

Inverse Problems and Imaging

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First lecture: friday, january 16, 2026, 9:00-12:00 (ENS, room 1Z25).

Material on the course website.

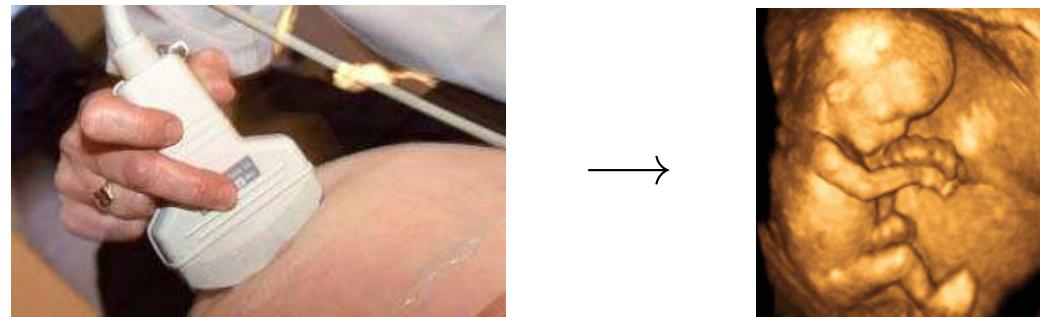
Validation: project (notebook jupyter + oral presentation).

Sensor array imaging

- Sensor array imaging (echography in medical imaging, sonar, non-destructive testing, seismic exploration, radar, etc) has two steps:
 - data acquisition: an unknown medium is probed with waves; waves are emitted by a source (or a source array) and recorded by a receiver array.
 - data processing: the recorded signals are processed to identify the quantities of interest (reflector locations, etc).

- Example:

Ultrasound echography



- Standard imaging techniques require:

- good receivers,
- suitable conditions for wave propagation (ideally, the “target” is embedded in a homogeneous medium),
- controlled and known sources.

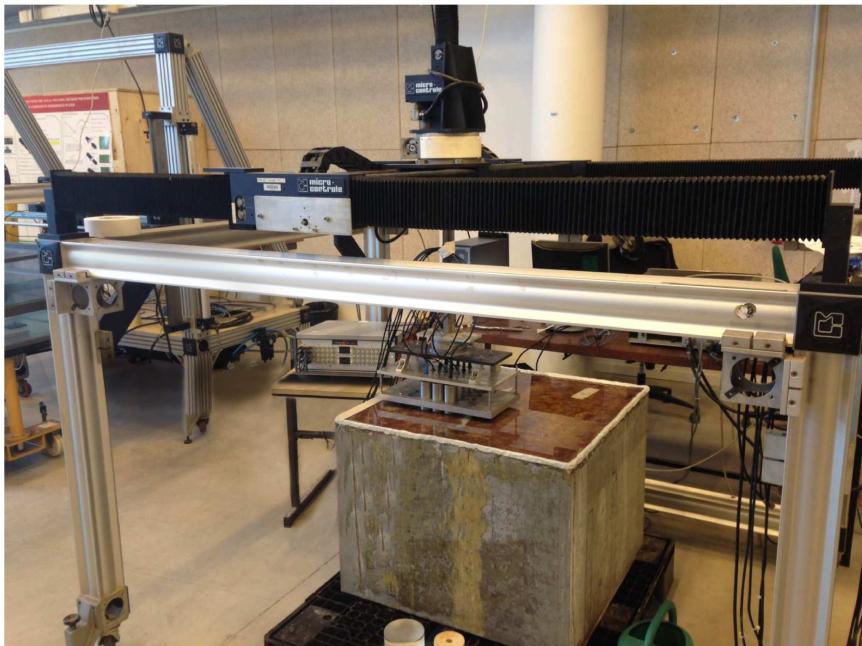
Sensor array imaging

- Goal: Propose and study imaging techniques that are robust with respect to:
 - measurement noise,
 - the complexity of the medium (heterogeneous medium),
 - the control and the knowledge of the sources.
- More generally: resolution of ill-posed inverse problems.

