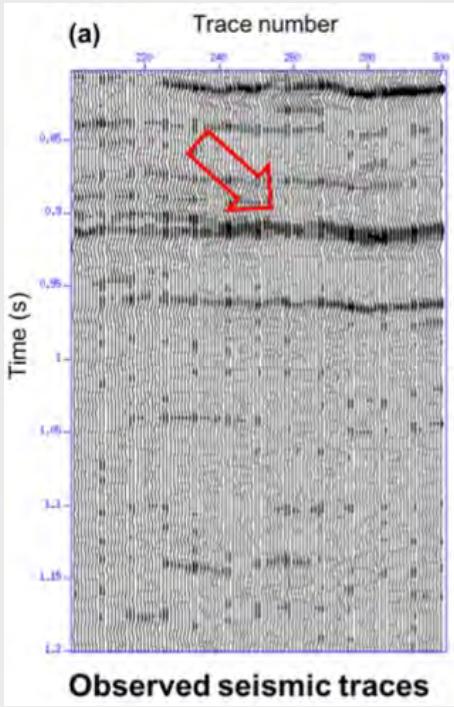
Reflectivity inversion based on conventional deconvolution (least-squares inversion)

$$L = \frac{1}{2} \| s(t) - h(t) * r'(t) \|_2^2$$



$$r'(t)$$

Reflectivity inversion with Machine Learning



$$s(t) = h(t) * r(t)$$

↑ source wavelet
observed data

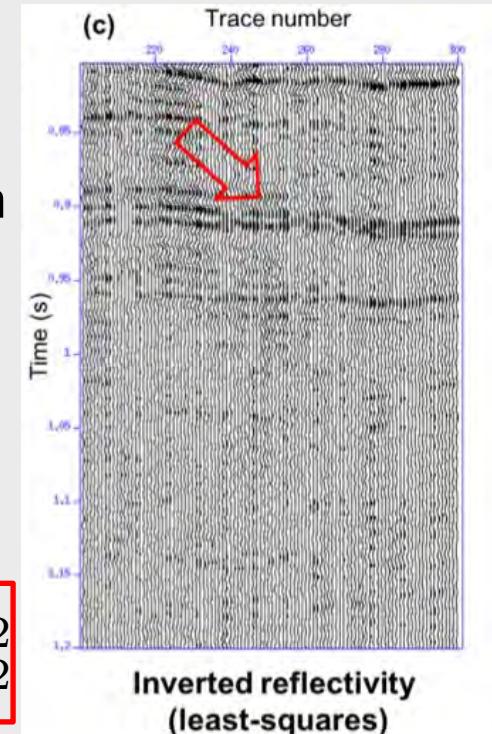
Reflectivity inversion based on conventional deconvolution (least-squares inversion)

