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Evidence for radial anisotropy in the lower crust of the Apennines from Bayesian ambient noise tomography in Europe

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SUMMARY

Probing seismic anisotropy of the lithosphere provides valuable clues on the fabric of rocks. We present a 3-D probabilistic model of shear wave velocity and radial anisotropy of the crust and uppermost mantle of Europe, focusing on the mountain belts of the Alps and Apennines. The model is built from Love and Rayleigh dispersion curves in the period range 5 to 149 s. Data are extracted from seismic ambient noise recorded at 1521 broadband stations, including the AlpArray network. The dispersion curves are first combined in a linearised least squares inversion to obtain 2-D maps of group velocity at each period. Love and Rayleigh maps are then jointly inverted at depth for shear wave velocity and radial anisotropy using a Bayesian Monte-Carlo scheme that accounts for the trade-off between radial anisotropy and horizontal layering. The isotropic part of our model is consistent with previous studies. However, our anisotropy maps differ from previous large scale studies that suggested the presence of significant radial anisotropy everywhere in the European crust and shallow upper mantle. We observe instead that radial anisotropy is mostly

localized beneath the Apennines while most of the remaining European crust and shallow upper mantle is isotropic. We attribute this difference to trade-offs between radial anisotropy and thin (hectometric) layering in previous studies based on least-squares inversions and long period data (>30 s). In contrast, our approach involves a massive dataset of short period measurements and a Bayesian inversion that accounts for thin layering. The positive radial anisotropy ($V_{SH} > V_{SV}$) observed in the lower crust of the Apennines cannot result from thin layering. We rather attribute it to ductile horizontal flow in response to the recent and present-day extension in the region.

Key words: Seismic anisotropy; Seismic noise; Surface waves and free oscillations; Seismic tomography; Europe

1 INTRODUCTION

The convergence between the large Eurasian and African plates, which started in the early Cretaceous, is responsible for the creation of several orogens in the Mediterranean region such as the Alps, Carpathians, Apennines and Dinarides (e.g. Handy et al. 2010) (Figure 1). Several microplates were involved during the stages of subduction and collision. Among them, the Adriatic microplate (or Adria) played a key role in the current geodynamic structure of the Western and Central Mediterranean region and its collisional mountain belts. Adria is the upper plate of continental subduction in the Western and Central Alps, while it is the lower plate in the Apenninic and Dinaric subductions (e.g. Wortel & Spakman 2000; Lippitsch et al. 2003; Spakman & Wortel 2004). The question of which plate is subducting underneath the other in the Eastern Alps is still unresolved (e.g. Lippitsch et al. 2003; Koulakov et al. 2009; Mitterbauer et al. 2011; Zhao et al. 2016a). This complex tectonic setting is associated with high deformation in the crust (e.g. Nocquet & Calais 2004), switches in subduction polarity in the Alps-Apennines transition (Vignaroli et al. 2008; Zhao et al. 2016a) and Eastern Alps (Lippitsch et al. 2003) and complicated geometry of slabs in the mantle (e.g. Vignaroli et al. 2008; Faccenna et al. 2014). In addition, Adria plays an important part in the seismic hazard of the region as illustrated by the high concentration of earthquakes in the Apennines, Eastern

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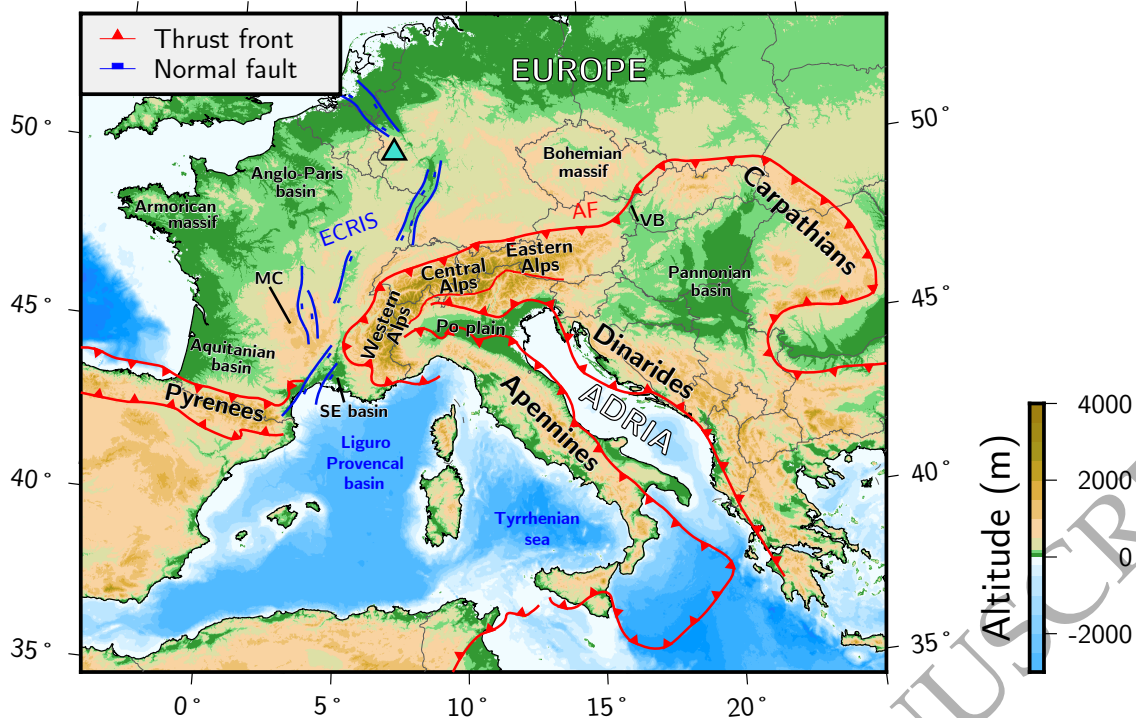