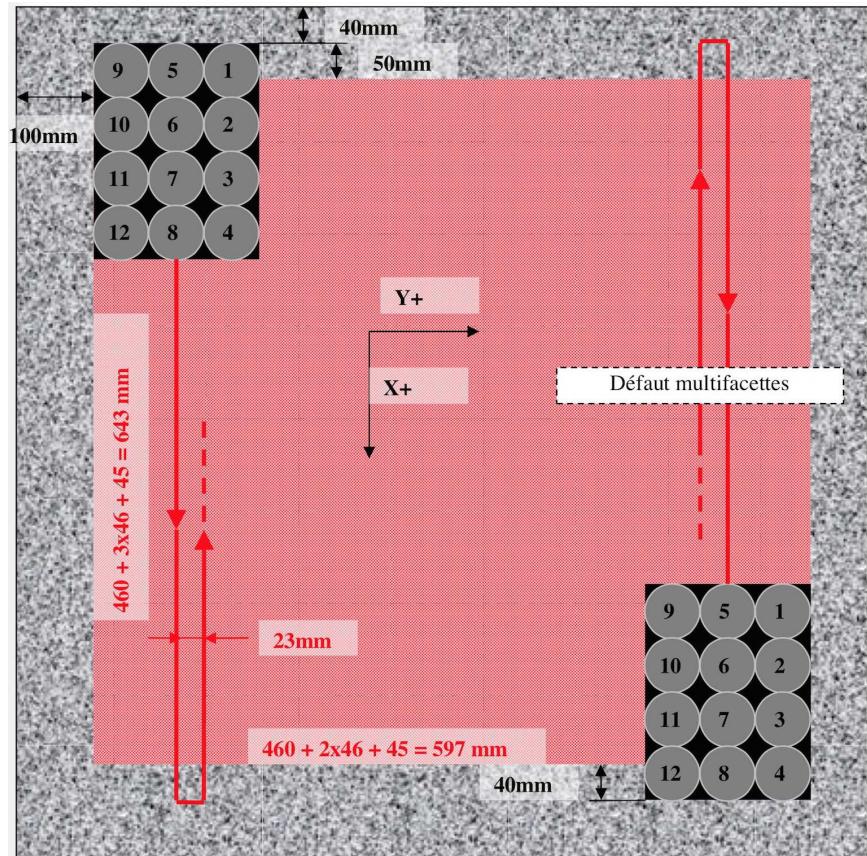
→ Introduce probabilistic and statistical techniques:

- Least squares optimization,
- Bayesian analysis,
- Random matrix theory,
- Spectral theory for stationary processes,
- Gaussian processes.