$$L = \frac{1}{2} \| s(t) - h(t) * r'(t) \|_2^2$$

$$L = \frac{1}{2} \| r_s(t) - h_{\Theta}^{\dagger}(t) * s_s(t) \|_2^2$$

synthetic data

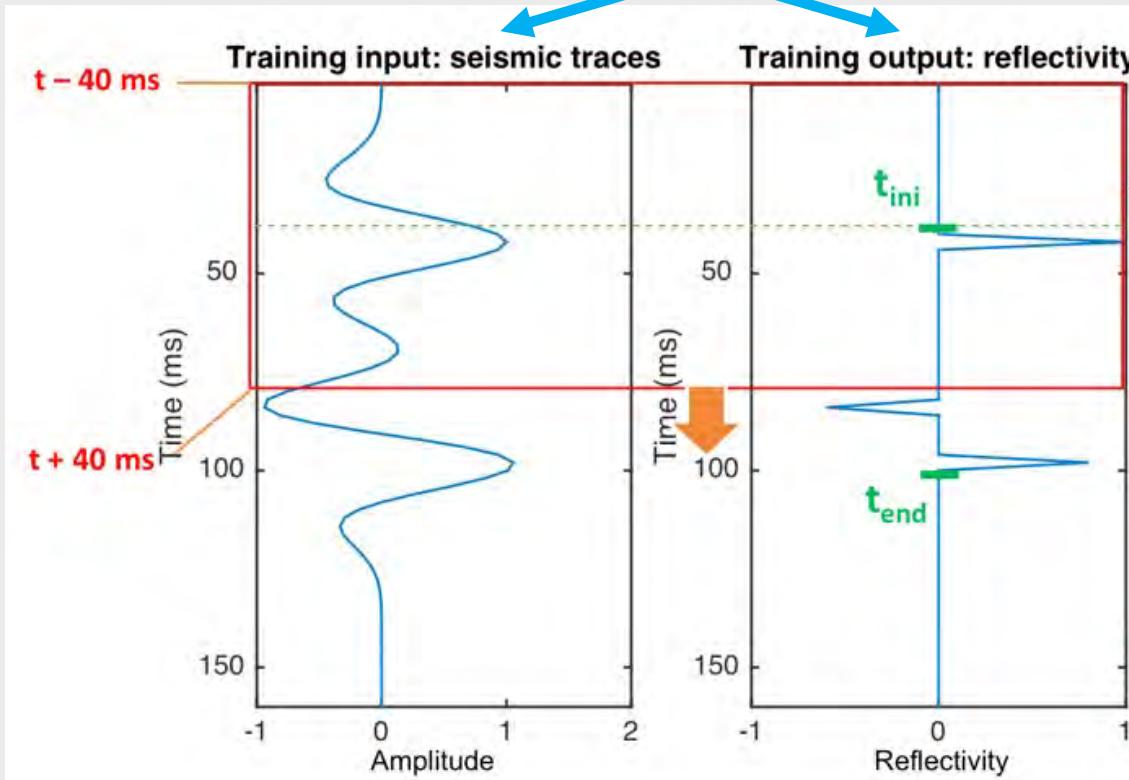
Mapping operator



$$r'(t)$$

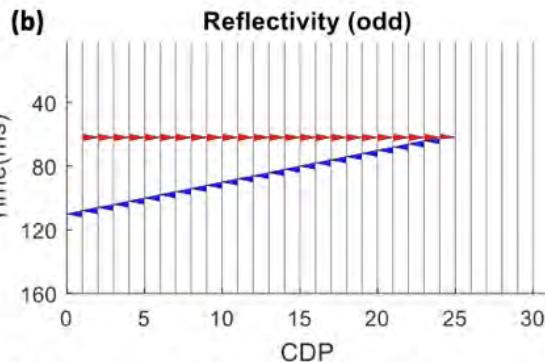
Synthetic training set

$$L = \frac{1}{2} \| (r_s(t) - h_{\Theta}^{\dagger}(t) * s_s(t)) \|_2^2$$

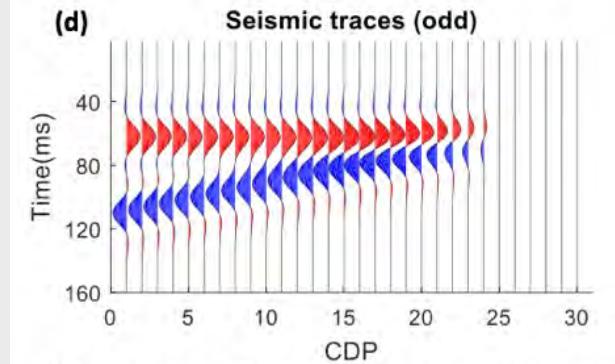


Resolution test

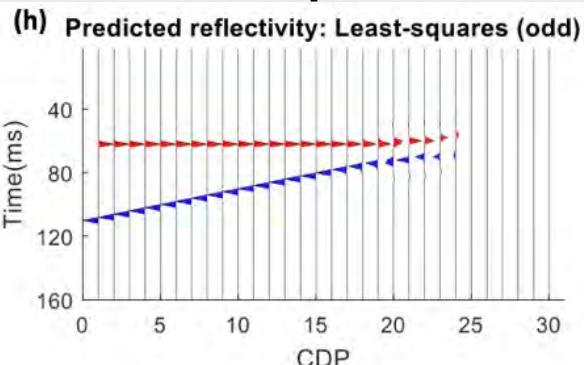
True model



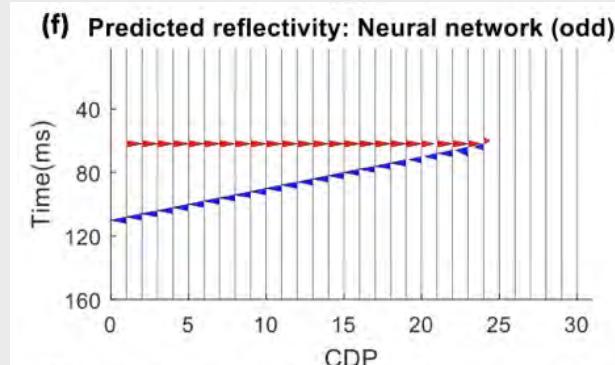
