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Eikonal tomography using coherent surface waves extracted from ambient noise by iterative matched filtering – Application to the large-N Maupasacq array

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Key Points:

- Coherent surface wave trains can be extracted from ambient noise by matched filtering
- Robust phase velocity maps are obtained by eikonal tomography from the extracted coherent wave trains
- The method is applied on noise recorded by the Maupasacq array to obtain phase velocity maps of the Arzacq-Mauleon basins for periods 2-9 s

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Abstract

Standard ambient noise tomography relies on cross-correlation of noise records between pairs of sensors to estimate empirical Green's functions. This approach is challenging if the distribution of noise sources is heterogeneous and can get computationally intensive for large-N seismic arrays. Here, we propose an iterative matched filtering method to isolate and extract coherent wave fronts that travel across a dense array of seismic sensors. The method can separate interfering wave trains coming from different directions, to provide amplitude and travel time fields for each detected wave front. We use the eikonal equation to derive phase velocity maps from the gradient of these travel time fields. Artefacts originating from scattered waves are removed by azimuthal averaging and spatial smoothing. The method is validated on a synthetic test and then applied to the data of the Maupasacq experiment. Rayleigh wave phase velocity maps are obtained for periods between 2 and 9 s. These maps correlate with surface geology at short period ($T < 3$ s) and reveal the deep architecture of the Arzacq and Mauleon basins at longer periods ($T > 4$ s).

Keywords:

Eikonal tomography, Ambient noise tomography, Surface waves, large-N arrays

1 Introduction

Massive deployments of cheap and easy-to-install geophone nodes, which were so far devoted to active source acquisitions for the oil and gas industry, have recently received increasing interest from the academic world for passive imaging studies (e.g., de Ridder & Dellinger, 2011; de Ridder & Biondi, 2015; Lin et al., 2013; Mordret et al., 2013; Hand, 2014). These large-N passive deployments open important new perspectives for the studies of active faults (Ben-Zion et al., 2015; Roux, Moreau, et al., 2016; Taylor et al., 2019; Y. Wang et al., 2019), reservoirs (Vergne et al., 2017), volcanoes (Hansen & Schmandt, 2015; Brenguier et al., 2016; Nakata et al., 2016), glaciers (Roux, Gimbert, et al., 2016), or landslides (J. Wang et al., 2008).

In the period range 1-10 s, i.e. in the so-called "secondary microseismic noise" band, the predominant source of noise is the interaction between oceanic waves and the solid earth, which produces surface waves that can be recorded by seismic stations located far away in the middle of continents (Longuet-Higgins, 1950). These surface waves in the

ambient noise have been extensively used for passive imaging of the crust (e.g., Shapiro et al., 2005; Lin et al., 2008, 2009; Ritzwoller et al., 2011). The standard approach consists in estimating empirical Green’s functions from the correlation of noise recorded by pairs of sensors (Lobkis & Weaver, 2001; Shapiro & Campillo, 2004). This approach has proven especially well suited for dense seismic arrays, because it can be fully automated and can thus be applied to massive data sets. Several variants of the method have been proposed to handle the case of an imperfectly homogeneous noise distribution (Curtis & Halliday, 2010; Wapenaar et al., 2011; Roux, Moreau, et al., 2016; Ermert et al., 2017) or to jointly invert for the velocity model and the source distribution (Yao & van der Hilst, 2009; Fichtner et al., 2017; Sager et al., 2018).

Alternatively, dense seismic arrays allow seismologists to measure directly the local velocity and propagation direction of a wave front. For example, the Progressive Multichannel Cross-Correlation method (PMCC, Cansi, 1995) estimates these parameters using phase differences computed over triangles of close stations. This method has been later adapted to large aperture arrays such as the USArray Transportable Array (Fan et al., 2018). Local phase velocities can also be estimated by gradiometry (de Ridder & Biondi, 2015) or by local beamforming (Roux & Ben-Zion, 2017). Unlike classical ambient noise tomography, these approaches remain applicable even if the distribution of sources is highly heterogeneous. However, since they rely on a limited subset of stations, they are more vulnerable to the effects of interfering arrivals, which can strongly bias the measurements.

In this study, we introduce a matched filtering method to isolate and extract surface wave fronts from the ambient noise using a dense seismic array. We then exploit the phase of the extracted wave fronts to derive phase velocity maps by eikonal tomography (Lin et al., 2009). We apply the method on the ambient noise field recorded by the large-N Maupasacq (Mauleon Passive Acquisition) seismic array.

The paper is organized as follows. In section 2, we describe the Maupasacq experiment and characterize the ambient noise field recorded in the 1-10 s period band. We find evidence for highly directive and energetic surface wave trains that are generated in the oceans. Section 3 details the matched filtering method that can isolate these coherent wave fronts. We determine phase velocity maps from the phase gradient of these surface wave fields with the eikonal equation. In section 4, the method is tested and val-

idated on a synthetic data set that mimics the Maupasacq acquisition. In section 5, it is applied to the Maupasacq data set to obtain phase velocity maps of the area for periods between 2 and 9 s. After a preliminary geological interpretation of our results, we discuss some perspectives opened by this new way to exploit data from dense seismic arrays.

2 Characterization of ambient noise recorded during the Maupasacq experiment

2.1 The Maupasacq experiment

The Maupasacq array (presented in Polychronopoulou et al., 2018) operated from March to October 2017 in the Mauleon basin (western Pyrenees, France). It includes a total of 442 temporary 3-component sensors (Fig. 1): 48 broad-band stations (BB) with a cut-off period of 20 s in complement to the 6 permanent broad band stations (PBB, cut-off period 100 s) available in the area, 197 short period stations (SP, cut-off period 1 s) and 191 geophone nodes (GN, cut-off period 0.1 s). The sampling rates are 100 samples per second for the BB and SP stations, and 250 samples per second for the GN. The inner part of the network forms a regular 50×30 km grid with an inter station spacing of about 7 km for the BB, and 3 km for the SP. The GN were installed along 5 lines in the WNW-ESE direction with a spacing of 1 km, and along 3 lines in the SSW-NNE direction with a spacing of 1.5 km. The dense part of the array is surrounded by two circles of BB and SP stations extending the total aperture to about 120×130 km.

[Fig. 1]

2.2 Characterization of the ambient noise field

In this work, we focus on the ambient wave field in the 1-10 s period band, which roughly corresponds to the period band of secondary microseismic noise. We first characterize the noise recorded by the Maupasacq array in the spectral, spatial and temporal domains.

2.2.1 Power spectral density and noise polarization

A spectrogram of the vertical component record at permanent station ATE during the Maupasacq experiment is shown in Figure 2a. Several higher energy patches emerge

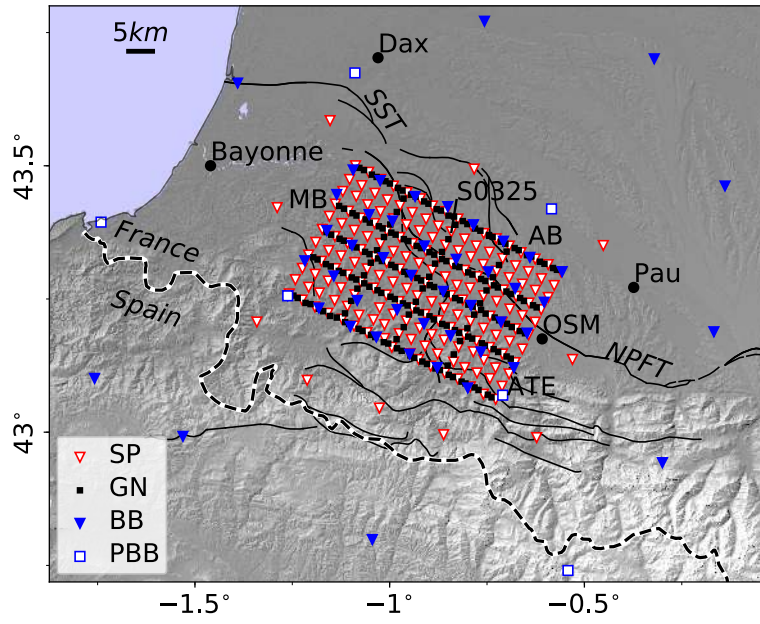


Figure 1. Map of the Maupasacq array. BB: broad-band, PBB: permanent broad-band, SP: short-period, GN: Geophone nodes. OSM: Oloron-Sainte-Marie, AB: Arzacq Basin, MB: Mauleon Basin, NPFT: North Pyrenean Front Thrust, SST: Sainte-suzanne Thrust. The solid lines indicate some of the main faults of the area after Saspiturry et al. (2019). The dashed line corresponds to the French-Spanish border.