Figure 1. Topographic map of the study area and main geological features. MC: Massif Central; VB: Vienna basin; SE basin: Southeast basin; ECRIS: European Cenozoic rift system; AF: Alpine front. Turquoise triangle: Eifel hotspot.

Alps and Dinarides due to the convergence of Adria and Europe, which is still active in those areas.

In such a complex setting, seismic imaging is a tool of decisive importance to provide constraints on the lithospheric structure and fabric of crustal and mantle rocks that are necessary to build geodynamical models of the region. Teleseismic travel-time P-wave tomography of the mantle has allowed to map subducting slabs at large scale beneath Europe down to the transition zone and to observe slab detachments (Spakman et al. 1993; Lucente et al. 1999; Wortel & Spakman 2000; Piromallo & Morelli 2003). However, teleseismic travel-time tomography has a poor vertical resolution in the upper mantle because of steeply incident rays. Alternatively, long period surface wave tomography has been used to produce regional 3-D models of shear wave velocity beneath Europe (Boschi et al. 2004; Pasyanos 2005; Weidle & Maupin 2008; Schivardi & Morelli 2011). Nevertheless, all these tomographic models are based on earthquake sources, so their spatial resolution is limited by the ray coverage.

In order to increase resolution, Zhu et al. (2012, 2015) proposed an adjoint approach to simultaneously invert body and surface waveforms. One advantage of inverting full waveforms is to mitigate the issue of ray coverage, as waveforms contain scattering and a broader sensitivity

to structure (Fichtner et al. 2010; Lekić & Romanowicz 2011). They found many smaller-scale structures such as more detailed slab geometries, mantle upwellings and lithospheric delaminations. However, the resolution in full-waveform inversion is limited by the frequency content of inverted waveforms. As a result, the studies from Zhu et al. (2012, 2015) are focused on the mantle structure and do not provide detailed information in the crust.

Ambient noise tomography has proven to be a reliable technique to improve both data coverage and resolution at crustal depth and at continental scale in Europe (Stehly et al. 2009; Verbeke et al. 2012; Molinari et al. 2015; Kästle et al. 2016, 2018; Lu et al. 2018, 2020; Zhao et al. 2020). In the Alps and Apennines, Molinari et al. (2015) invert simultaneously Rayleigh wave group and phase velocities between 5 and 37 s to retrieve the distribution of 3-D shear velocity in the crust. Lu et al. (2018) invert Rayleigh group velocities in a broader period range (5-150 s) to reconstruct the 3-D shear velocity structure in the crust and the uppermost mantle beneath Europe. Kästle et al. (2018) combine ambient noise and earthquake data to map shear velocities down to 220 km depth.

These ambient noise tomographic models agree on the first-order isotropic structures and show a good agreement with both active and passive seismic studies. As opposed to teleseismic studies, their level of resolution (a few tens of km laterally and a few km vertically) allows to map crustal heterogeneities and shallow structures such as sedimentary basins. However, they do not account for seismic anisotropy. This is a clear limitation, as including azimuthal or radial anisotropy helps to map the fabric of crustal and mantle rocks, whose knowledge is of great interest in regions that have undergone high strain during their tectonic history.

