

# Non periodic homogenization for elastic wave forward and inverse problems in seismology

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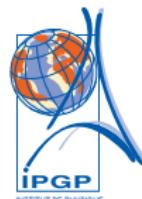
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# Science is such a difficult and dangerous job



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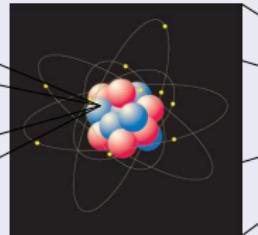
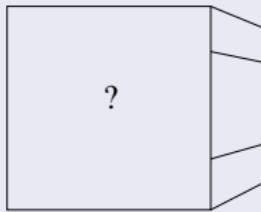
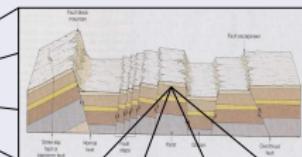
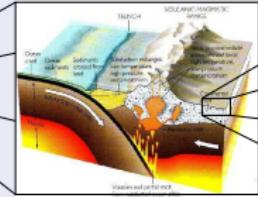
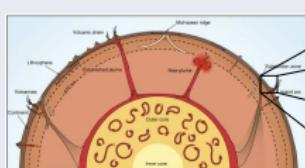
# Science is such a difficult and dangerous job



# Introduction and motivations

- Seismologists often work with frequency band limited data. → **the corresponding wavefield has a minimum wavelength**
- Observations indicate that waves of a **given wavelength** are sensitive to inhomogeneities of scales much smaller than this wavelength only in an **effective** way. This is one of the reasons why seismic waves can be used to image the earth.

The earth contains inhomogeneities at all scales.



# Small scales from the forward modeling point of view

Considering a 3-D medium, for which we can define

- a smallest inhomogeneities typical size  $\lambda_h$
- a minimum wave speed  $\alpha_{min}$

Solving the wave equation (with FD, SEM ...) for

- $N_s$  sources
- maximum frequency  $f_{max} \Rightarrow \lambda_{min} = \frac{\alpha_{min}}{f_{max}}$

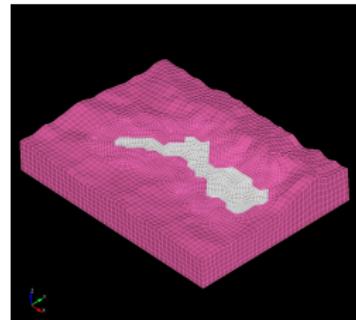
The computing time  $t_c$  is

- ① in a smooth 3-D medium (that is  $\lambda_h \gg \lambda_{min}$ )

$$t_c \propto N_s \lambda_{min}^{-4}$$

- ② in a rough 3-D medium (that is  $\lambda_h \ll \lambda_{min}$ )

$$t_c \propto N_s \lambda_h^{-4}$$



In a rough medium, the objective of homogenization is to bring back the numerical cost from  $N_s \lambda_h^{-4}$  to  $N_s \lambda_{min}^{-4}$

# Two scale homogenization : the non-periodic case

Homogenization is computing a smooth effective elastic tensor for a given minimum wavelength  $\lambda_{min}$  :

$$\mathbf{c}^{*,k_0} = \mathcal{H}^{k_0}(\mathbf{c})$$

where  $k_0 > k_{max}$  (typically  $k_0 = 2k_{max}$ , with  $k_{max} = 1/\lambda_{min}$ )

$\mathcal{H}^{k_0}$  is computed with a non-periodic homogenization technique. It implies

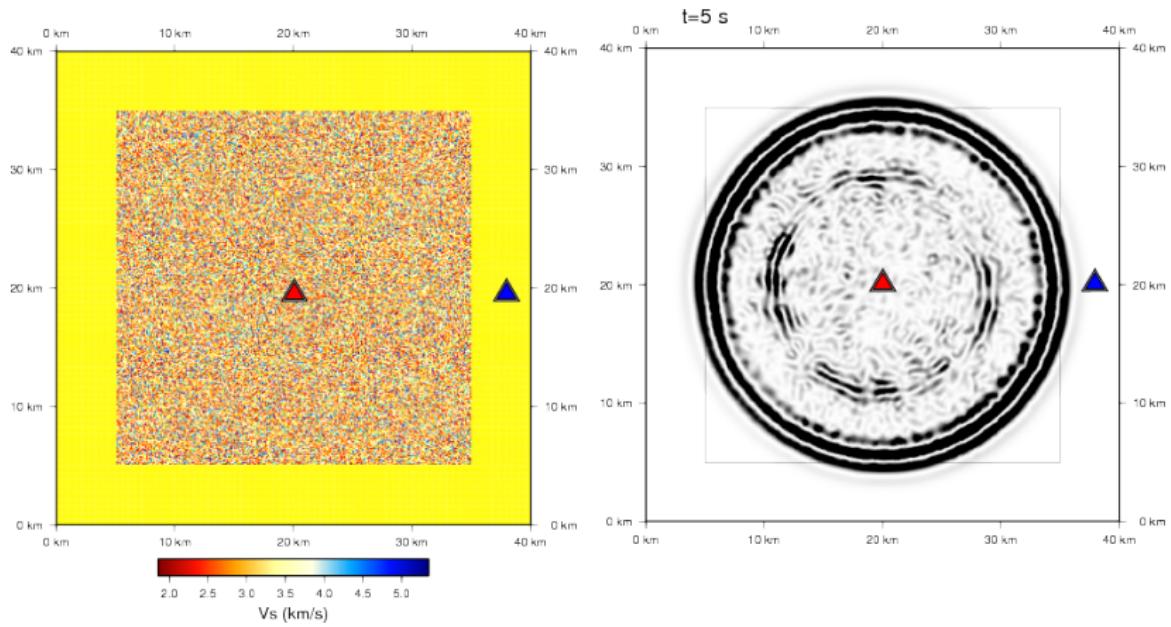
- solving a set of static loading problems (a PDE : the **cell problem**)
- use of smoothing operator (a lowpass spatial filter).