Input seismic data



Least squares

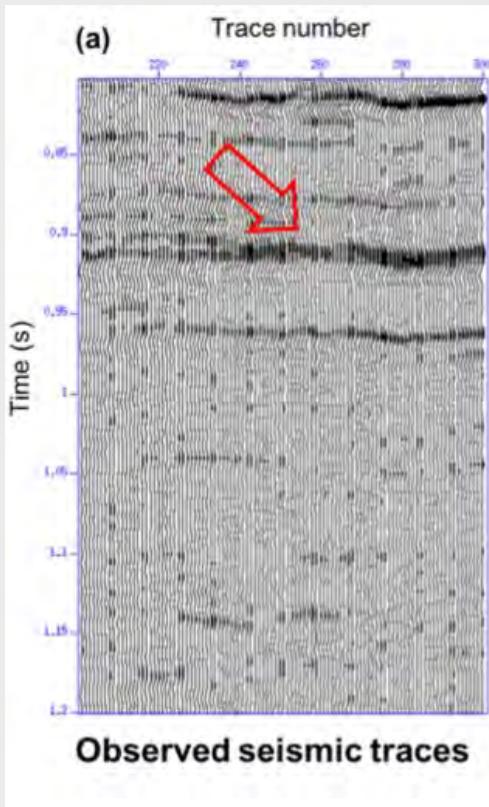


Neural network

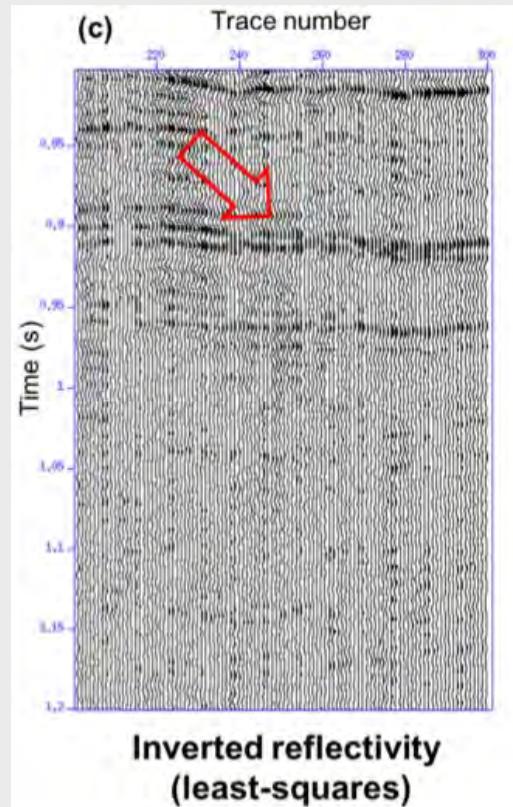


Field data example

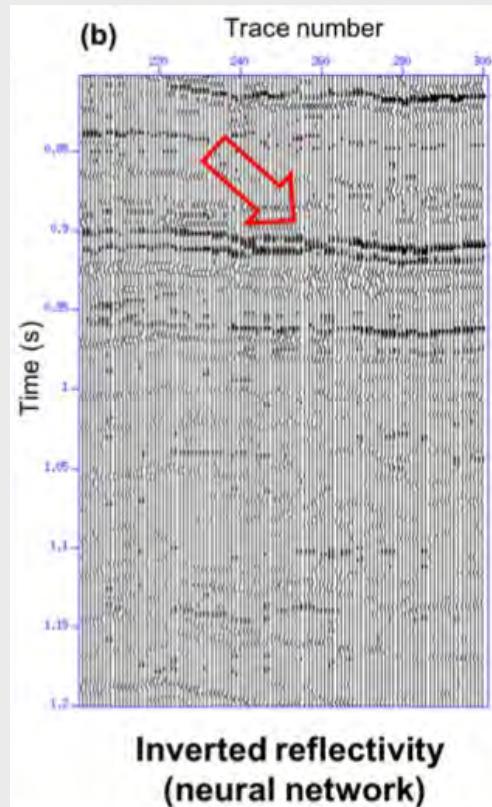
Input seismic data



Least squares

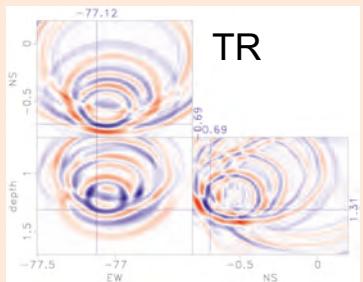


Neural network

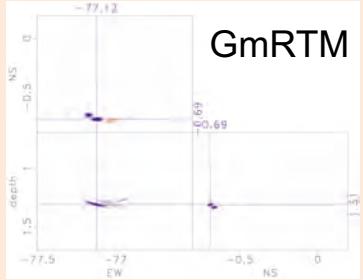


Studies for human-induced seismicity

Source parameters of micro earthquakes



GmRTM

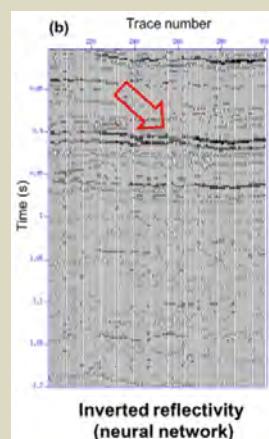
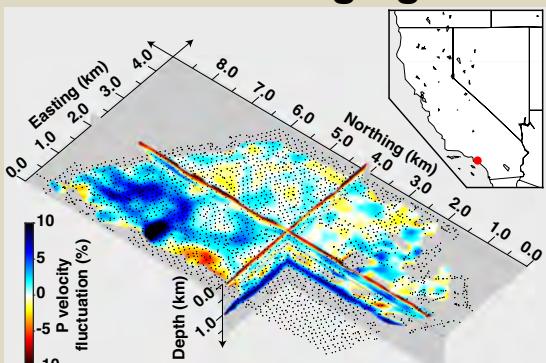