Our knowledge of azimuthal anisotropy beneath the Alps and Italy mostly comes from SKS splitting studies. Fast velocity directions are generally parallel to the mountain chain in the Western and Central Alps (Barruol et al. 2004; Lucente et al. 2006; Barruol et al. 2011; Salimbeni et al. 2018), the Eastern Alps (Bokermann et al. 2013; Qorbani et al. 2015) and the Apennines (Palano 2015). A few studies have addressed azimuthal anisotropy in Europe from surface wave analysis but they are usually at local scale (e.g. Fry et al. 2010; Schippkus et al. 2020) or use periods that do not resolve crustal structure well (e.g. Nita et al. 2016). Radial anisotropy beneath Europe is mostly known from very large scale or global surface wave studies where the crust is not resolved (Kustowski et al. 2008; Weidle & Maupin 2008; Boschi et al. 2009; Zhu et al. 2015; Schivardi & Morelli 2011). All these studies are constructed from long period observations, where the crust is not inverted for but fixed from a reference model such as for example CRUST1.0 (Laske et al. 2013). However, although details within the crust cannot be resolved by long-period seismic waves, errors in the

reference crustal model may bias images of the upper mantle. Indeed, trade-offs exist between heterogeneities in the crust (e.g. Moho topography) and anisotropy in the mantle (Bozdağ & Trampert 2008; Panning et al. 2010; Chang & Ferreira 2017). In particular, apparent positive radial anisotropy ($V_{SH} > V_{SV}$) in long wavelength tomographic models may be induced by horizontal layering (e.g. Backus 1962; Bodin et al. 2015; Alder et al. 2017). This may be an important effect in Western Europe, as thin (hectometric) layering of the lower crust has been inferred from ubiquitous observations in deep seismic reflection profiles of the Variscan crust (Paul & Nicollin 1989; Singh & McKenzie 1993; Rey 1993). In the long period studies cited above, radial anisotropy is ubiquitous as it may improve data fit without the need for modelling isotropic small-scale structures. In this context, radial anisotropy cannot reliably be used to interpret models derived from long-period surface waves in terms of structural fabric, i.e. lattice preferred orientation (LPO) of anisotropic minerals, resulting from deformation. At shorter periods, Stehly et al. (2009); Kästle et al. (2018) measured both Rayleigh and Love dispersion but they did not invert them jointly for radial anisotropy at depth.

The aim of this study is to build the first 3-D model of radial anisotropy for the crust and uppermost mantle beneath Europe, in particular beneath the Alps and Apennines. We use a massive ambient noise surface wave dataset analysed in a broad frequency range (5 to 149 s). Taking advantage of the AlpArray temporary network (Hetényi et al. 2018) and surrounding permanent networks, we assemble a new large dataset of ambient noise Rayleigh and Love group velocities recorded at more than 700,000 station pairs. Two reasons led us to use group velocity instead of phase velocity. First, group velocities are easier to extract from noise correlation than phase velocities. Their estimations are therefore usually more robust and have proven to be a reliable choice in ambient noise tomography (e.g. Shapiro & Campillo 2004; Shapiro et al. 2005; Yang et al. 2007; Stehly et al. 2009; Lu et al. 2018). In addition, group velocities are sensitive to shallower structure than phase velocities for the same period range (e.g. Forsyth 1975). Our goal is to focus on the V_s structure of Europe from crust to uppermost mantle. Given the period range available, the use of fundamental-mode group velocities is then justified by the targeted depth range.

A standard 2-D linear least-squares tomography is first performed at each period on Rayleigh and Love dispersion data. In a second step, Rayleigh and Love dispersion data are jointly inverted at each geographical location for a 1-D depth profile of shear wave velocity and radial anisotropy. This 1-D inversion is done within the transdimensional hierarchical Bayesian framework of Bodin et al. (2012b, 2016); Yuan & Bodin (2018), where the trade-off between horizontal layering and radial anisotropy is accounted for. That is, the number of

layers in the inverted 1-D models is determined by the data themselves and anisotropy is added in each layer only if required by the data. Finally, all the 1-D models are assembled to create a regional 3-D probabilistic anisotropic model.

This study is focused on radial anisotropy. Such a goal requires an identical data coverage for Love and Rayleigh waves to perform a joint inversion of Love and Rayleigh group velocity maps with identical resolutions. Since Love waves observed on the correlations of horizontal components (transverse-transverse) have a lower signal-to-noise ratio than Rayleigh waves observed on vertical-vertical correlations, we have to reject a large part of the Rayleigh measurements that would have been used for imaging isotropic structures. We therefore emphasize that the goal of this study is not to improve the resolution of isotropic structures compared to previous high-resolution ambient noise models of the region (e.g. Lu et al. 2018; Kästle et al. 2018) but instead to image radial anisotropy across Europe in a way that is not biased by different data coverage of Rayleigh and Love waves.