The whole process is non-linear (it is not just a linear filter of slowness or elastic parameters). If  $\mathcal{F}^{k_0}$  is simple lowpass filter operator ( $\mathcal{F}^{k_0}(g)(\mathbf{x}) = w^{k_0} * g(\mathbf{x})$ ), then

$$\mathcal{H}^{k_0}(\mathbf{c}) \neq \mathcal{F}^{k_0}(\mathbf{c})$$

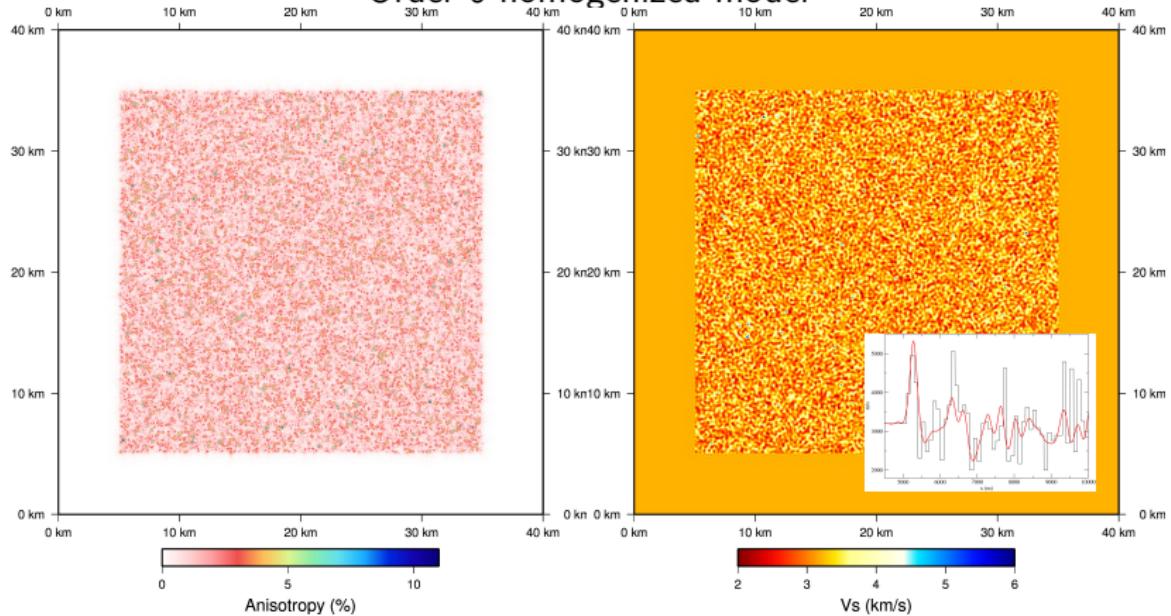
# A 2-D random example

The inner square is a  $300 \times 300$  homogeneous squares cells ( $100 \times 100$  m $^2$ ). In each cell, the density,  $\lambda$  and  $\mu$  are determine randomly with contrast up to 50%.