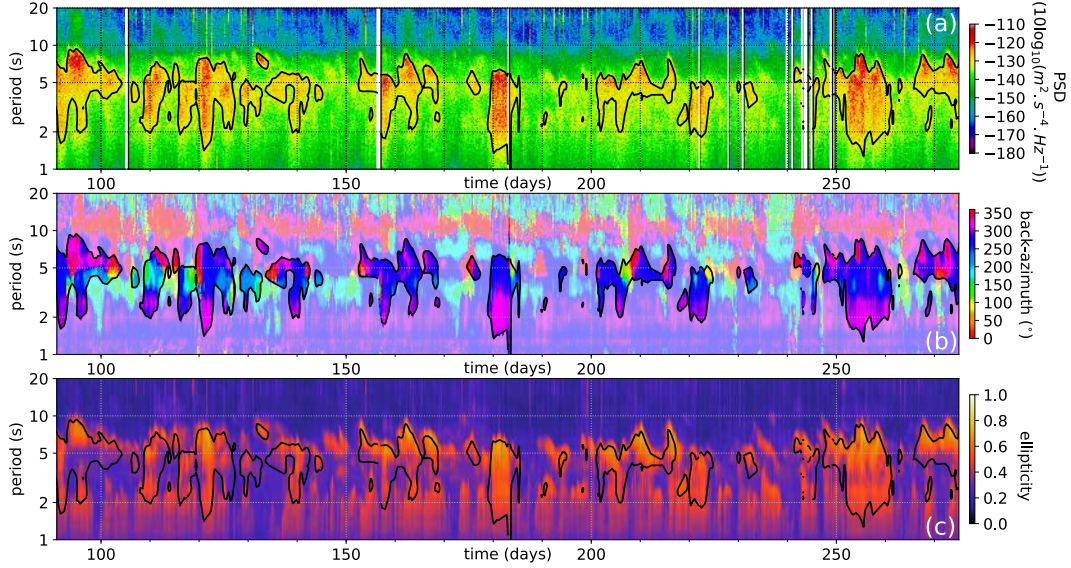


Figure 2. (a) Power spectral density of the vertical component of the permanent broad-band station ATE, with a time resolution of 1 hour, from April 1st to October 2nd, 2017. Amplitudes are expressed in dB relative to $1 \text{ m}^2 \cdot \text{s}^{-4} \cdot \text{Hz}^{-1}$. Black contours correspond to the -132 dB noise level. (b) Polarization back-azimuth averaged over all the broad-band stations during the same period. The measurements associated to low noise levels are shadowed using the same contours as (a). (c) Average ellipticity. A value of 1 corresponds to circular polarization, while a value of 0 corresponds to linear polarization.

in the period band of the secondary microseismic peak (Fig. 2a, contoured patterns). The upper period bound of these patches extends to about 10 s at the beginning and end of the recording period (e.g., days 95 or 255) and decreases to 6-7 s during summer (e.g., day 180). This change in noise amplitude and frequency content reflects the seasonal decline in the intensity and number of oceanic storms during summer in the northern hemisphere (Stutzmann et al., 2000).

[Fig. 2]

To further characterize the azimuthal distribution of the noise sources over time and period and to determine the wave type that dominates during the high energy patterns observed in the power spectral analysis, we estimate the dominant polarization state of the ambient noise for periods between 1 and 20 s using the BB and PBB stations (Figure 1). The three-component waveforms are cut into consecutive 1 hour windows. The seismic records are deconvolved from the instrumental response, detrended, tapered, and

down-sampled at 2.5 samples per seconds. Inside each one-hour window, we compute the ensemble average inter components cross-spectral matrix using a 400 s sliding window with 50% overlap. The dominant polarization is then determined from the eigenvalue decomposition of the ensemble average cross-spectral matrix at each period (Park et al., 1987). The complex eigenvector associated with the highest eigenvalue corresponds to the dominant polarization component of the signal. The real and imaginary parts of this vector correspond to the semi-major and semi-minor axes of the polarization ellipse, respectively. For each station, we estimate the polarization ellipticity as the ratio between the norms of the imaginary and real parts of the first eigenvector (Vidale, 1986). The propagation direction is estimated from the azimuth of the real part of the first eigenvector. The 180° ambiguity of the propagation direction is resolved by assuming a retrograde polarization. The polarization attributes are then averaged for all the broad band stations of the array in the time-period domain (Figure 2b, c). The average propagation direction is obtained by averaging the ellipticity orientation observed at each station:

$$\beta = \arg \left(\frac{1}{N} \sum_{j=0}^{N-1} e^{i\beta_j} \right), \quad (1)$$

where β_j is the back azimuth estimated at station j , in radians, and $\arg$ denotes the complex argument.

The high energetic patterns observed in the vertical component spectrogram (Fig. 2a) are characterized by higher average ellipticity (Fig. 2c), which confirms the predominance of Rayleigh waves. As can be seen in Fig. 2b, these waves originate mostly from a few specific directions in good agreement with the dominant noise directions previously reported by Chevrot et al. (2007) in the nearby Quercy region : N275° (Galicia), N340° (Northern Atlantic) and N130° (Mediterranean sea, see for example day 206).

2.2.2 Beamforming analysis

To complement the polarization study, we analyse the vertical component records using beamforming at several periods between 2 and 9 s. We introduce the plane wave beamforming operator (e.g., Jensen et al., 2000):

$$B(\mathbf{s}, \omega) = \frac{1}{N} \sum_{j=0}^{N-1} U(\mathbf{r}_j, \omega) e^{i\omega \mathbf{s} \cdot (\mathbf{r}_j - \tilde{\mathbf{r}})}, \quad (2)$$

where $\mathbf{s}$ is the horizontal slowness vector, ω the angular frequency, N the number of stations, $U(\mathbf{r}_j, \omega)$ the Fourier transform of the wave field observed at station j located in $\mathbf{r}_j$, and $\tilde{\mathbf{r}}$ the reference location, taken here as the center of the array. To account for the dispersive nature of surface waves, we evaluate the beam-power near a center pulsation $\tilde{\omega}$ using spectral integration of the band-pass filtered traces:

$$E(\mathbf{s}, \tilde{\omega}) = \frac{1}{2\pi} \int |G(\omega, \tilde{\omega}) \cdot B(\mathbf{s}, \omega)|^2 d\omega, \quad (3)$$

where G denotes the Fourier domain Gaussian band-pass filter centered on pulsation $\tilde{\omega}$ with a bandwidth controlled by a coefficient α (e.g., Bensen et al., 2007):

$$G(\omega, \tilde{\omega}) = e^{-\alpha(|\omega/\tilde{\omega}| - 1)^2}. \quad (4)$$

Figure 3 shows the results of beamforming for two different storms recorded between days 250 and 260 (a-d) and between days 205 and 208 (e-h) at different periods. The fundamental mode Rayleigh waves correspond to phase velocities near 2.6 km/s at 2 s period and 3.1 km/s at 9 s period. The propagation directions obtained by beamforming are in excellent agreement with the polarization analysis (Fig. 2a). In particular, prominent energetic arrivals are detected from back azimuths N275°, N340° and N130°. The relative amplitudes of these noise sources vary strongly as a function of the time interval considered and the period. For example, the signal recorded between days 250 and 260 (Fig. 2a) is dominated by noise sources from Galicia for periods between 2 and 6 s (Fig. 3b-d and blueish colors on Fig. 2b), while the northern direction dominates at periods near 9 s (Fig. 3a and pinkish colors on Fig. 2b). Secondary arrivals are also observed from the Mediterranean sea at periods below 3 s (Fig. 3c,d). These Mediterranean sources are rarely dominant. Most of the time they are masked by the energetic arrivals coming from the northern and western directions, which explains why they are under represented in the polarization diagram in Fig. 2b.