source

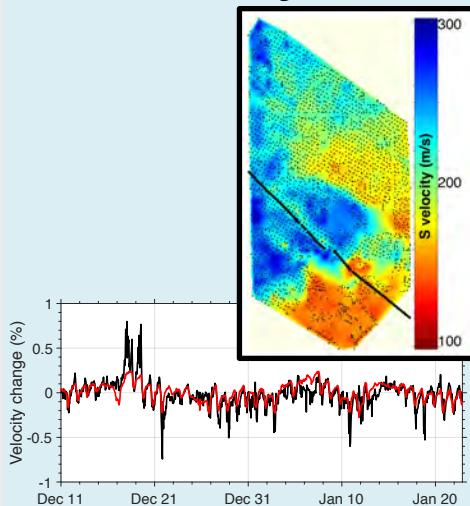
site

path

3D/4D High-resolution structural imaging

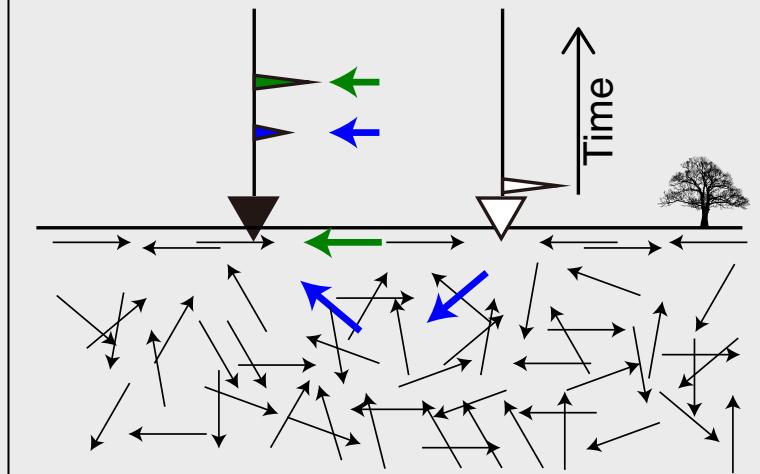
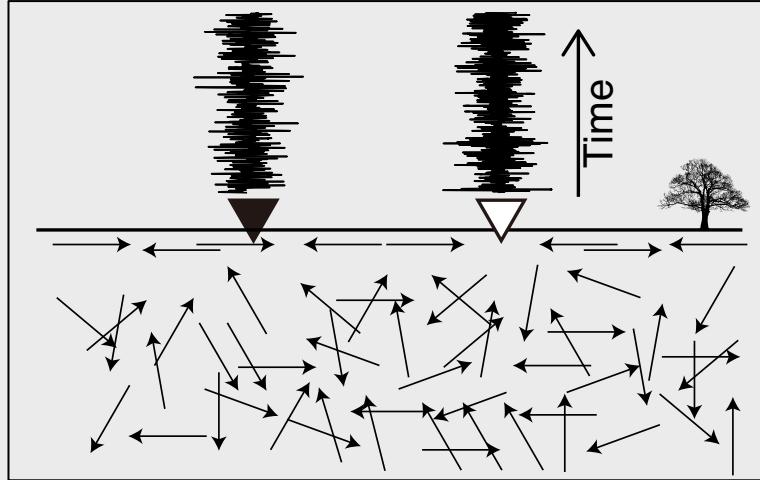


Near surface velocities and their dynamics



Publications:
<http://www.mit.edu/~nnakata>

Ambient noise correlation

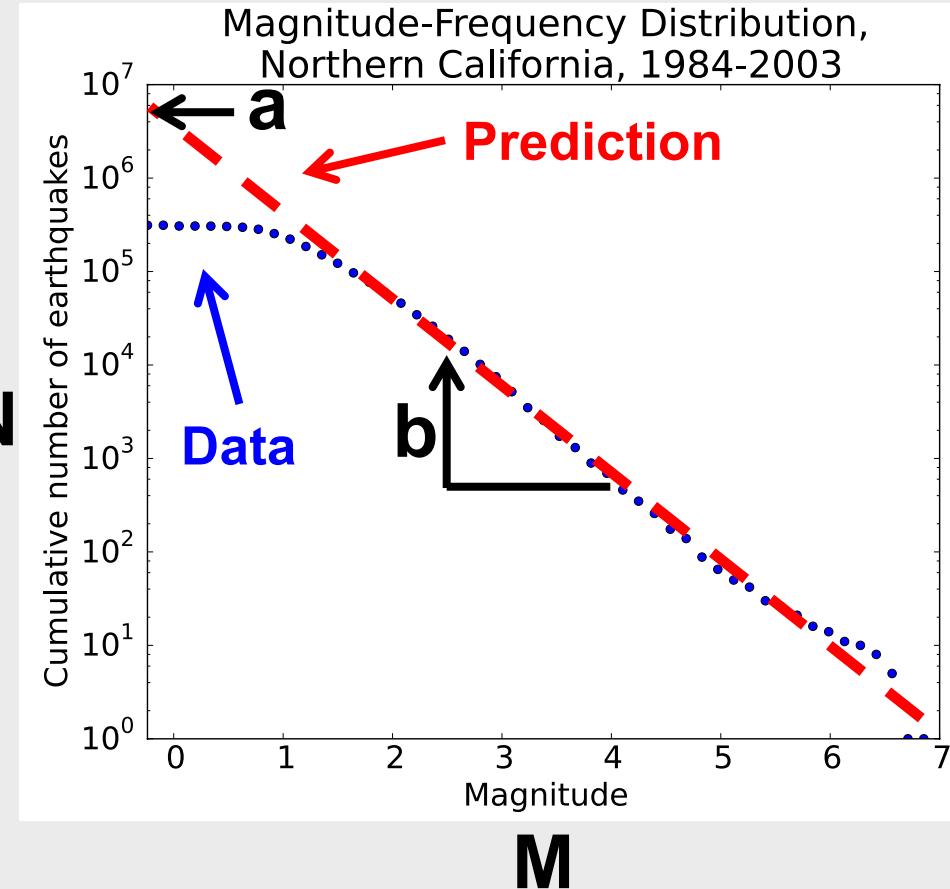


= ambient-field
correlation
∈ seismic interferometry

Processing

- crosscorrelation
- spectral whitening
- time and/or space averaging
- normalization
- etc...