Application 1: Ultrasound echography in concrete

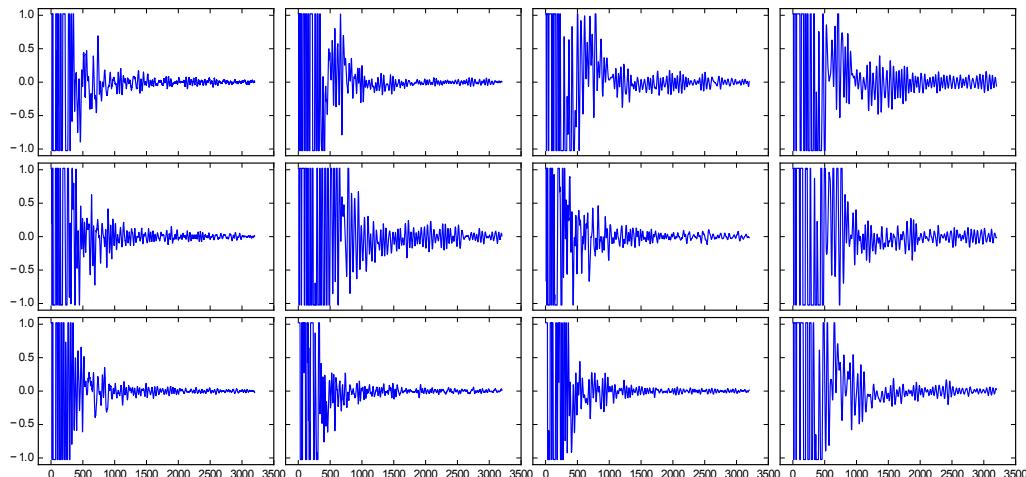


Experimental configuration

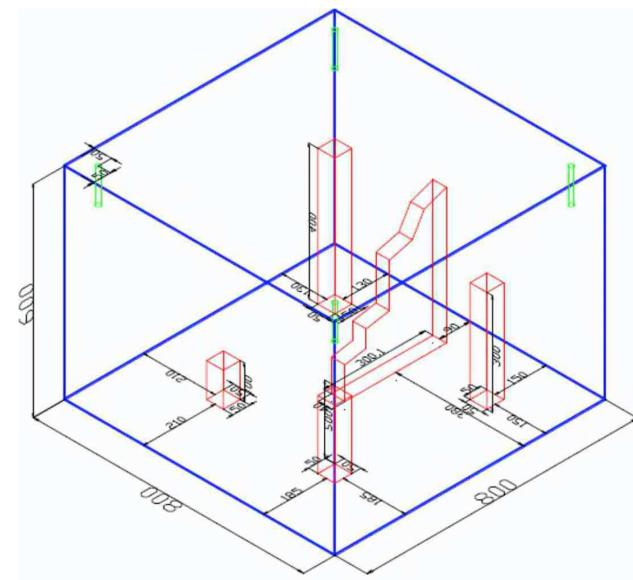


Top view of the acquisition geometry

Application 1: Ultrasound echography in concrete

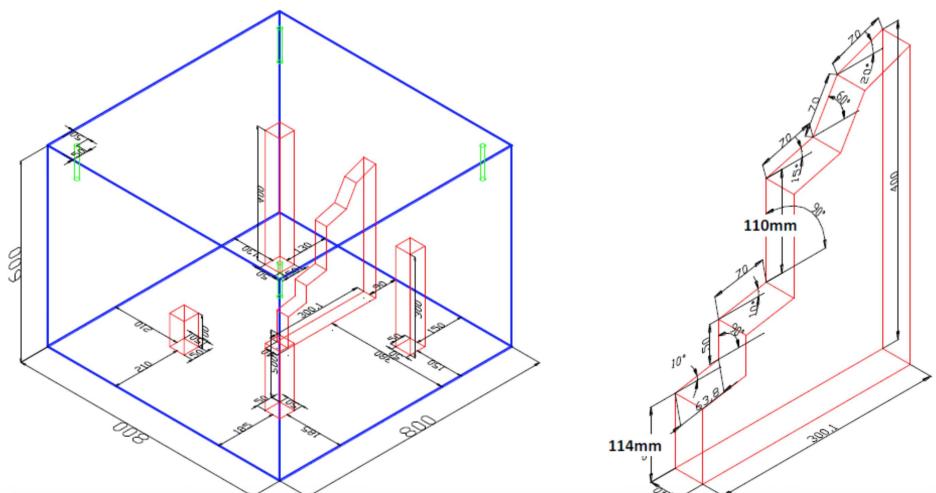


Data

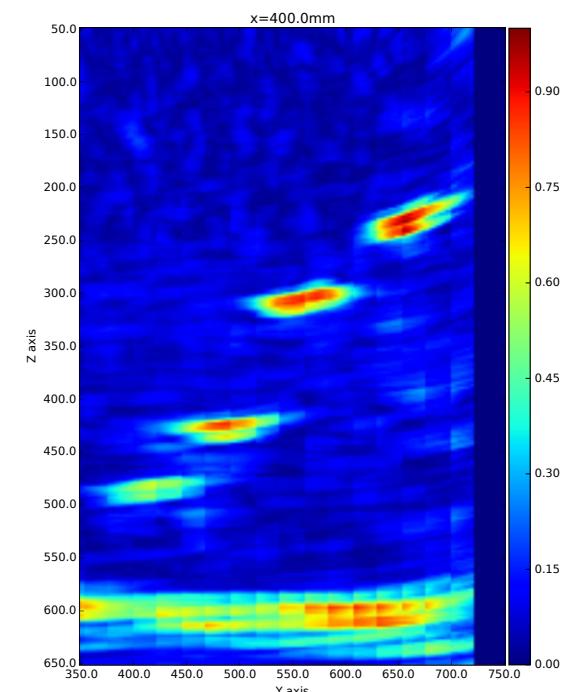


Real configuration

Application 1: Ultrasound echography in concrete

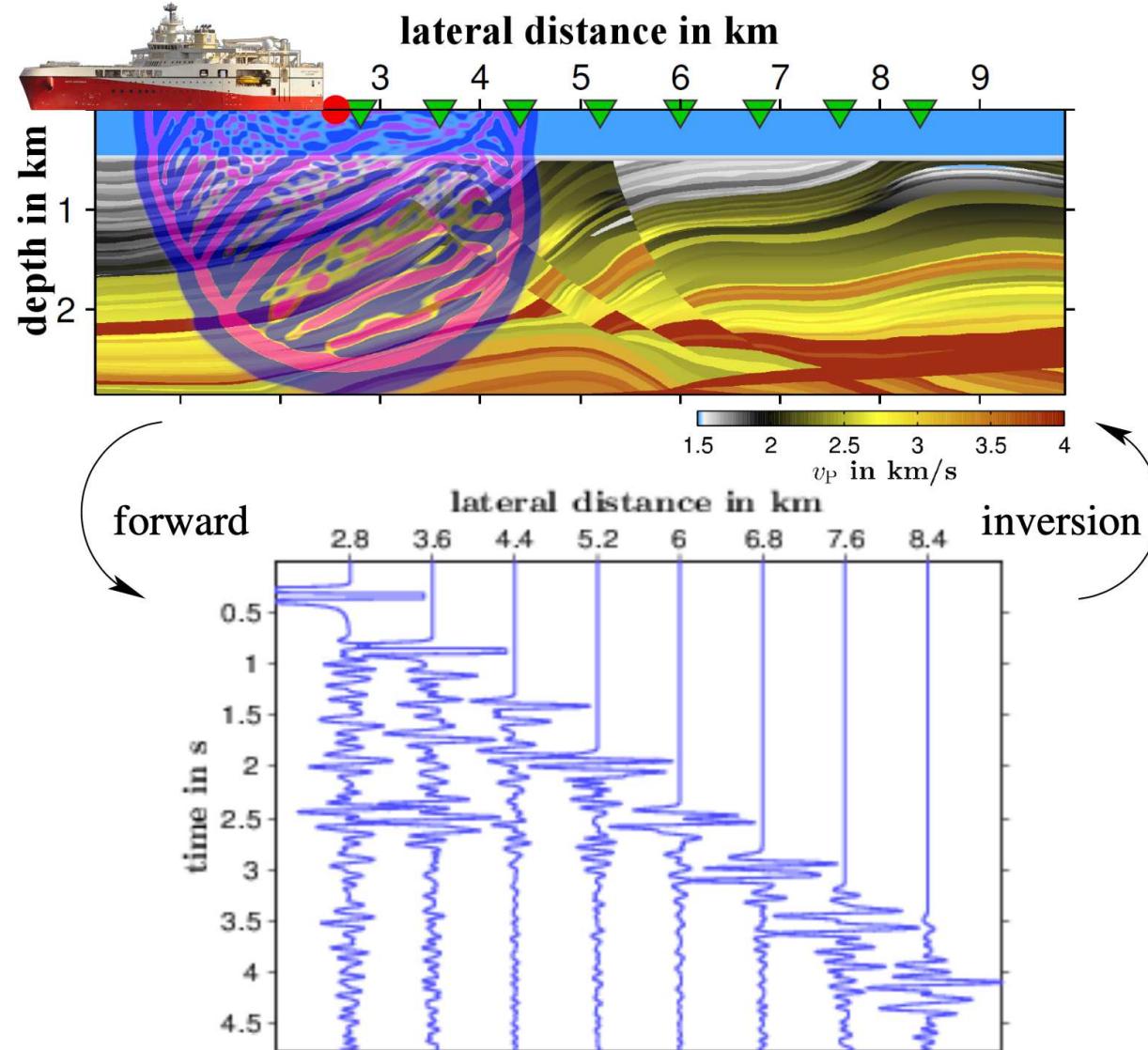


Real configuration



2D Image (along the complex defect plane)

Application 2: Reflection seismology



An inverse problem: Velocity estimation problem