Beamforming analysis also revealed a weak but pervasive noise source with a dominant period around 7 s that produces waves with a very fast apparent phase velocity, indicating a nearly vertical incidence (e.g., Figure. 3e). Similar waves have been reported in the literature and attributed to distant oceanic sources (e.g., Landès et al., 2010; Meschede et al., 2017; Pedersen & Colombi, 2018). They could be responsible for the spurious large

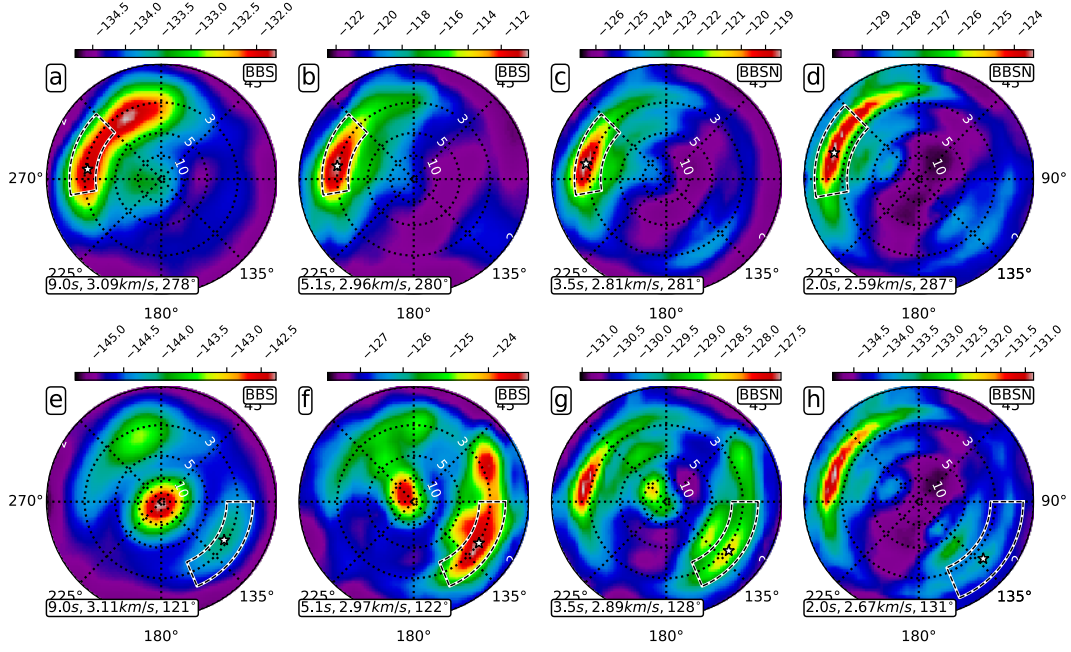


Figure 3. Beam power estimates for two storms recorded from days 250 to 260 (a-d) and days 205 to 208 (e-h) for several periods between 2 and 9 s. The amplitudes are expressed in dB relative to $1 \text{ m}^2 \cdot \text{s}^{-2} \cdot \text{Hz}^{-1}$. The angles correspond to back-azimuths in degrees and the radial ticks correspond to phase velocities in km/s. All the station are used for the periods between 2 and 3.5 s (label BBSN) and only the BB and SP stations are used for the periods above 3.5 s (label BBS). The white stars indicate the local maxima found in the white dashed areas interpreted as the fundamental mode Rayleigh wave from the Galicia direction (a-d) and from the Mediterranean direction (e-h). The period, phase velocity and back-azimuth of these maxima are indicated in the legend of each diagram.

amplitude arrival near the zero lagtime that is often observed in noise correlation functions (Villaseñor et al., 2007; Taylor et al., 2019).

[Fig. 3]

2.2.3 Coherent surface wave trains from the oceans

In the time domain, the high energy patterns detected in the spectrogram (Fig. 2a) correspond to coherent wave trains that cross the entire array. Fig. 4a,c display the waveforms recorded during day 181 at periods 5.2 s and 3 s (see the shaded traces), where the noise comes mostly from the Galicia direction (Fig. 2b). To highlight the phase align-

ments of the waves emitted by the dominant noise source, the seismic traces are sorted as a function of increasing distance measured along the propagation direction determined from beamforming. The black solid lines in Fig. 4a,c correspond to the reference wavelets obtained at the center of the array, using the iterative matched filtering algorithm detailed below.

2.2.4 Concluding remarks on the nature of the ambient wave field

We conclude from these detailed observations that the ambient wave field in the 1-10 s period band is a superposition of coherent quasi plane waves coming from a few well-defined dominant directions and not a purely diffusive wave field. The relative amplitudes of the dominant noise sources strongly vary with the time interval and with the period considered.

[Fig. 4]

3 Methods

3.1 Extraction of coherent wave trains with an iterative matched filtering approach

We now propose a method to separate and extract the coherent Rayleigh wave fields that we have identified in the ambient noise. The method is applied to consecutive and non overlapping one-hour long time windows, detrended, deconvolved from the instrument response and filtered with a Gaussian band-pass filter near a central pulsation $\tilde{\omega}$. No temporal or spectral normalization (e.g., Bensen et al., 2007) is applied to the data. The algorithm involves three successive steps, which are described in more details in the following:

- A beamforming, which allows us to estimate the apparent slowness and back azimuth of the dominant noise source. These beam parameters are used to obtain an initial reference wavelet at the center of the array.
- An iterative matched filter. This step iteratively improves the reference wavelet and the values of time and amplitude fields at each station.
- A subtraction of the matched wave field from each trace.

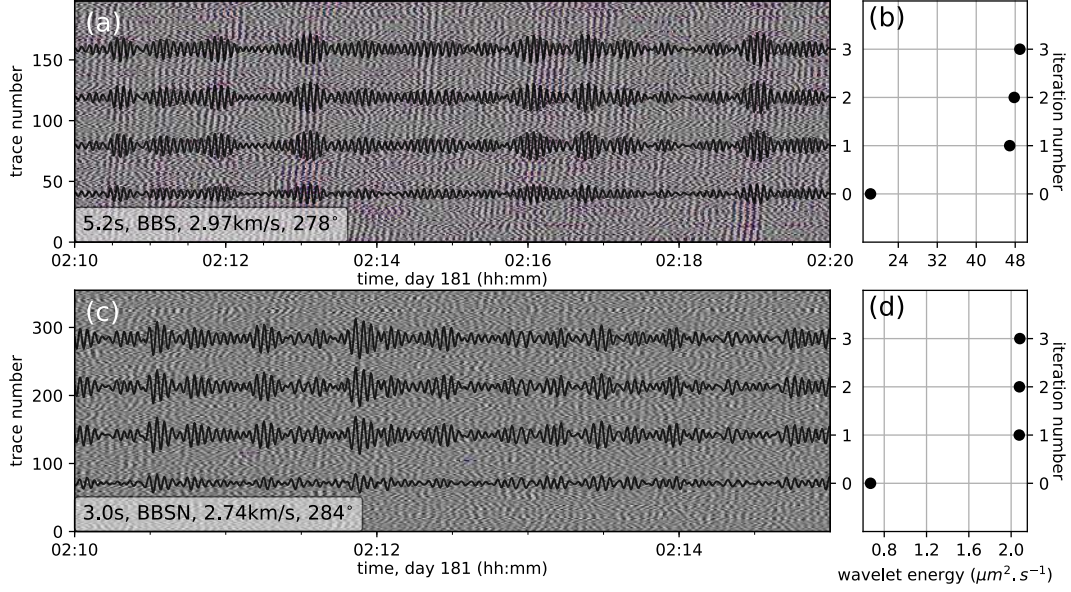


Figure 4. Oceanic wave trains and convergence of the reference wavelet. (a) The background gray-scaled field corresponds to the vertical component waveforms recorded by the PBB, BB and SP stations on Julian day 181. The traces are filtered around 5.2 s period using a Gaussian filter and sorted by increasing distance along the azimuth of the source estimated from plane-wave beamforming (step 1). The black traces show the reference wavelets obtained at each iteration of the matched filtering approach (step 2). (b) Energy of the reference wavelet shown in (a) as a function of the iteration number. (c, d), same as (a, b) near 3 s period using the PBB, BB, SP and GN stations.

This algorithm is iterated until all the coherent wave trains are extracted and the remaining wave field is incoherent.

3.1.1 Step 1: Beamforming and initial wavelet estimate

In the first step, we determine the slowness vector $\mathbf{s}_{max}(\tilde{\omega})$, which maximizes the plane-wave beam-power in the neighborhood of the center angular frequency $\tilde{\omega}$ (eq. 3). We obtain an initial estimate of the reference wavelet by forming the beam corresponding to the dominant plane wave

$$w_0(\tilde{\mathbf{r}}, t) = \frac{1}{N} \sum_{j=0}^{N-1} u(\mathbf{r}_j, t + \mathbf{s}_{max}(\tilde{\omega}) \cdot (\mathbf{r}_j - \tilde{\mathbf{r}})) \quad (5)$$

where subscript 0 indicates the first iteration, $u(\mathbf{r}_j, t)$ is the wave field observed at station j located in $\mathbf{r}_j$, filtered with a Gaussian filter centered at pulsation $\tilde{\omega}$ (eq. 4), and $\tilde{\mathbf{r}}$ is the reference location, chosen at the center of the array.