Small EQs occur more frequently than large EQs



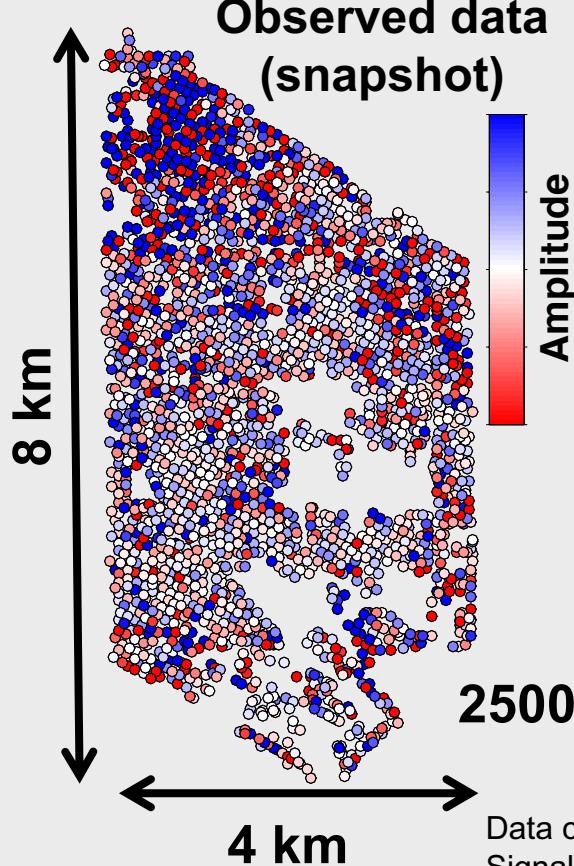
Gutenberg-Richter law (b value)

$$\log_{10} N = a - bM$$

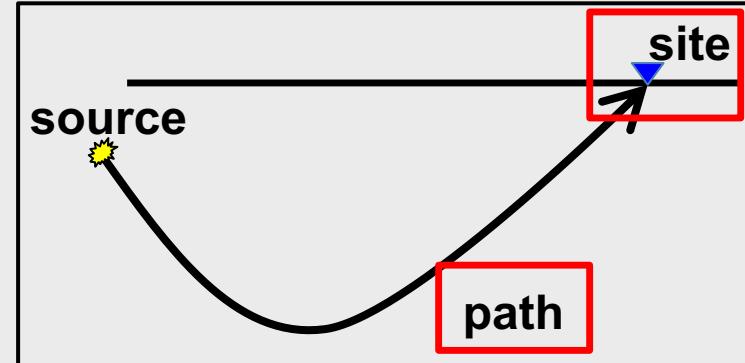
Ambient noise

Long Beach array (2012)

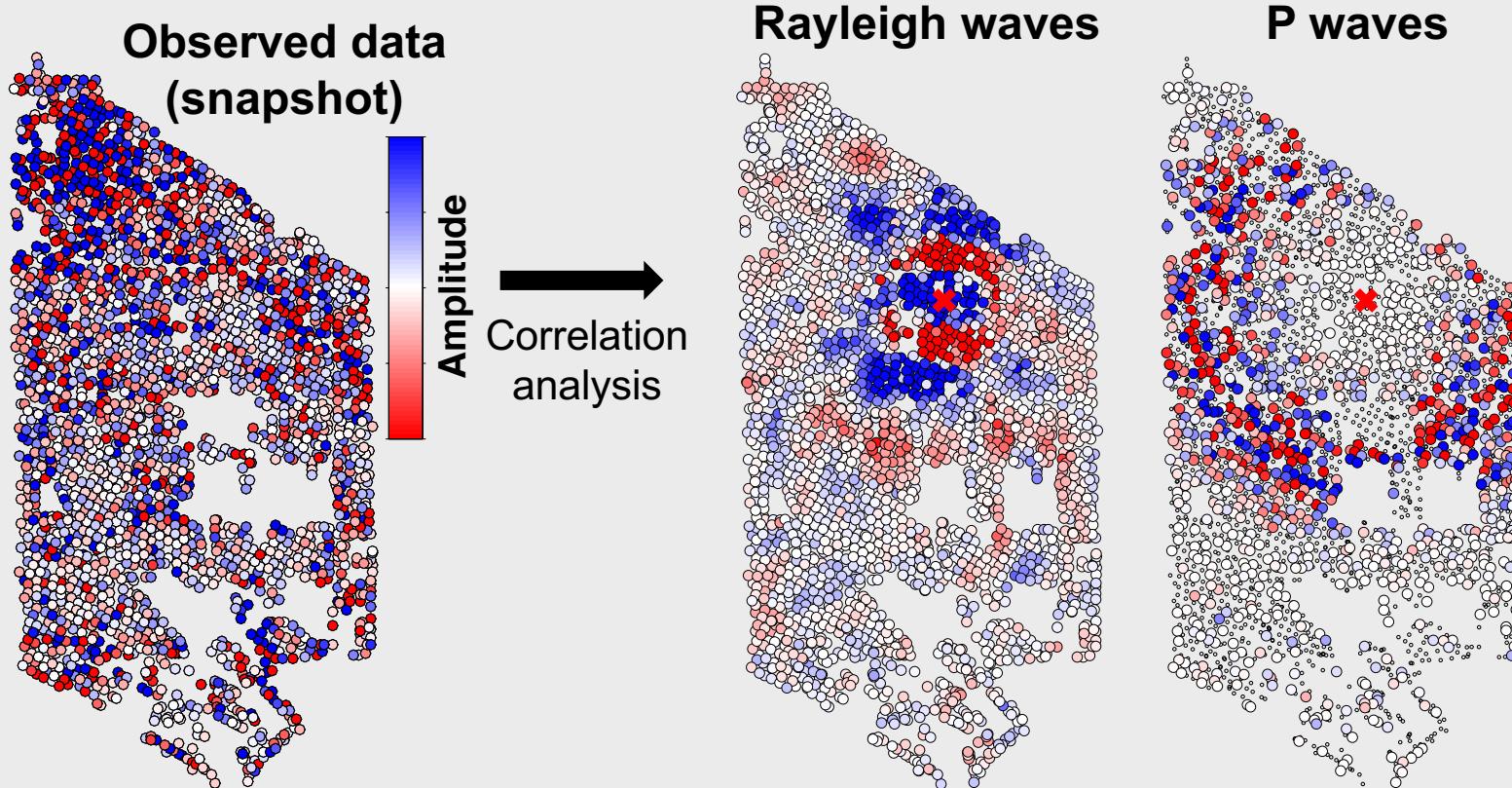
Observed data
(snapshot)