- *Direct problem:* Given the velocity map $c = (c(x))_{x \in \Omega}$ of the medium, compute the wavefield solution of the wave equation

$$[\partial_t^2 - c^2(x)\Delta]p^{(s)}(t, x) = f(t)\delta(x - x_s), \quad t \in \mathbb{R}, x \in \Omega,$$

starting from $p^{(s)}(t, x) = 0$, $t \ll 0$.

At the locations of the receivers:

$$d_{r,s}(t) = p^{(s)}(t, x_r), \quad r, s = 1, \dots, N$$

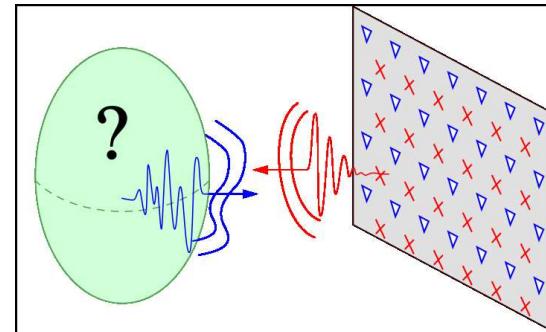
→ forward map

$$\mathcal{D} : c \mapsto \mathbf{d}$$

where $\mathbf{d} = ((d_{r,s}(t))_{r,s=1}^N)_{t \in [t_{\min}, t_{\max}]}$, is the array response matrix.

- *Inverse problem:*

Given the time-resolved measurements \mathbf{d} , determine the velocity map c .