3.1.2 Step 2: Iterative matched filtering

Matched filtering is a method used to search for a known signal in a data stream. It is a concept commonly employed in engineering to decide if and where a reference signal occurs in a background noise (e.g., radar or sonar return signals, Turin, 1960). In seismology, the method has been introduced to measure relative time delays and amplitude ratios between observed and modeled seismograms (Sigloch & Nolet, 2006). Here, we use this approach to search for the delayed reference wavelet in each trace. To do this, we cross correlate the reference wavelet with each recorded trace. The correlation function normalized by the energy of the reference wavelet, hereafter referred to as the Matched Filter Correlation Function (MFCF) is defined as:

$$c_n(\mathbf{r}_j, t) = \frac{w_n(\tilde{\mathbf{r}}, t) * u(\mathbf{r}_j, t)}{\int w_n(\tilde{\mathbf{r}}, t)^2 dt}, \quad (6)$$

where $*$ denotes the time domain cross correlation operator and n the iteration number. The denominator in (6) is a normalization constant which ensures that the amplitude of the MFCF at the maximum lag time is the relative amplitude of the extracted wave at station j relative to the reference location.

231 The MFCFs verify the wave equation because the correlation of a wave field that
 232 verifies the wave equation with any other spatially invariant signal still satisfies the wave
 233 equation (Lin et al., 2013, their equation 3). By the linearity of the correlation function,
 234 this assertion holds true if u is the superposition of several wave fields verifying the wave
 235 equation.

236 The maximum of the MFCF occurs at lag-time $T_n(\mathbf{r}_j, \tilde{\omega})$, which corresponds to the
 237 time delay between the station j and the reference location in the direction of propaga-
 238 tion of the dominant coherent wave. We then relax the plane-wave assumption. The traces
 239 are realigned with the new time delays and stacked to determine a new reference wavelet
 240 :

$$w_{n+1}(\tilde{\mathbf{r}}, t) = \frac{1}{N} \sum_{j=0}^{N-1} u(\mathbf{r}_j, t + T_n(\mathbf{r}_j, \tilde{\omega})) \quad (7)$$

241 that will be used at iteration $n+1$ to recompute the MFCF according to eq. (6). The
 242 process is iterated until convergence, i.e. until the energy of the reference wavelet reaches
 243 a plateau, which usually occurs after a few iterations.

244 Examples of wavelets obtained at the successive iterations of the matched filter-
 245 ing algorithm are shown in Fig. 4a,c (black solid lines). The right part of Fig. 4 (b,d)
 246 shows the variation of the wavelet energy as a function of the iteration number. The fi-
 247 nal wavelet energy is usually 2 to 4 times larger than the energy of the initial plane-wave
 248 estimate. This increased energy results from an improved alignment of the seismic traces,
 249 which accounts for the lateral variations of phase velocity beneath the Maupasacq ar-
 250 ray. From the last version of the MFCF, noted c_f , we get the travel time and amplitude
 251 at each station, measured as the maximum lag time and maximum amplitude of c_f

$$\begin{cases} T(\mathbf{r}_j, \tilde{\omega}) = \operatorname{argmax}\{t \rightarrow c_f(\mathbf{r}_j, t)\} \\ A(\mathbf{r}_j, \tilde{\omega}) = c_f(\mathbf{r}_j, T(\mathbf{r}_j, \tilde{\omega})) \end{cases} \quad (8)$$

252 Figure 5 illustrates the travel time and amplitude of the dominant wave field at 5.2 s pe-
 253 riod extracted from a one-hour long noise time window.

254 [Fig. 5]

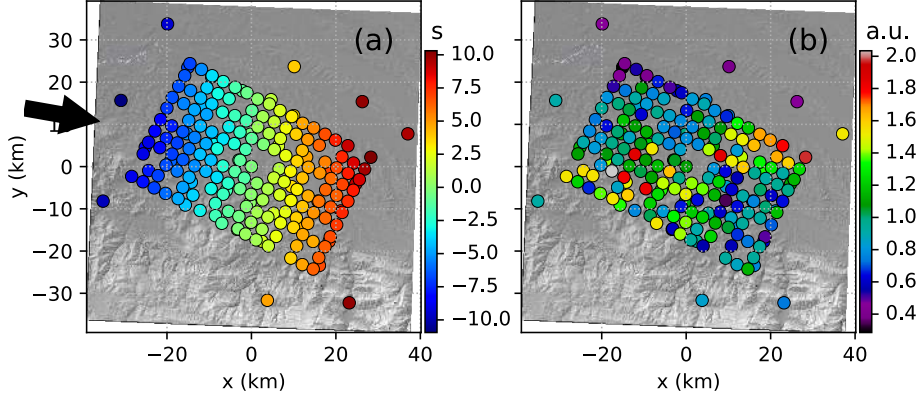


Figure 5. Travel time (a) and amplitude fields (b) of the MFCFs at 5.2 s period obtained for one hour of signal recorded on Julian day 181, between 2 and 3 a.m. (UTC) by the BB, PBB and SP stations. The amplitude and travel time fields correspond to a dominant noise source from back-azimuth 277° (black arrow) detected at the first iteration of the 3-step procedure.

255 *3.1.3 Step 3 : Subtraction of the dominant wave field from the signal*

256 The last step of the algorithm consists in subtracting the wave field matched to the
 257 reference wavelet from the traces. The matched wave field is built from a time shifted
 258 and scaled version of the reference wavelet at each station:

$$u(\mathbf{r}_j, t) = A(\mathbf{r}_j, \tilde{\omega}) \times w(\mathbf{r}_j, t - T(\mathbf{r}_j, \tilde{\omega})). \quad (9)$$

259 The 3-step algorithm is then re-applied to the residual wave field, to search for secondary
 260 coherent wave fronts.

261 *3.1.4 Average MFCF*

262 For each one-hour long time window and each iteration of the 3-step matched fil-
 263 tering procedure, we obtain a collection of MFCFs associated to a dominant average wavenum-
 264 ber vector. The MFCFs obtained for similar wave vectors are averaged in order to in-
 265 crease the signal to noise ratio and minimize the cross-terms due to fortuitous correla-
 266 tions between uncorrelated noise sources (Snieder, 2004).

3.2 Regularized eikonal tomography

In practice, the travel time field is only known at the locations of the seismic stations. Computing the spatial gradient of the travel time field thus requires spatial interpolation, for example using splines in tension (Lin et al., 2009; Mordret et al., 2013). But this approach gives only poor control on the value and smoothness of the velocity model. Here we formulate a regularized interpolation to search for the travel time field θ discretized on a regular Cartesian grid that minimizes the misfit with the observed travel times at each station, and such that its gradient leads to a phase slowness model that is close to an a priori slowness model. The cost function of the problem is:

$$\begin{aligned}
 f(\theta) = & \sum_{j=0}^{N-1} (T(\mathbf{r}_j) - (L.\theta)_j)^2 + \alpha^2 \sum_{m=0}^{M-1} (\|\nabla\theta\|_m - s_{prior\ m})^2 \\
 & + \beta^2 \sum_{m=0}^{M-1} (\Delta\theta_m)^2 + \gamma^2 \sum_{m=0}^{M-1} (\Delta\|\nabla\theta\|_m)^2.
 \end{aligned} \tag{10}$$

In this expression, $T(\mathbf{r}_j)$ is the travel time field observed at each station, $\mathbf{r}_j$ the location of the j^{th} station, $(L.\theta)_j$ the travel time estimated at station j using the bi-linear interpolation operator L , s_{prior} the vector containing the prior phase slowness model over the same grid as θ . The dependency of θ , T and s_{prior} to the center pulsation $\tilde{\omega}$ is dropped for the sake of simplicity. The first term in eq. (10) quantifies the misfit between the predicted and observed travel times. The second term measures the misfit between the eikonal phase slowness derived from the inverted parameter using the eikonal equation $\|\nabla\theta\|$ and the prior slowness model s_{prior} . The last two terms are penalty constraints that are introduced in order to minimize the total curvature of the travel time and slowness models respectively, which are quantified with the norm of the Laplacian (Smith & Wessel, 1990). The parameters α , β and γ control the penalty constraints that regularize the solution. The inverse problem (10) is solved with a conjugate gradient algorithm. The travel time model θ used to initiate the inversion is computed in the prior slowness model s_{prior} using the fast marching method (FMM, Sethian, 1996). The inversion is first applied to the longest period using a homogeneous prior model. The period is then progressively decreased, using the solution of the inversion at the current period as an a priori for the next period. The regularization terms in eq. (10) allow us to smooth the inverted travel time (third term in eq. (10)) and the corresponding eikonal tomographic model (fourth term in eq. (10)), which mitigates the influence of the outliers and improves the accu-

racy of the velocity model by canceling the interference between the incident and scattered wave fields. We adjust these regularization parameters by a systematic grid search. The criterion is to obtain smooth travel time and slowness models that provide a good fit of travel time data.