The novelty of our approach is to combine a unique and massive dataset of Rayleigh and Love wave measurements over a broad period range with a Bayesian inversion. This Bayesian approach allows to sample the ensemble of possible anisotropic layered models, where each layer can be either isotropic or anisotropic. In this way, anisotropy is not imposed at the outset but is introduced only if required by the data. This data-driven procedure prevents us from getting an anisotropic signal everywhere in the model, as it is the case in previous studies in Europe. Indeed, we show that radial anisotropy is overall not pervasive in the crust and uppermost mantle. Radial anisotropy is only unambiguously present in specific areas, mainly in the lower crust of the Apennines, where it could be related to the recent and present-day extensive deformation of the Adriatic plate.

2 SURFACE WAVE DATASET

2.1 Group velocity measurements from ambient noise correlations

We use a dataset of group velocity dispersion curves in the period range 5-149 s for both Rayleigh and Love waves. The group velocity measurements are obtained from ambient noise correlation following an approach similar to the one described in Lu et al. (2018). We summarize here its main steps and refer to Lu et al. (2018) for a more detailed description.

Dispersion measurements are obtained from cross-correlations of ambient seismic noise recorded on horizontal and vertical components at 1521 broadband European stations. The locations of receivers, which belong to the AlpArray temporary experiment (Hetényi et al.

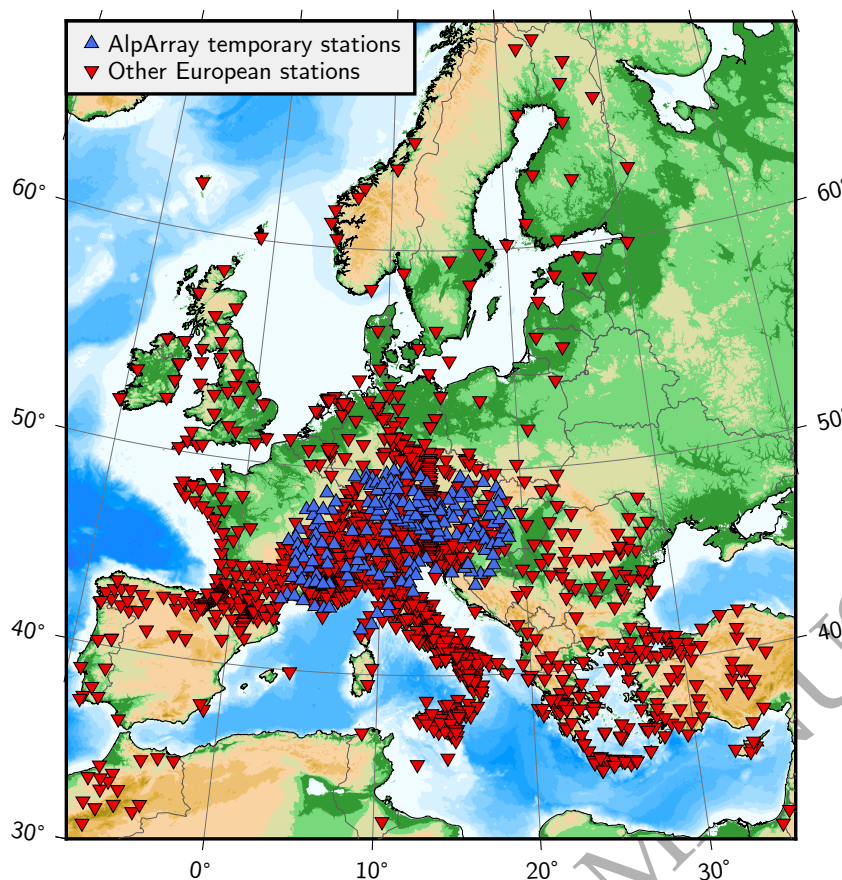


Figure 2. Map of the 1521 broadband stations used in this study. Blue triangles represent receivers belonging to the AlpArray temporary seismic network (Hetényi et al. 2018). Red triangles correspond to stations from other temporary and permanent networks in and around Europe.