Full Waveform Inversion (FWI)

- FWI fits the model prediction $\mathcal{D}[\mathbf{c}]$ with the measured data \mathbf{d}_{meas} (least-square minimization):

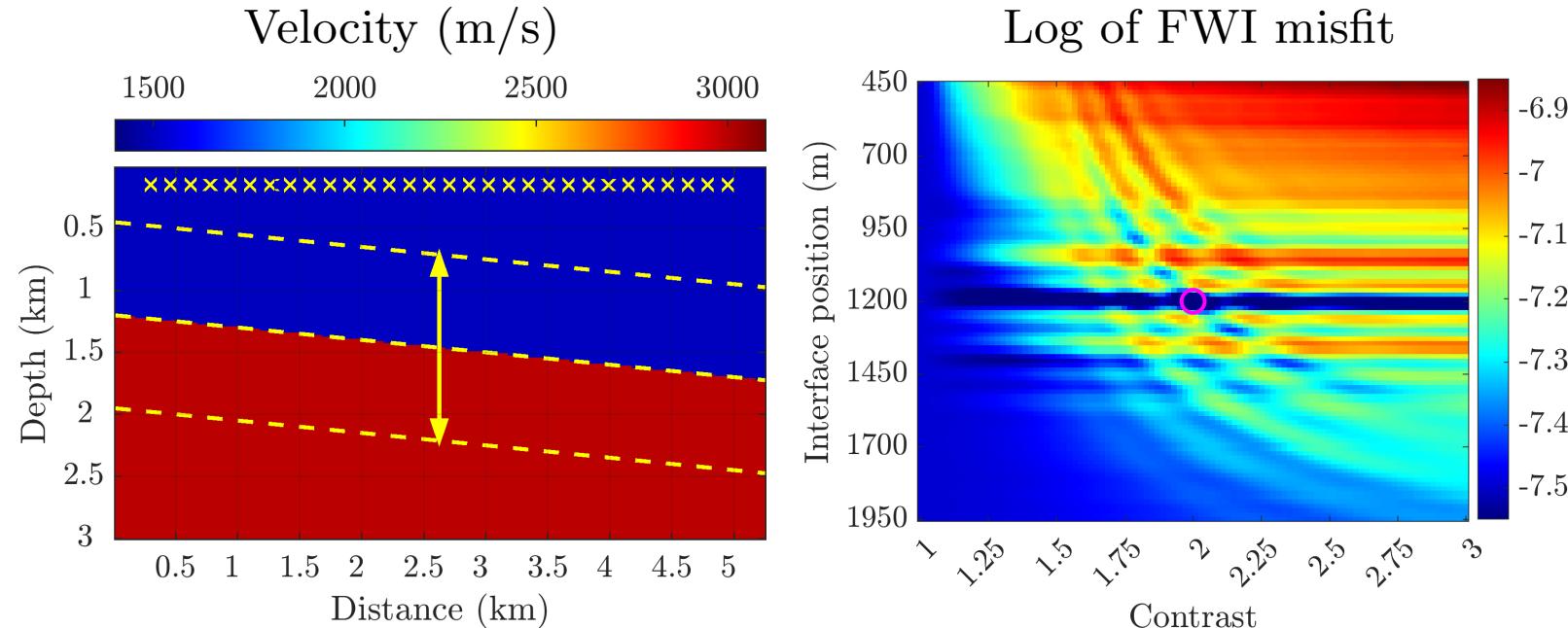
$$\hat{\mathbf{c}} = \underset{\mathbf{c}}{\operatorname{argmin}} \mathcal{O}_{FWI}[\mathbf{c}],$$

$$\mathcal{O}_{FWI}[\mathbf{c}] = \|\mathcal{D}[\mathbf{c}] - \mathbf{d}_{meas}\|^2 = \sum_{r,s=1}^N \int_{t_{\min}}^{t_{\max}} |\mathcal{D}[\mathbf{c}](t)_{r,s} - d_{meas}(t)_{r,s}|^2 dt,$$

with $\mathbf{c} = (c(x))_{x \in \Omega}$.

- Mathematical formulation: a PDE constrained minimization problem.
- Resolution by iterative methods (Newton, Gauss-Newton, steepest-descent, ...).
- Problem: The objective function $\mathcal{O}_{FWI}[\mathbf{c}]$ is not convex in \mathbf{c} .
→ optimization needs hard to get good initial guess.