4 Validation of the matched filtering and eikonal tomography methods on a synthetic test

4.1 Synthetic experiment set-up

We now validate our method with a numerical experiment. The objective of this synthetic test is to demonstrate the ability of the matched filtering method to measure travel time and amplitude fields on waves that simultaneously arrive from different directions and interfere in the time domain.

We consider one-day long synthetic noise time series containing 5 s Rayleigh waves propagating in the structural phase velocity model shown in Fig. 6a. The synthetic noise records are built using temporal wavelets obtained by bandpass filtering white noise series around 5 s period. These noise series are then shifted in time and scaled in amplitude to simulate the propagation of the waves across the array. The phase travel time shifts and amplitude corrections at each station are obtained by resolving the scalar wave equation with a finite difference method for each noise source. As a boundary condition, we impose the displacement produced by an incident plane wave coming from the azimuth of the noise source with a phase velocity of 3.0 km/s. We implement Stacey boundary conditions to absorb the outgoing scattered wave field (Stacey, 1988). Fig. 6b,c show the phase travel time and amplitude fields modeled for an incoming wave with a back-azimuth of 280° . We apply independent phase and envelope travel time shifts in the frequency domain. The signal is reformed afterwards in the time domain. For the envelope shifts, we use the group travel times estimated with the FMM algorithm in the synthetic group velocity model shown in Fig. 7.

We add the two synthetic noise records modelled for back-azimuths 280° and 130° to simulate the interference of the Atlantic Ocean and Mediterranean Sea noise sources. The amplitude of the Mediterranean source is taken as one third of the Atlantic source. Additional incoherent noise in the form of random Gaussian noise with an amplitude of

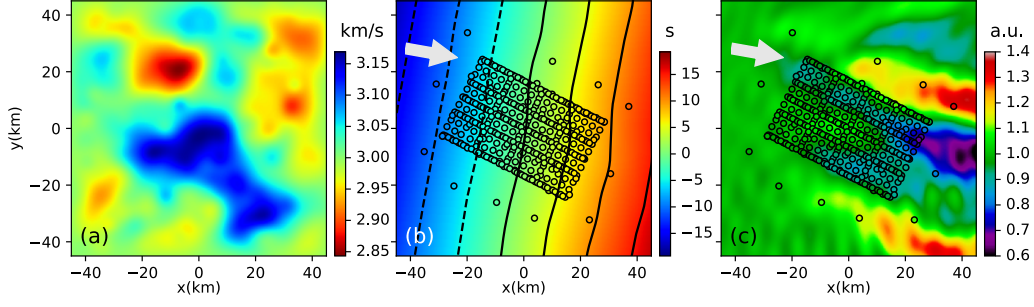


Figure 6. Synthetic surface wave modeling at period 5 s. (a) Synthetic structural velocity model for the fundamental mode Rayleigh wave at period 5 s, expressed in km/s. (b) Travel time of the modeled wave field relative to the center of the array (T), obtained by unwrapping the phase field from the solution of the wave equation. Travel time isovalues are indicated with black lines. (c) Amplitude of the modeled wave field, (A) in arbitrary units. The white arrows indicate the back azimuth of the incident wave (N280°). The dots indicate the station locations.

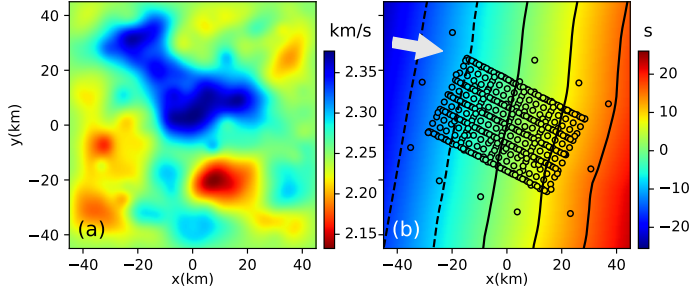


Figure 7. Synthetic surface wave modeling at period 5 s. (a) Group velocity synthetic model for the fundamental mode Rayleigh wave at period 5 s, expressed in km/s. (b) Travel times of the Rayleigh wave envelope relative to the center coordinate in seconds, computed with the FMM algorithm for an incident plane wave from back azimuth 280°.

325 20 % of the Atlantic noise source is added to each trace to simulate instrumental noise
 326 or local uncorrelated noise sources.

327 [Fig. 6]

328 [Fig. 7]

329 4.2 Results

330 The 3-step matched filtering method is applied to the 1-day synthetic noise series,
 331 cut into consecutive and non-overlapping 1-hour time windows. In each window, we re-

cover the azimuth of both the primary and secondary wave fronts after the first two iterations of the 3-step procedure (i.e. back-azimuths N280° and N130°, not shown). The phase travel times of the average MFCFs obtained at each station for the Atlantic and Mediterranean directions are shown in Figs. 8a,b. They are in good agreement with the expected travel time fields (8c,d) obtained from the phase of the modeled wave fields for these two directions as in Fig. 6b. The lateral distribution of the amplitudes at each station from the average MFCFs also reproduces the expected ones for the two directions imposed (Fig. 9).

We invert the travel time fields at each station to obtain the eikonal phase velocity models for the two directions using eq. (10) (Fig. 10a,b). We compare them with the expected eikonal models, determined from the gradient of the expected travel time fields of Figs. 8c,d. The expected velocity models are smoothed laterally to ensure a fair comparison with the inverted phase velocity maps (Fig. 10b,d). The differences between the expected phase velocity models for back azimuths 280° and 130° are attributed to the errors introduced by the eikonal approximation, which discards the information carried by the amplitude field. This test validates the ability of our method to separate interfering wave fronts emitted by uncorrelated noise sources. Note that the method is also able to separate coherent wave fronts from incoherent background noise.

[Fig. 8]

[Fig. 9]

[Fig. 10]

Figure 11a shows the velocity model obtained by averaging the Atlantic and Mediterranean sources eikonal models. It is in very good agreement with the true structural model shown in Fig. 11c, at least beneath the dense part of the seismic array. The strong similarity between the eikonal model derived from a complete azimuthal coverage (Fig. 11b) and the true structural model (Fig. 11c) demonstrates that in practice eikonal tomography is sufficient to obtain accurate and robust phase velocity models provided a good azimuthal coverage.

[Fig. 11]

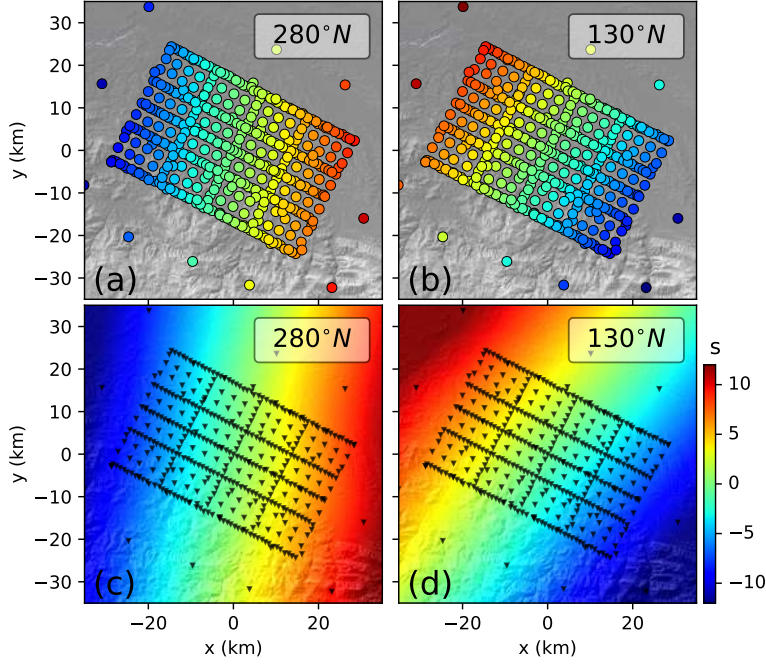


Figure 8. Results of the synthetic interfering wave fronts test for travel times. (a, b) phase travel time of the MFCFs at each station for the extracted dominant Atlantic beam (N280°, a) and secondary Mediterranean beam (N130°, b), (c, d) expected travel time fields from the solution of the wave equation for an input plane wave coming from the Atlantic (c) and Mediterranean sea (d).