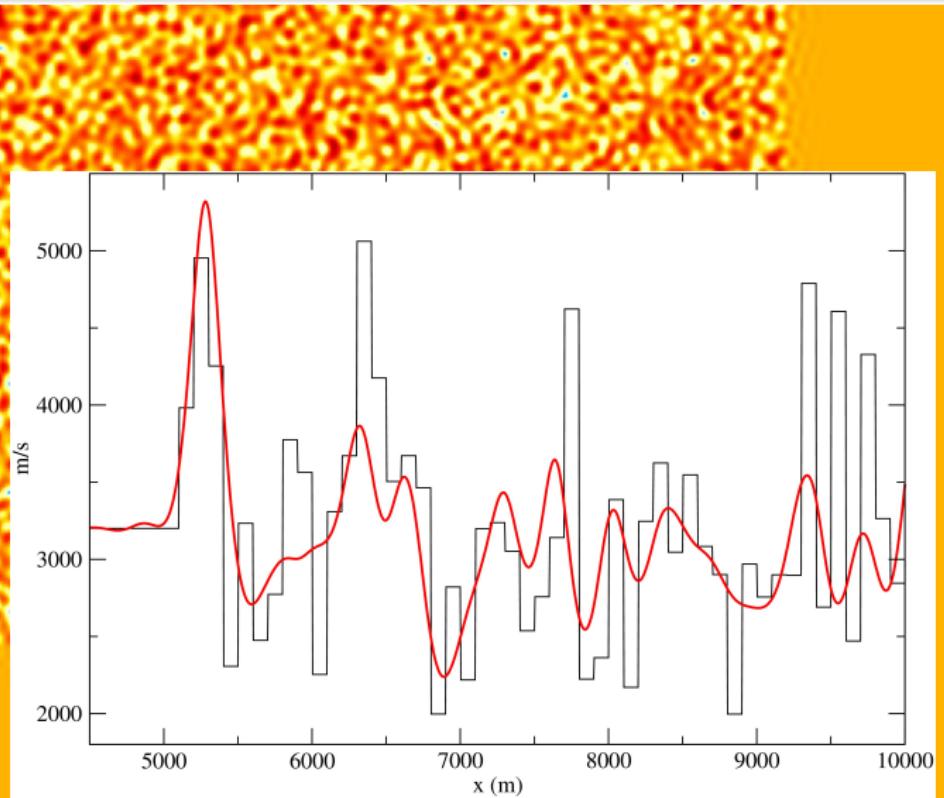


# A 2-D random example

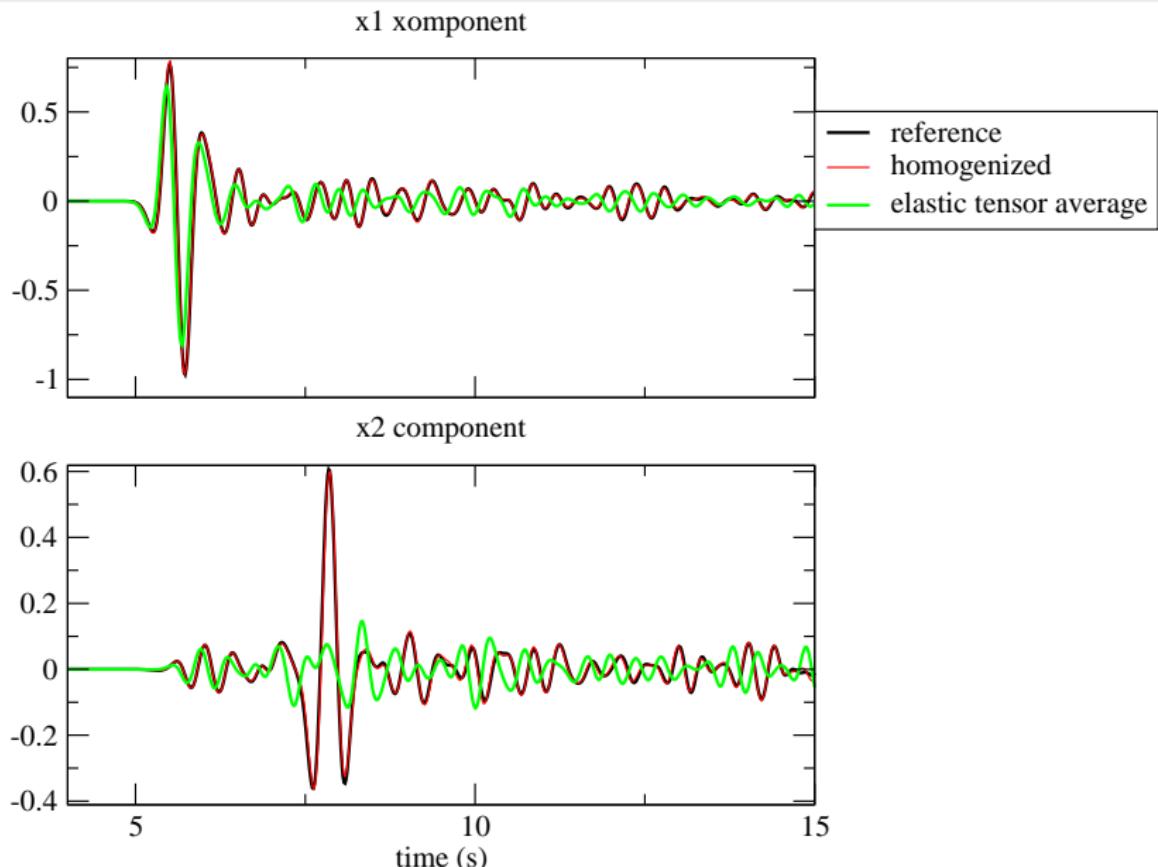
Order 0 homogenized model



# A 2-D random example



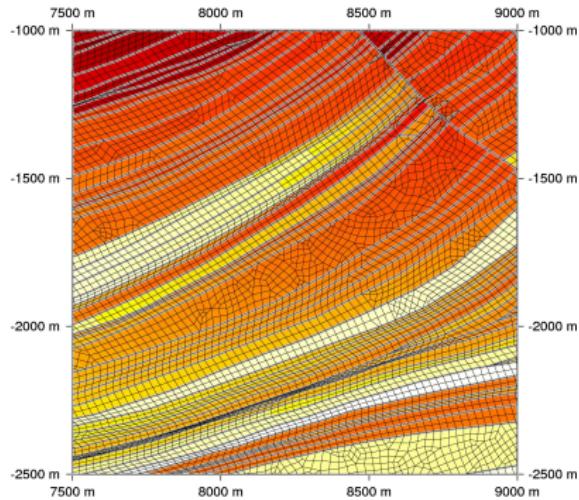
# A 2-D random example : source at the center of the square



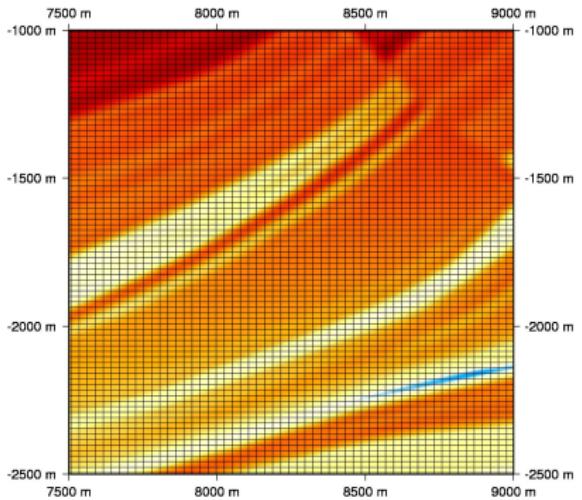
# Marmousi2 example

Homogenization allows for simple meshes, even for complex media.

Mesh for the original model



Mesh for the homogenized model



# Small scales from the inverse problem point of view

## Inverse problem, full waveform inversion (FWI)

- FWI can only be performed in a limited frequency band.
- The best that can be recovered is what is “seen” by the wavefield. It is an **effective** version  $(\rho^*, \mathbf{c}^*)$  of the real earth  $(\rho, \mathbf{c})$ .
- **We expect a FWI result to be at best an effective medium.**  
As a first step, we test this intuitive idea in the layered media simple case.

# Residual homogenization of layered media

- Non-periodic homogenization of layered media is simple because an analytically solution exists to the cell equation. In such case, the Backus parameters are involved (still with  $\varepsilon_0 = (k_0 \lambda_{min})^{-1}$ ) :

$$\rho^{\varepsilon_0*} = \mathcal{F}^{k_0}(\rho)$$

$$\frac{1}{C^{\varepsilon_0*}} = \mathcal{F}^{k_0}\left(\frac{1}{C}\right)$$

$$\frac{1}{L^{\varepsilon_0*}} = \mathcal{F}^{k_0}\left(\frac{1}{L}\right)$$

$$A^{\varepsilon_0*} - \frac{(F^{\varepsilon_0*})^2}{C^{\varepsilon_0*}} = \mathcal{F}^{k_0}\left(A - \frac{F^2}{C}\right)$$