Topography of the FWI objective function



- Probing pulse is a modulated Gaussian pulse with central frequency 6Hz and bandwidth 4Hz .
- $N = 30$ sensors and $N_t = 39$ time samples at interval $\tau = 0.0435\text{s}$.
- Search velocity has two parameters: the bottom velocity and depth of the interface (the angle and top velocity are known).
- Objective function:

$$\mathcal{O}_{FWI}[c] = \|\mathbf{d}_{meas} - \mathcal{D}[c]\|_2^2$$

A short history of Full Waveform Inversion (FWI)

- Regularization [Virieux and Operto 2009]:

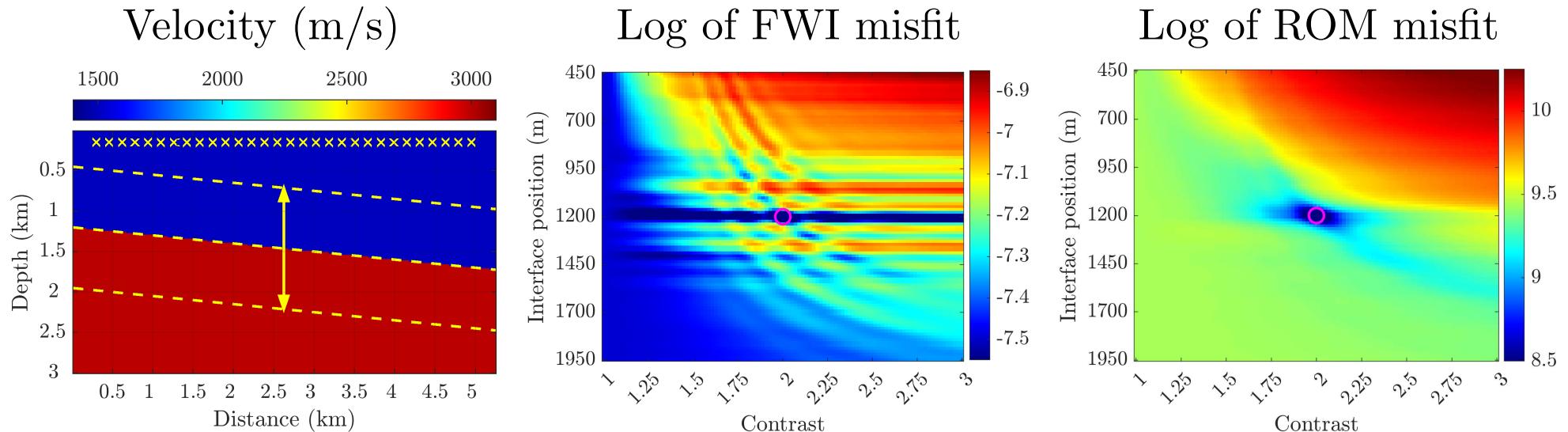
$$\hat{\mathbf{c}} = \operatorname{argmin}_{\mathbf{c}} \{ \|\mathcal{D}[\mathbf{c}] - \mathbf{d}_{meas}\|^2 + \lambda \operatorname{Reg}[\mathbf{c}] \},$$

with $\operatorname{Reg}[\mathbf{c}] = \|\mathbf{c}\|_{L^2}^2, \|\mathbf{c}\|_{L^1}, \|\mathbf{c}\|_{\text{TV}}, \dots$ (Bayesian interpretation).

- Progressive time continuation (layer stripping): Proceed hierarchically from the shallow part to the deep part [Sheng et al. 2006].
- Progressive frequency continuation: Successive inversion of subdata sets of increasing high-frequency content [Bunks et al. 1995].
- Optimal transport: Wasserstein distance instead of least-squares [Engquist et al. 2016].
- Extension (or relaxation) strategies: model-space extension [Symes 2008], source-space extension [Huang et al. 2018], receiver-space extension [Benziane et al. 2025].
- Use of reduced-order models [Borcea et al., 2024].

→ What about robustness/stability with respect to measurement noise, complexity of the medium, control of the sources ?