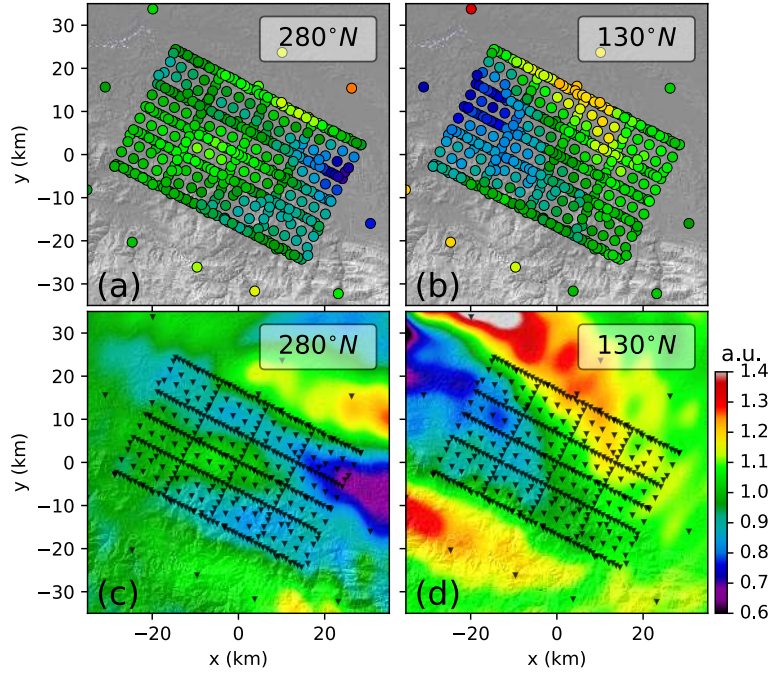


Figure 9. Results of the synthetic interfering wave fronts test for amplitudes. (a, b) amplitude of the MFCFs at each station for the dominant Atlantic beam (N280°, a) and secondary Mediterranean beam (N130°, b), (c, d) expected amplitude fields from the solution of the wave equation for an input wave plane coming from the Atlantic (c) and Mediterranean sea (d).

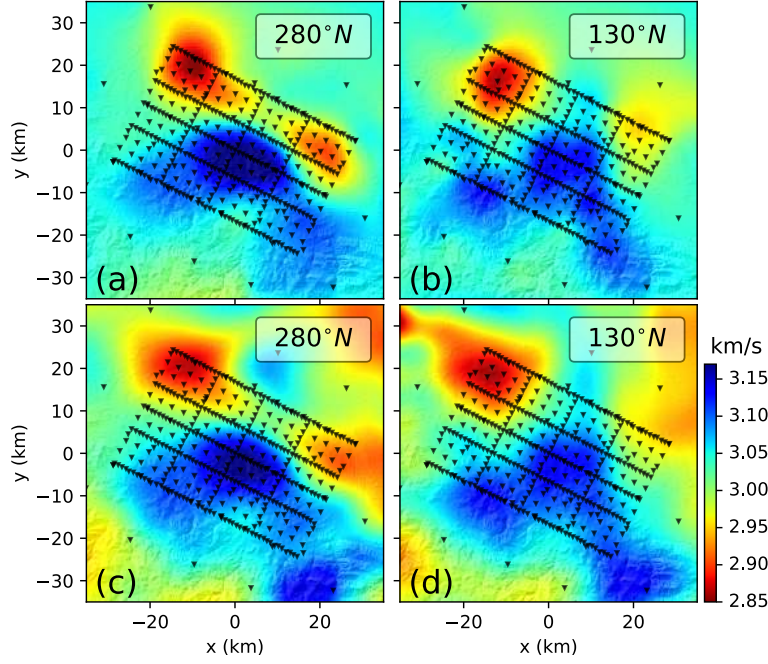


Figure 10. (a, b) eikonal phase velocity models inverted from the travel times shown in Fig. 8a and b respectively, (c, d) expected smooth eikonal phase velocity models obtained for the Atlantic (c) and Mediterranean sea (d) directions.

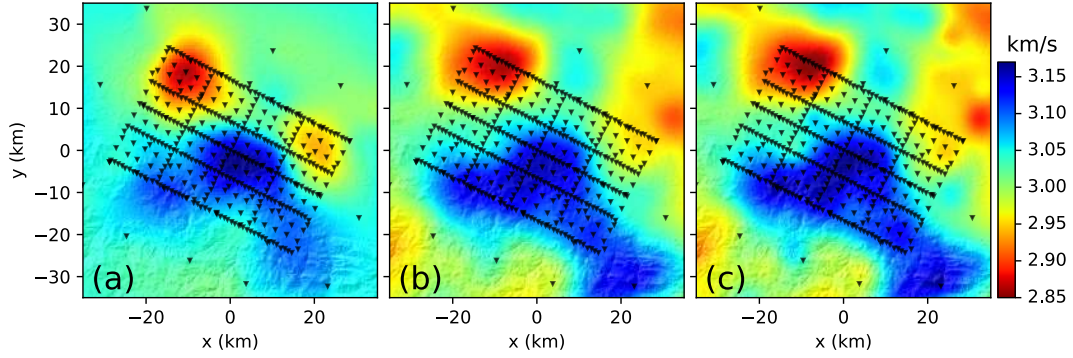


Figure 11. (a) Eikonal phase velocity models obtained by averaging models from directions N280 and N130 (10a,b). (b) Eikonal phase velocity model obtained from a complete azimuthal coverage. (c) True structural phase velocity model.

5 Application to the Maupasacq dataset

5.1 data pre-processing

We pre-process the vertical component records as for the polarization analysis presented above. After removing the anomalous traces, the seismograms are deconvolved from their instrumental response, detrended, re-sampled at 2.5 samples per second, and cut into one-hour non overlapping time windows.

5.2 Extraction of coherent wave trains by iterative matched filtering

We apply the matched filtering approach at a number of periods spaced logarithmically between 2 and 9 s. To mitigate the instrumental noise level in the seismograms, we use only the broad-band stations for periods above 7 s (Fig. 1, BB and PBB), the short period and broad-band stations for periods between 3 and 7 s (BB, PBB and SP) and all the stations for periods below 3 s (BB, PBB, SP and GN). Inside a given time window, we can usually extract several (sometimes up to 10) coherent wave fronts. We only keep the wave fronts whose average phase velocity corresponds to the fundamental mode Rayleigh wave. The phase velocity ranges used to isolate the Rayleigh wave component are 2.8-3.2 km/s at period 5.6 s, 2.6-3.0 km/s at period 3.48 s and 2.4-2.8 km/s at period 2.34 s (Fig. 12). The first detections (dark red dots on Fig. 12) are in good agreement with the dominant noise sources previously identified with the polarization and beamforming analyses. Interestingly, the later detections, which correspond to less energetic noise sources, often come from other back-azimuths. Thus, by iterating and exploiting all the coherent but less energetic arrivals, we can improve the azimuthal coverage very significantly. However, very few arrivals are detected in the North-East and South-West directions.