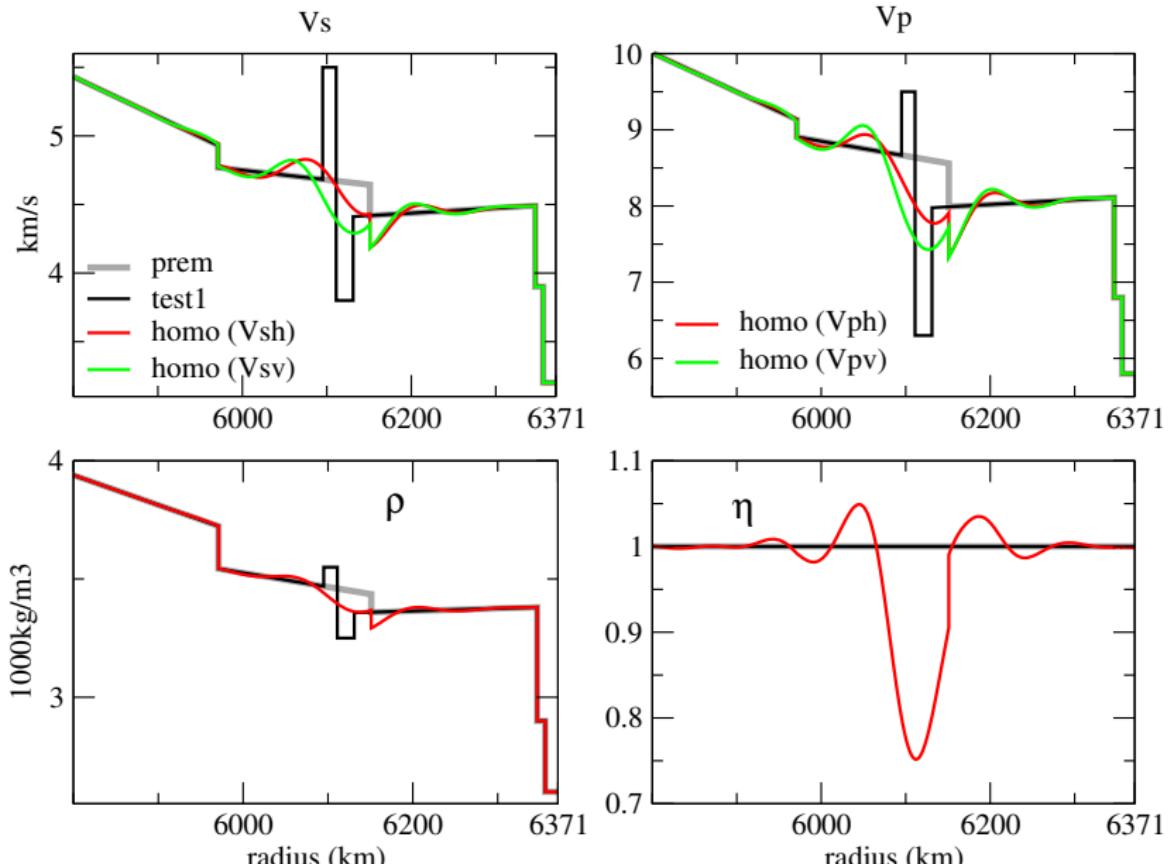
$$\frac{F^{\varepsilon_0*}}{C^{\varepsilon_0*}} = \mathcal{F}^{k_0}\left(\frac{F}{C}\right)$$

$$N^{\varepsilon_0*} = \mathcal{F}^{k_0}(N)$$

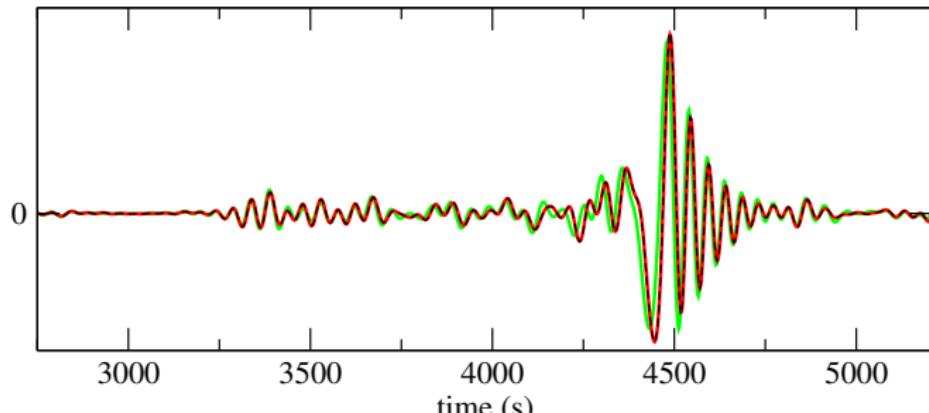
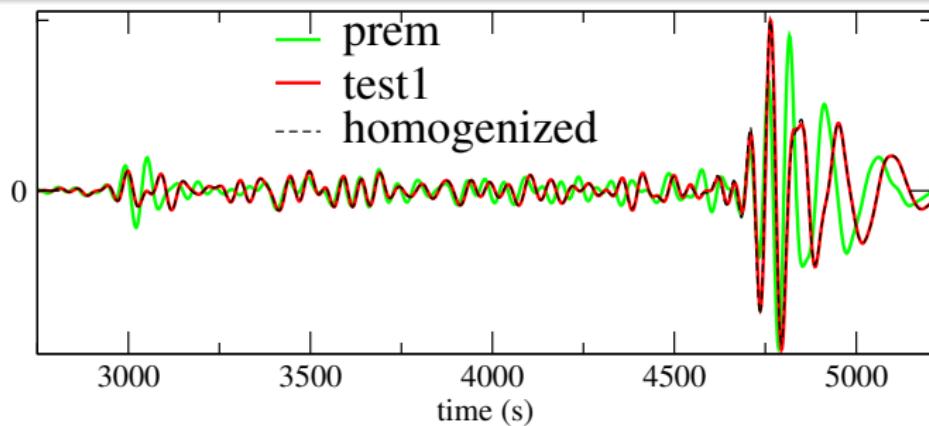
- We introduce the notion of residual homogenization, that allows to homogenize only the “difference” between a reference model and given model. For the  $C$  parameter, it gives :

$$\frac{1}{C^{\varepsilon_0*}} = \frac{1}{C_{\text{ref}}} + \mathcal{F}^{k_0}\left(\frac{1}{C} - \frac{1}{C_{\text{ref}}}\right)$$

# Residual homogenization in layered media



# Residual homogenization in layered media



# Full waveform inversion in layered media

To test the idea that a FWI can recover at best an homogenized model

- cost function to be minimized :

$$\Phi(\mathbf{m}) = {}^T[\mathbf{g}(\mathbf{m}) - \mathbf{d}] \mathbf{C}_d^{-1} [\mathbf{g}(\mathbf{m}) - \mathbf{d}]$$

with  $\mathbf{g}$  : forward problem ;  $\mathbf{m}$  model parameters ;  $\mathbf{d}$  data set.