Topographies of the FWI and ROM objective functions



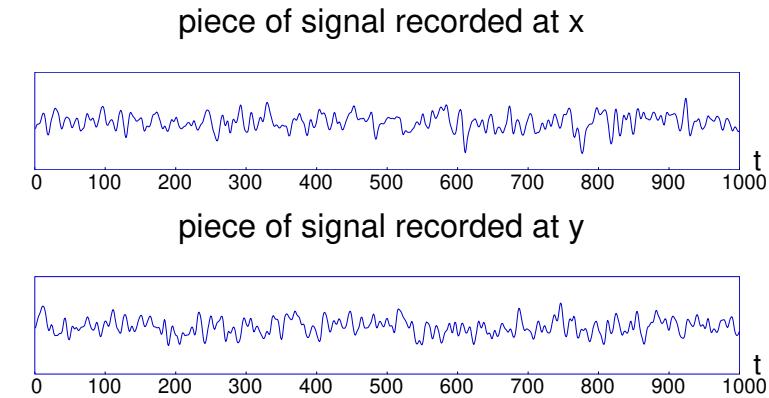
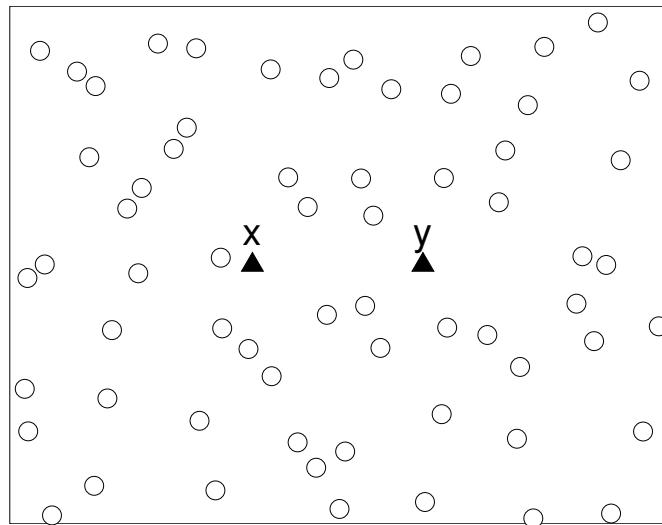
- Search velocity has two parameters: the contrast and the depth of the interface (the angle and top velocity are known).
- FWI objective function:

$$\mathcal{O}_{FWI}[c] = \|\mathcal{D}[c] - \mathbf{d}^{meas}\|_2^2$$

- ROM objective function:

$$\mathcal{O}_{ROM}[c] = \|\mathbf{A}^{rom}[c] - \mathbf{A}^{rom,meas}\|_2^2$$

Theory: Cross correlation of signals transmitted by noise sources

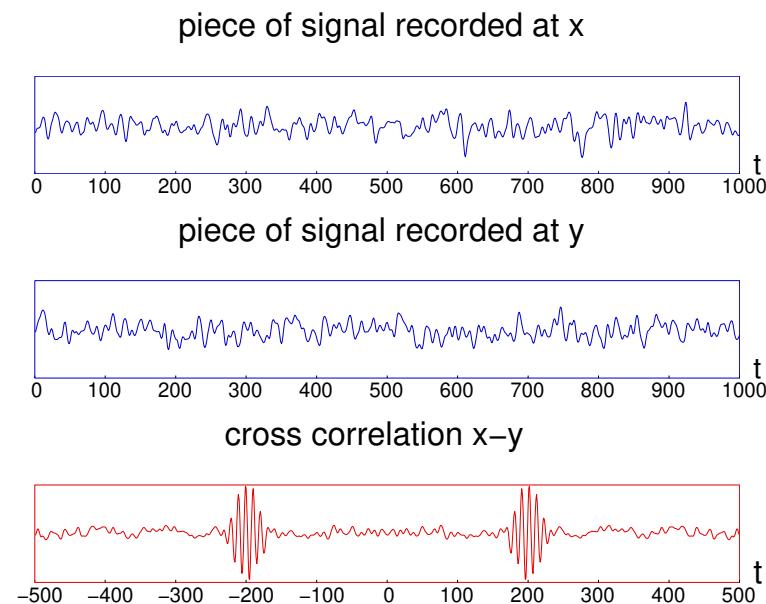
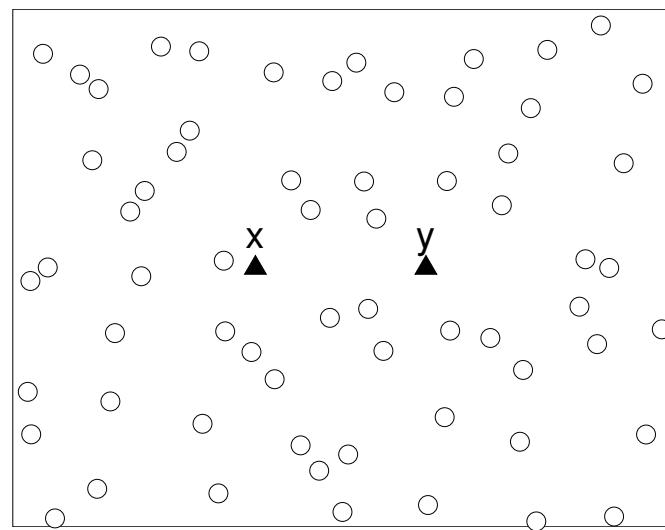


Numerical simulation of wave propagation
with many noise sources (○) and two receivers at x and y (▲)

How to extract information from the recorded signals $u_x(t)$ and $u_y(t)$?

These signals are just noise !