5.3 Results of matched filtering

The MFCFs obtained for similar directions are averaged over 5° azimuthal bins to increase the signal to noise ratio. Further details about the temporal variation and the convergence rate of the MFCF are given in the supplementary materials. The MFCFs obtained at each station for period 5.2 s for the two back-azimuths N280° and N130° are displayed on Fig. 13a,d. The MFCFs are ordered as a function of their distance from the noise source. The origin time is defined as the onset time at the center of the array. The

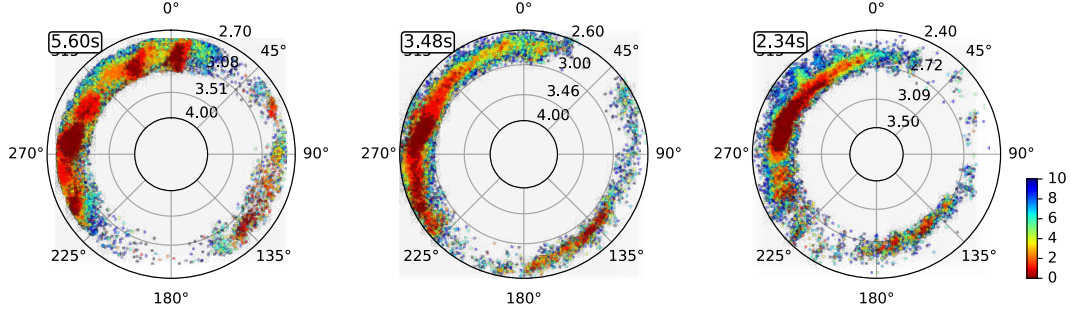


Figure 12. Detections of coherent fundamental mode Rayleigh wave trains during the 6 months of the Maupasacq experiment at three different periods: 5.60 s (left), 3.48 s (middle), and 2.34 s (right). Each dot corresponds to one extraction inside a 1-hr time window. The colors indicate the extraction rank (0 stands for the first extraction). The radial ticks correspond to phase velocities in km/s.

travel time and amplitude fields derived from the averaged MFCF at each station for the two directions are shown respectively on Fig. 13b, c and 13e, f. Movies showing the propagation of the extracted surface wave fronts across the network for several back azimuths and periods are given in the supplementary materials.

Significant distortions of the travel time field can be observed as a result of lateral variations of velocity beneath the array. For example, the wavefront curvature observed for the back azimuth N280° (Fig. 13b) in the north-eastern part of the array reveals the low velocity anomaly of the Arzacq basin. We also observe significant amplitude variations within the array, ranging from about 0.25 to 2 at period 5.2 s (Fig. 13c,f). The lateral distribution of the amplitudes vary with the incoming direction, which suggests that this distribution is firstly controlled by the heterogeneities that are crossed by the waves, rather than local structural amplifications. For example, the higher amplitudes observed in the Arzacq basin for the N280° direction (Fig. 13c) results from the focusing of the wave front produced by the negative velocity anomaly in the North-East corner of the array.

[Fig. 13]

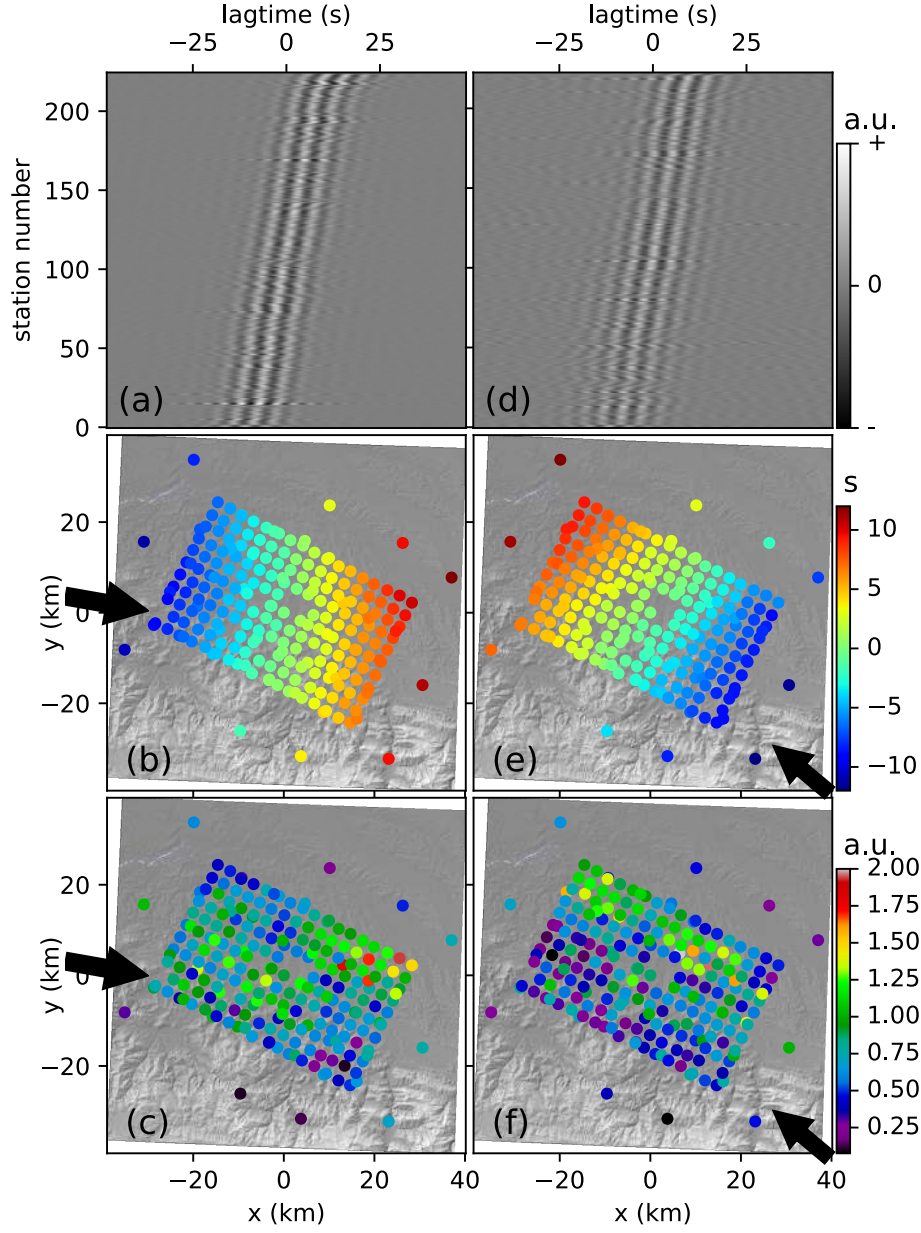


Figure 13. Matched filtering results obtained for the broad-band and short period sensors at period 5.2s, for the two directions 280° (a-c) and 130° (d-f) (black arrows) averaged over the full recording period. (a) Average MFCFs. (b) Travel time field obtained from the maximum lag time of the average MFCFs, expressed in s. (c) Amplitude field derived from the average MFCFs. (d-f) same as (a-c) for a the 130° back-azimuth.

5.4 Phase velocity maps

We use the regularized interpolation approach to estimate the eikonal phase velocity dispersion map for each back azimuth from the average MFCFs (as exemplified on Fig. 13b, e). We start the inversion using the wave fronts originating from the Atlantic direction (from the average MFCFs obtained for the back-azimuths between N280° and N285°) at the longest period (9 s) assuming a homogeneous 3.1 km/s prior model based on the average slowness measured by beamforming. The regularization parameters (α, β, γ) in eq. (10) are adjusted to balance all the terms of the misfit function. The resulting phase slowness model is then used as the prior model for the next period (8.32 s) shifted by a constant slowness value to fit the mean slowness of the detected wave fronts at period 8.32 s. The procedure is repeated for the 20 periods spaced logarithmically down to period 2 s. The phase slowness models obtained at each period for the incoming Atlantic direction are then used as prior models to invert the travel time fields in the other directions. The azimuthal average of the phase velocity models obtained are shown in Fig. 14.