- Gauss-Newton iterative least square minimization :

$$\mathbf{m}^{i+1} = \mathbf{m}^i + \left( {}^t \mathbf{G}^i \mathbf{C}_d^{i-1} \mathbf{G}^i + \lambda^i \right)^{-1} \left[ {}^t \mathbf{G}^i \mathbf{C}_d^{i-1} (\mathbf{d}^i - \mathbf{g}^i(\mathbf{m}^i)) \right],$$

where

- $\mathbf{m}^i$  model at iteration  $i$
- $\mathbf{G}^i$  partial derivative matrix
- $\lambda^i$  damping at iteration  $i$

We use normal modes to compute  $\mathbf{g}$  in  $\mathbf{m}^i$  and normal mode full coupling to compute  $\mathbf{G}^i$  at each iteration.

# Full waveform inversion in layered media

## **m** parameter set :

- we use the Backus parameters set  $p_i(r)$  ( $p_1 = \rho$  ;  $p_2 = \frac{F}{C}$  ;  $p_3 = \frac{1}{L}$  ;  $p_4 = A - \frac{F^2}{C}$  ;  $p_5 = \frac{F}{C}$  ;  $p_6 = N$ )
- we invert relative we respect to a reference model (here PREM),  $\mathbf{p}_{ref}$   
 $\delta\mathbf{p} = \mathbf{p} - \mathbf{p}_{ref}$
- we use a radial Lagrange polynomials for the vertical parameters, which can be defined by values at given set of radius  $r_1 \dots r_N$ .

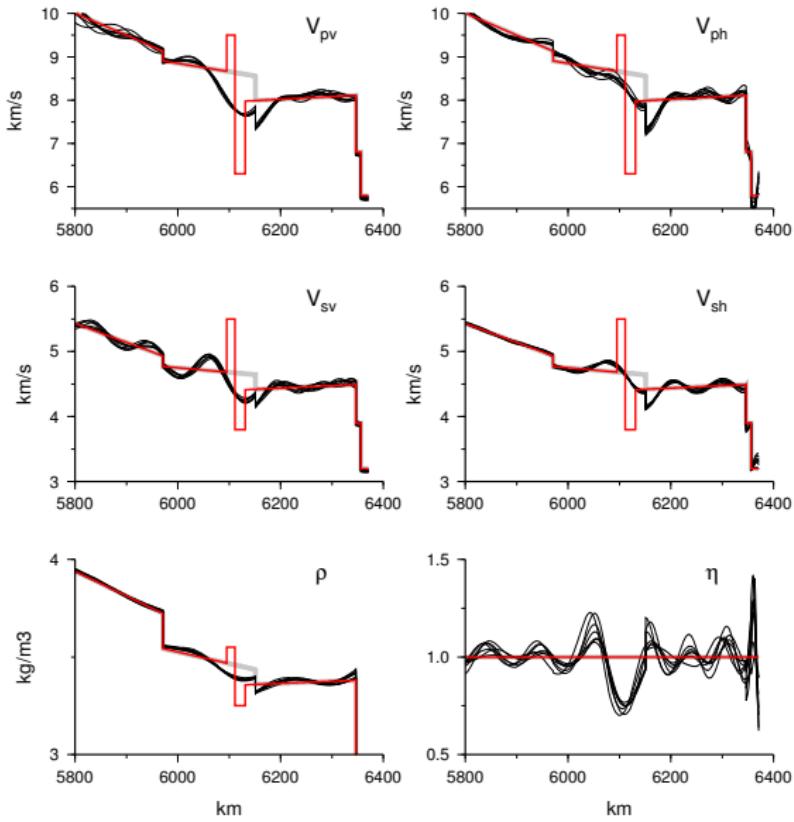
$$\mathbf{m} = (\delta p_1(r_1), \dots, \delta p_6(r_1), \delta p_1(r_2), \dots, \delta p_6(r_n))$$

## **d** data to be inverted

- Synthetic data **d** data to be inverted are computed with normal modes in TEST1 model ;
- a random noise of 3% amplitude is added to the data ;
- we use a single 3 components receiver in the period band [33s-1000s], two sources depth (10 km and 700 km) and an epicentral distance of 130°.