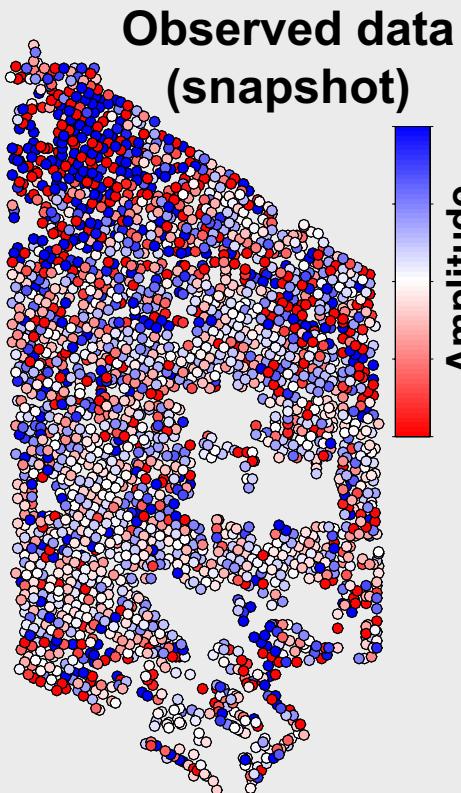
Data courtesy:
Signal Hill Petroleum, NodalSeismic