Theory: Cross correlation of signals transmitted by noise sources



Numerical simulation of wave propagation

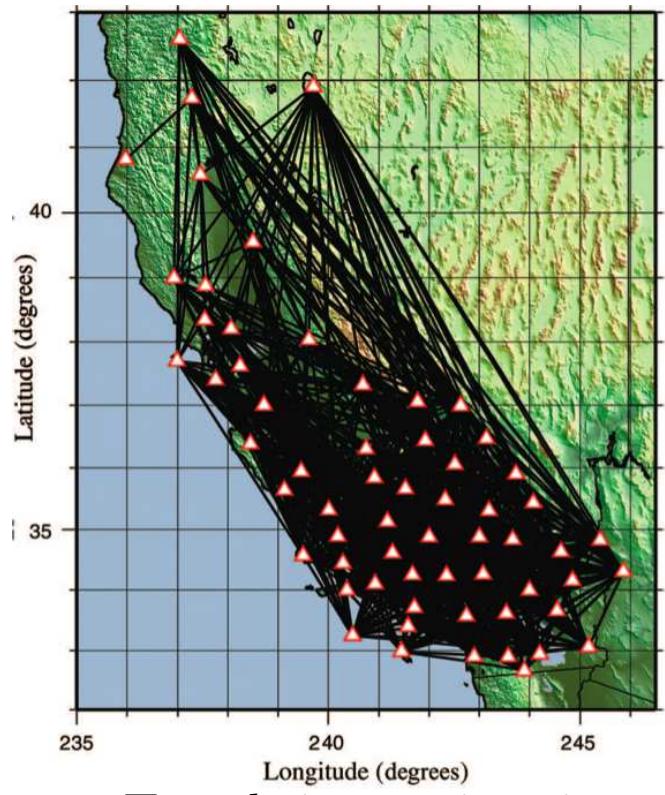
with many noise sources (○) and two receivers at x and y (▲)

→ Compute the cross correlation of the recorded signals

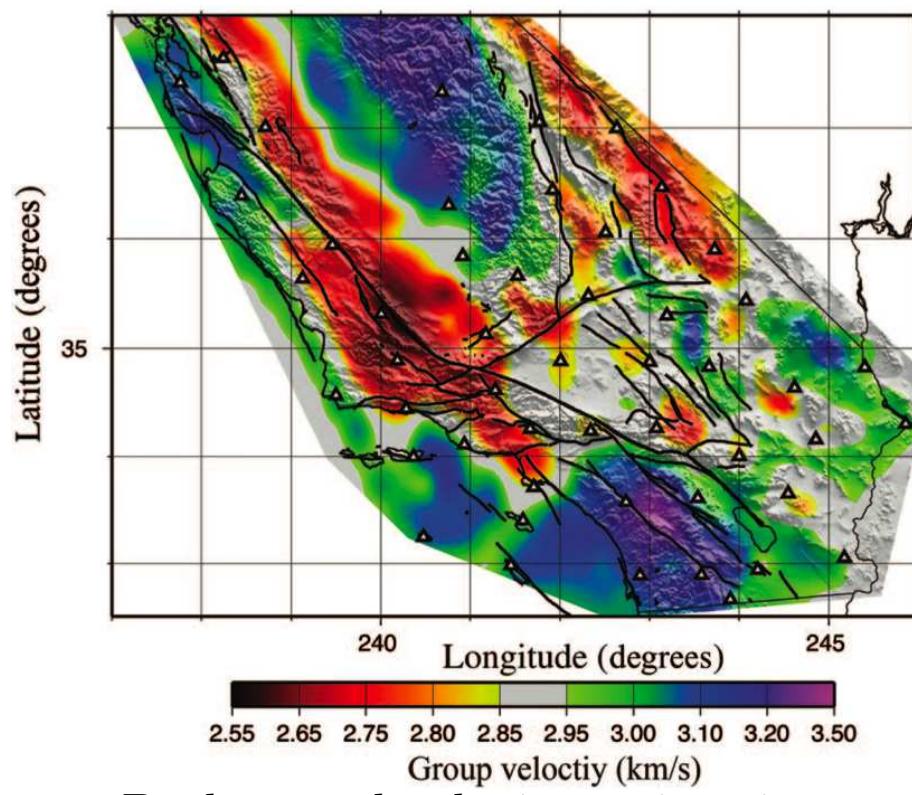
$$C_{\mathbf{x}, \mathbf{y}}^T(t) = \frac{1}{T} \int_0^T u_{\mathbf{x}}(s) u_{\mathbf{y}}(s + t) ds$$

and extract the travel time between the receivers at x and y .

Application 3: Seismic interferometry



Travel time estimation



Background velocity estimation

Application 3: Seismic interferometry