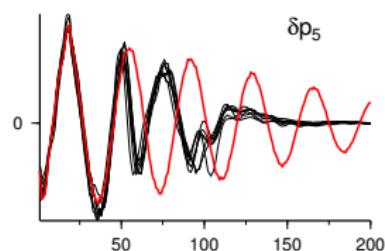
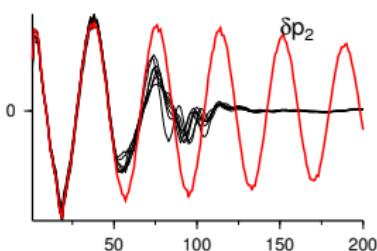
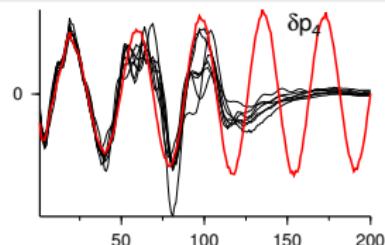
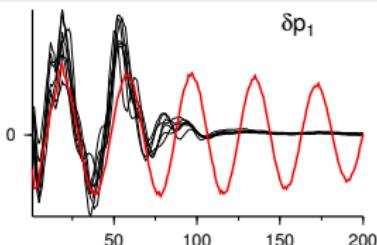
# Full waveform inversion in layered media : raw result

- grey line : reference model
- red line : target model
- black lines : inversion results for 8 different realizations of the 3% random noise



# Full waveform inversion : raw result in the spectral domain

- red line : target model
- black lines : inversion results for 8 different realizations of the 3% random noise



$$p_1 = \rho$$

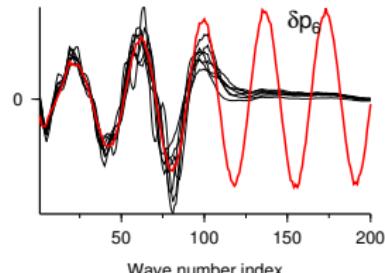
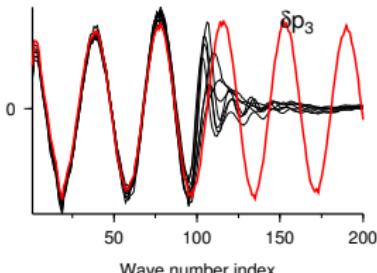
$$p_2 = \frac{1}{C}$$