Data mining to extract coherent waves

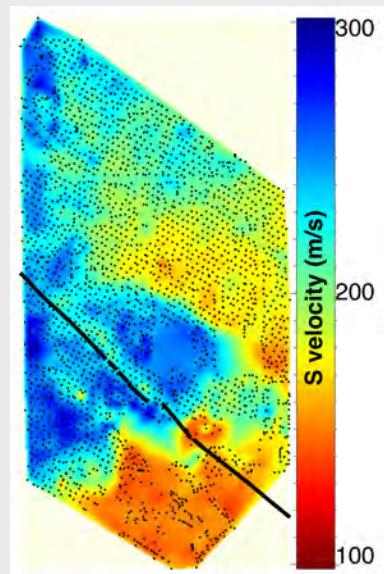


Velocity estimation

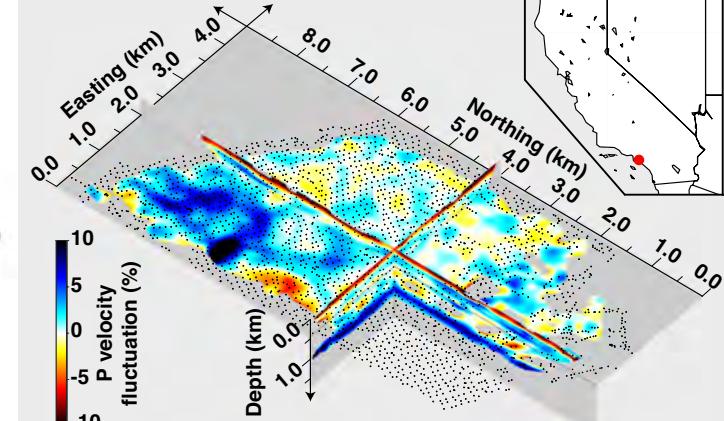


Correlation analysis

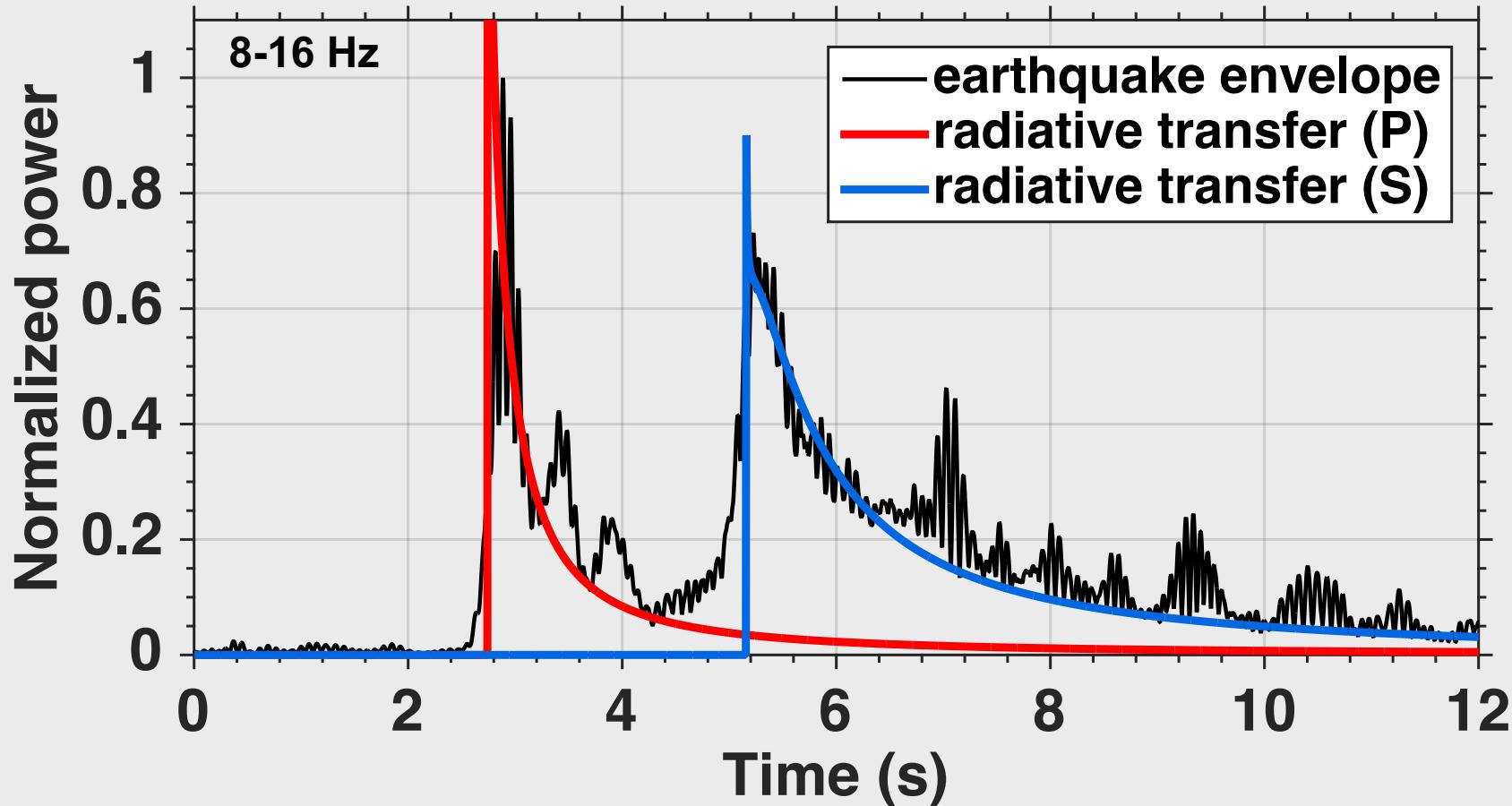
Near-surface
S velocity



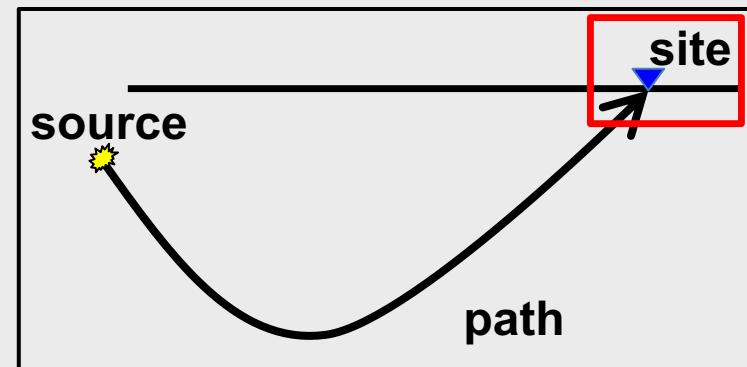
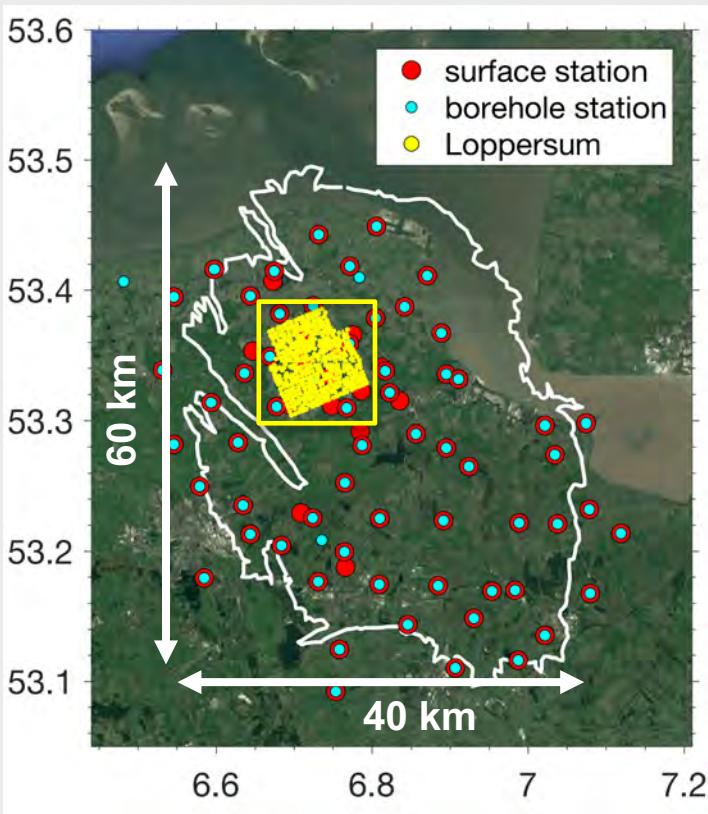
3D P velocity