[Fig. 14]

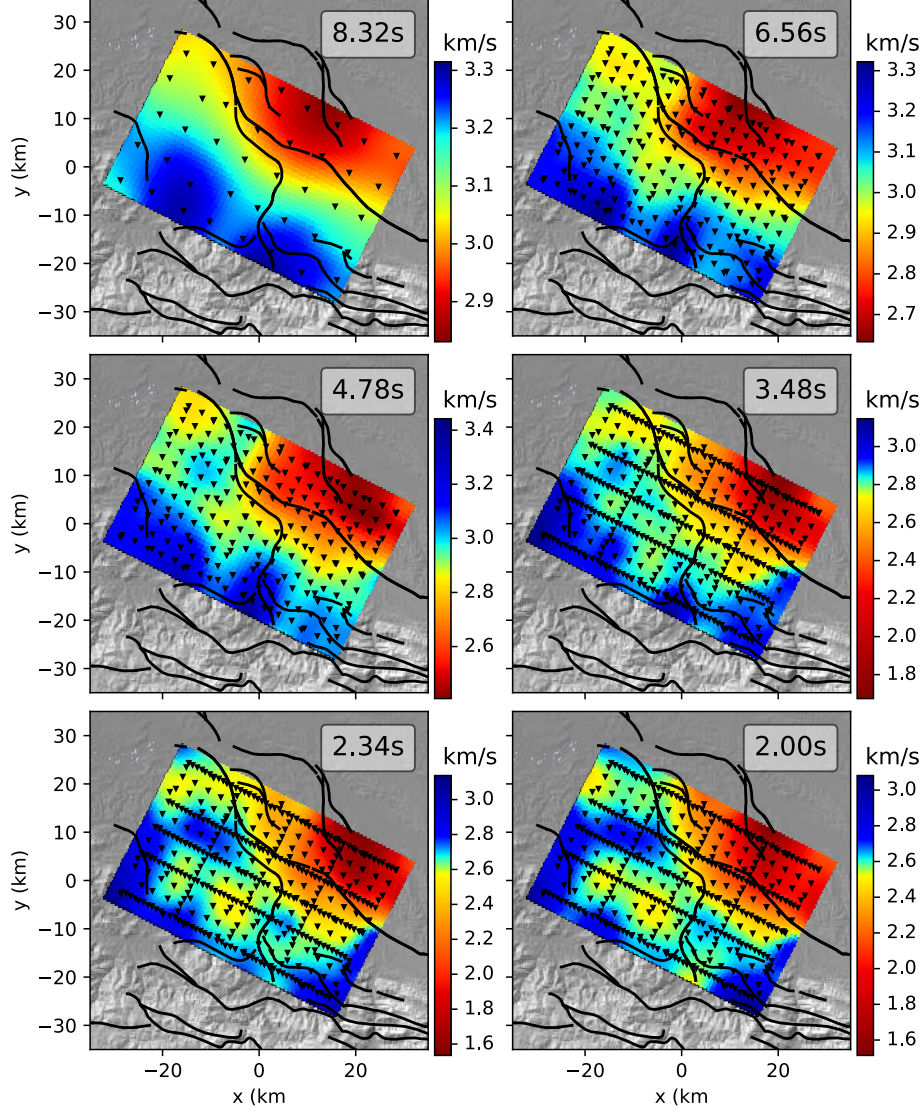


Figure 14. Rayleigh wave phase velocity models obtained at 6 periods between 9 and 2 s, in km/s.

6 Discussion

6.1 Comparison between matched filtering coherent wave field imaging and standard ambient noise tomography

The matched filtering method shares many similarities with standard ambient noise tomography. However, while both methods exploit the surface waves present in the ambient noise field for passive imaging, they do differ in several aspects.

First, in contrast to standard ambient noise tomography, the matched filtering approach does not rely on the implicit assumptions of a diffusive ambient noise field (Lobkis & Weaver, 2001) or of a spatially uniform distribution of uncorrelated noise sources (Roux et al., 2005). Instead, it exploits the strong directivity of noise sources to separate and extract coherent wave trains. While ambient noise correlation can usually provide relatively accurate travel time information even with heterogeneous noise source distributions or imperfectly diffuse wave fields, the matched filter approach better exploits the strong directivity of ambient noise fields. We have shown that the matched filtering method remains valid in the case of interfering wave trains originating from different directions, as observed with the Maupasacq array. The iterative extraction of coherent waves allows us to exploit sources from a broad range of azimuths thanks to the secondary noise sources that are most of the time buried beneath the predominant Atlantic and Mediterranean noise sources. Since the azimuthal averaging of eikonal phase velocity map can mitigate the error due to the eikonal approximation (Bodin & Maupin, 2008), we believe that the improved azimuthal coverage will ultimately lead to more robust isotropic tomographic models with less artefacts.

Second, the standard noise correlation method often implies some aggressive non-linear pre-processing filters like spectral whitening or temporal normalization (Bensen et al., 2007), which annihilate the amplitude information. Although it is usually admitted that those filters should be avoided for applications that try to exploit the amplitude information, no clear consensus can be found in the literature about the optimal pre-processing to simultaneously ensure a good reconstruction of the amplitudes and convergence of the noise correlations (e.g., Mordret et al., 2013; Liu et al., 2015; Bowden et al., 2015, 2017). With the matched filtering approach, since we normalize the correlations by the energy of the reference wavelet, the amplitude information is naturally preserved. The amplitude fields obtained by matched filtering should open important per-

spectives to better constrain attenuation and seismic anisotropy beneath dense local or regional seismic arrays.

Third, in ambient noise tomography, exploiting the noise correlation functions for inter-station distances smaller than a few wavelengths is generally challenging. In that case, surface waves are observed in the near field, which hinders the estimation of phase and group travel times, and errors due to heterogeneous noise source distributions become prominent (Weaver et al., 2009). For these reasons, close station pairs are often discarded (Bensen et al., 2007), which has a detrimental effect on the spatial coverage. The matched filtering approach does not suffer from this limitation.

Fourth, in the matched filtering approach, the separation and extraction of the different coherent wave trains scale with the number of sensors, while noise correlation scales with the square number of sensors when considering all the station pairs. Therefore, matched filtering can be used to obtain a tomographic image with lower computational cost. On the other hand, since noise correlations exploit the redundancy of information between a large number of station pairs, this can reduce the errors in the travel time estimates.

The main limitation of the matched filtering approach comes from the geometry of the acquisition. It first needs to be sufficiently dense to sample the wave field without spatial aliasing. In addition, the aperture of the array must be sufficiently large to be able to separate wave fields originating from nearby directions. In other words, the results of matched filtering will be controlled by the array response function at the periods considered (see supplementary information).

6.2 Going from eikonal to Helmholtz tomography

Despite the long term averaging of the MFCFs (over 6 months), the amplitude fields obtained are still noisy (Fig. 13c, f). A first source of problem is probably due to the use of different types of instruments and from an imperfect knowledge of their responses, especially for the SP sensors. In addition, since the periods considered in this study are well below the cut-off period of the SP and GN sensors, we have to face a significant level of instrumental noise. This will limit our capacity to exploit the amplitudes with Helmholtz tomography (Lin & Ritzwoller, 2011). In any case, amplitude fields still constitute untapped important sources of information that should allow us to further improve over the results of eikonal tomography in the near future.

6.3 Lateral variations of phase velocities beneath the Maupasacq array

The phase velocity maps show strong lateral variations, revealing the complexity of the structures in the upper 10 km. The main shallow feature is the low velocity anomaly in the north-eastern part of the model, which coincides with the Arzacq basin, a Tertiary unit separated from the Mauleon basin by the North Pyrenean Front Thrust (Fig. 1). The Mauleon basin, to the south west, is filled with Cretaceous calcareous and flysch rocks, which are characterized by much higher phase velocities. The low velocity anomaly observed in the south-western part of the Mauleon basin for periods shorter than 3.5 s are related to a large syncline approximately oriented along a WNW-ESE direction. It is nicely correlated with shallow post and syn-rift sedimentary deposits visible in the surface geological maps of the area (e.g., Masini et al., 2014). The phase velocity maps at periods longer than 4 s show a different and simpler pattern. In these maps, a sharp transition is observed which marks the limit between a high velocity domain in the south and a slower domain in the northernmost part of the Mauleon basin. Interestingly, this high velocity region is found on the top of the fast velocity anomaly imaged in Y. Wang et al. (2016), which has been interpreted as a mantle body exhumed during the Cretaceous episode of rifting. This would suggest that the top of the basement is at a much shallower depth in the southern part of the Mauleon basin, in good agreement with the geological sections recently published for this area (e.g., Saspiturry et al., 2019; Gómez-Romeu et al., 2019).

7 Conclusions

We analyzed the ambient noise recorded during 6 months by the large-N Maupasacq array in the Arzacq-Mauleon basins (Western Pyrenees, France) in the period band 2-9 s. Our study revealed strong coherent sources from several directions corresponding to the Galicia (N275°), the northern Atlantic (N340°) and the Mediterranean sea (N130°), in good agreement with the directions previously identified by Chevrot et al. (2007). This observation motivated the development of an iterative matched filtering method to separate and extract coherent wave fronts from the ambient noise field, which has been validated on a synthetic test case. We have shown that robust phase velocity maps can be obtained with the eikonal approach, after averaging the eikonal phase velocity maps obtained for the different incoming azimuths. The final phase velocity maps at short pe-

riod (2-3 s) are nicely correlated with surface geology. At longer period (>4 s), they reveal the deep architecture of the Arzacq and Mauleon basins.

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