$$p_3 = \frac{f}{L}$$

$$p_4 = A - \frac{F^2}{C}$$

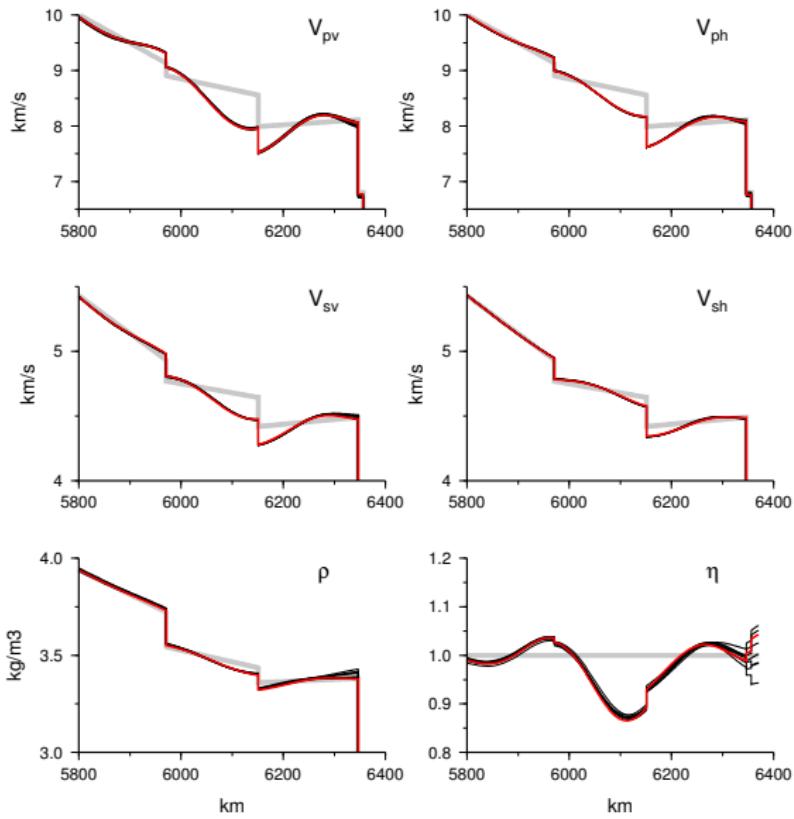
$$p_5(r) = \frac{F}{C}$$

$$p_6 = N$$



# Full waveform inversion : homogenized result

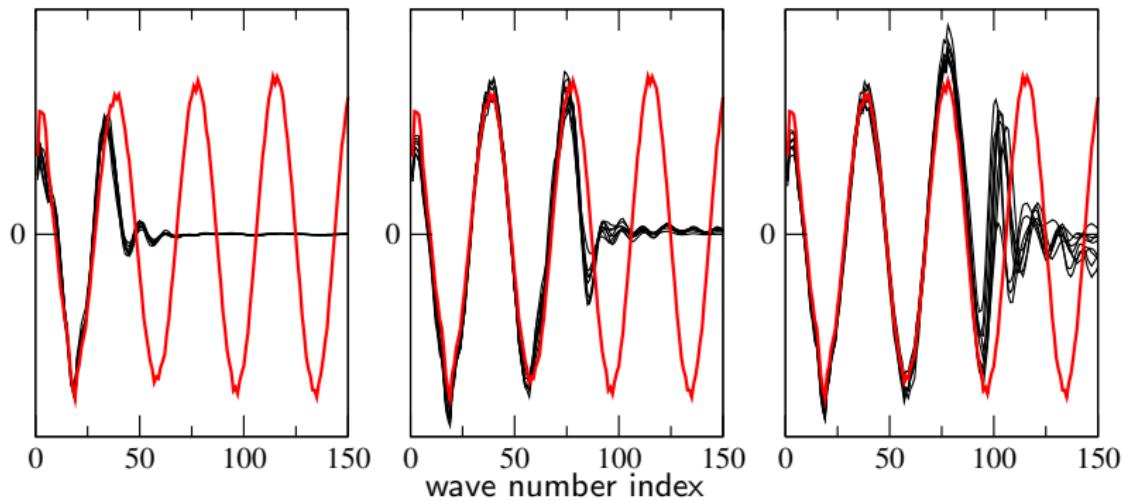
- grey line : reference model
- red line : homogenized target model
- black lines : homogenized inversion results for 8 different realizations of the 3% random noise



# Full waveform inversion : the inversion is multi-scale

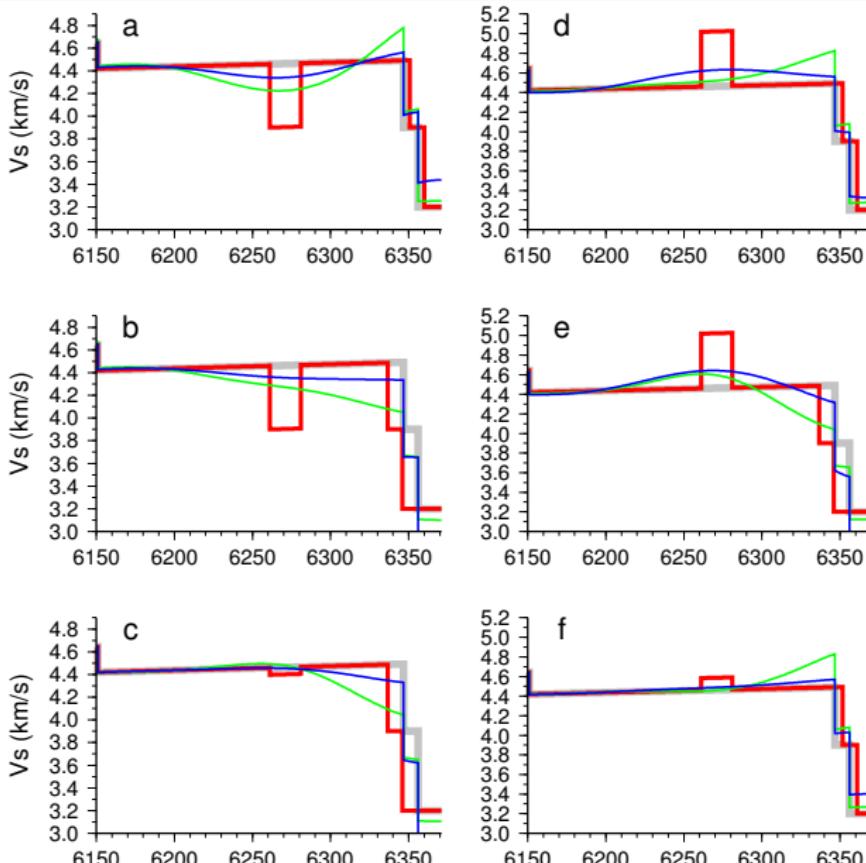
The inversion is designed such that the frequency band increases along with the iteration number.

Results for the  $\delta p_3$  residual ( $p_3 = \frac{1}{L}$ ) in the spectral domain for three iterations and therefore for 3 different frequency bands.



red line : target model ; black lines : inversion results for 8 different realizations of the 3% random noise

# Full waveform inversion : effect of a wrong a priori crust



# Conclusions and perspectives

## Conclusions

- So far, homogenization technique, by removing small scales, is useful for the forward problem (by easing the meshing problem for example).
- for the inverse problem, homogenization can allow to build a multi-scale inversion and a consistant parametrisation.
- an elastic model obtained from a FWI is indeed an homogenized version of the traget model (at least in the layered case)

## perspectives

- develop the down-scaling inverse problem (ongoing project)
- develop the higher dimensions (than layered) inverse problem (soon)
- up to what extent homogenization can lead to unicity of the inverse problem solution ?