High-frequency ground motion prediction

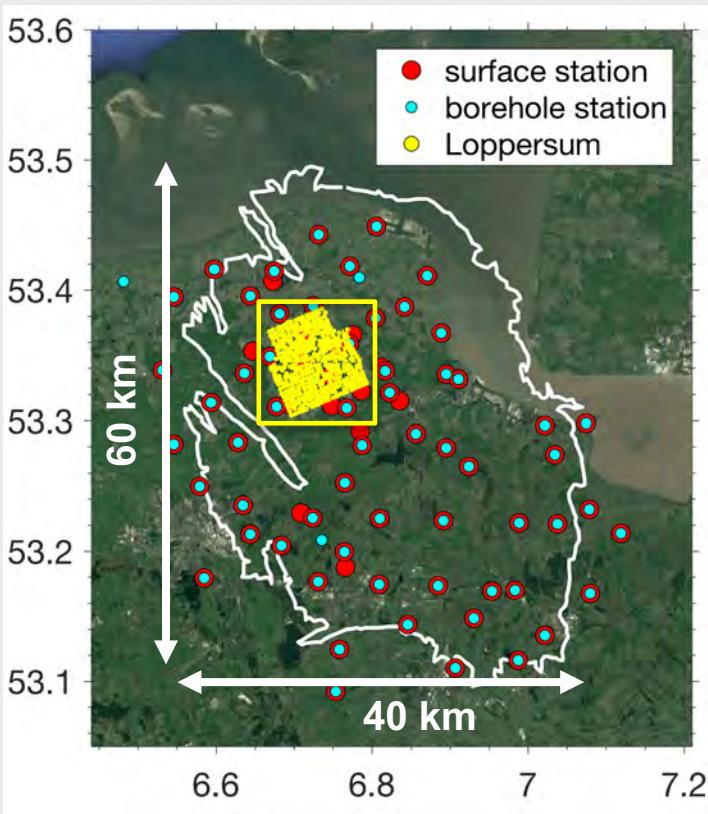


Groningen Gas Field



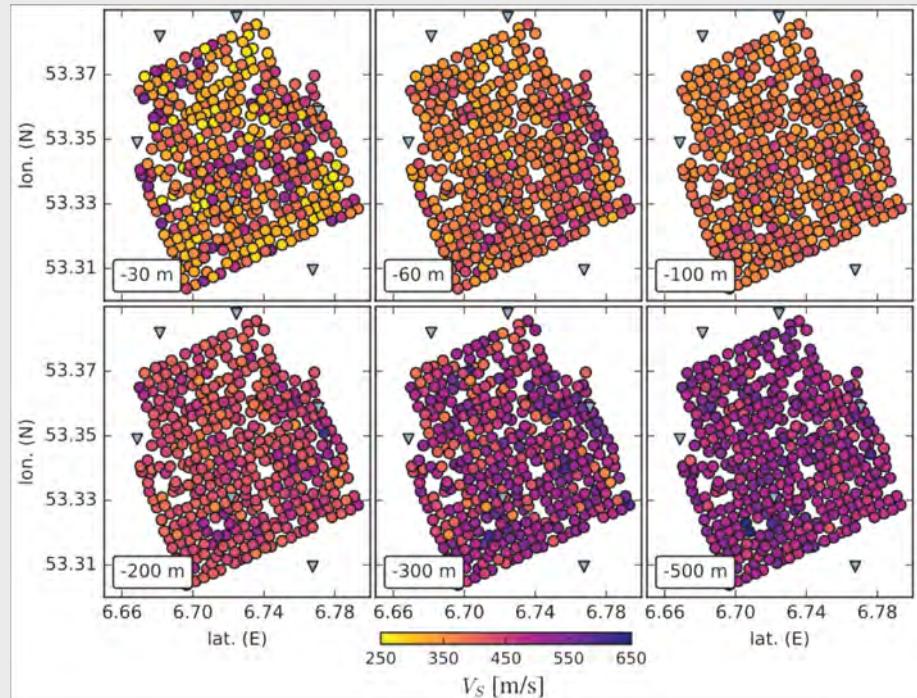
- **415 stations**
- **3 components**
- **Receiver spacing: 400 m**
- **One-month continuous record**

Groningen Gas Field

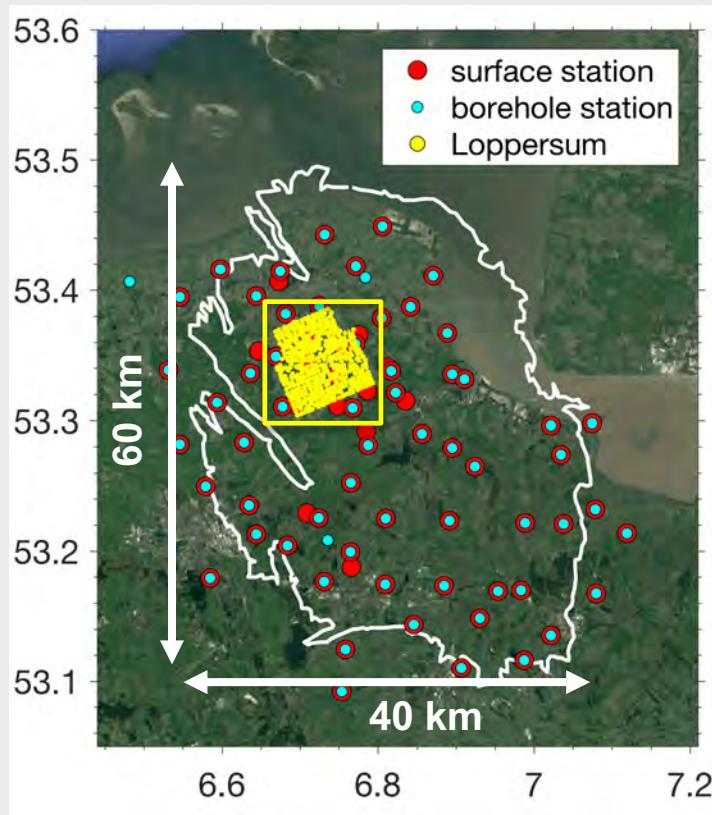


Near-surface velocity model estimated from

- H/V
- Rayleigh and Love waves



Groningen Gas Field



Hourly seismic velocity change