2018) and other European permanent and temporary seismic networks, are shown in Figure 2. Continuous seismic noise signals recorded between 2010 and 2016 are correlated in the period range 30-149 s for all stations that were in operation in that time period. Noise recordings from 2016 to 2017 are correlated in the period range 5-100 s for all available permanent and temporary stations including AlpArray. As in Lu et al. (2018), cross correlations are computed by segments of 4 hours before being stacked over the whole record duration. Group velocity dispersion curves for the fundamental mode of Love and Rayleigh waves are derived from the stacked correlations using a time-frequency analysis (Dziewonski et al. 1969; Herrmann 1973; Levshin et al. 1989). Rayleigh and Love wave dispersion curves are respectively estimated from the ZZ (vertical-vertical) and TT (transverse-transverse) correlations.

2.2 Measurements selection

At each period, we only select the best quality Rayleigh and Love group velocity measurements. The selection follows classical steps used in ambient noise tomography (e.g. Lu et al. 2018). They are based on signal-to-noise ratio, symmetry of the stacked correlations to minimize the error on group velocity estimations due to the directionality of noise sources (e.g. Stehly et al. 2006) -, inter-station distances to prevent the positive and negative times of the correlation from overlapping (e.g. Stehly et al. 2009) and the range of group velocity expected in the study region.

For each station pair, only measurements fulfilling the selection criteria for both Rayleigh and Love waves are kept in the final dataset. This ensures the same coverage for both types of surface waves at each period, thus resulting in an identical lateral resolution in the 2-D tomographic maps. By this way, we avoid differences that may introduce artificial radial anisotropy in our final 3-D model. Depending on period, we select only 2 % to 19 % of the initial dispersion dataset (Figure 3). A high level of rejection is commonly observed in ambient noise studies (Stehly et al. 2009; Lu et al. 2018). In our case, the lower signal-to-noise ratio on the horizontal components further increases this high level of rejection by excluding Love measurements from the selection. Nevertheless, we keep more than 20,000 station pairs for both Rayleigh and Love waves between 10 and 40 s, with more than 100,000 station pairs around 20 s period. The selection is designed to investigate radial anisotropy. It ensures identical data coverage for Love and Rayleigh waves but removes a large quantity of useful Rayleigh measurements that could have been used to derive a more detailed V_{SV} model. As a comparison, Lu et al. (2018) use a maximum of 230,000 Rayleigh measurements at 15 s to build their isotropic V_{SV} model of Europe from ambient noise, corresponding to 30 % of their initial dataset.

3 2-D SURFACE WAVE TOMOGRAPHY

3.1 Inversion method

Our dataset corresponds to the selected Rayleigh and Love waves group velocity measurements between station pairs in the period range 5 s to 149 s. Those group velocities are combined in a tomographic inversion to build 2-D isotropic velocity maps for Rayleigh and Love waves, the maps being generated independently for each period and each type of waves (Rayleigh or Love). The procedure follows the one described in Debayle & Sambridge (2004) and is a

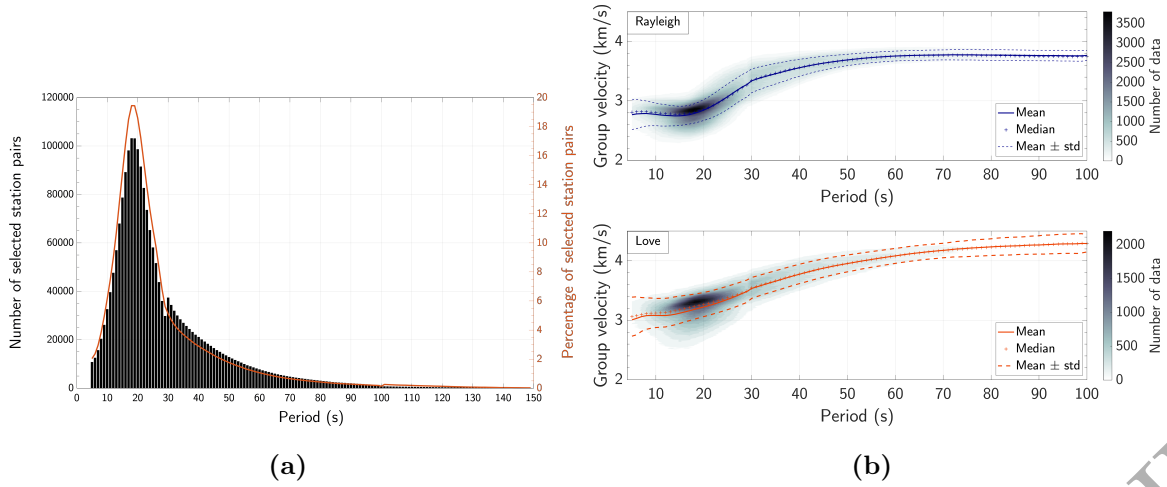


Figure 3. a) Number of selected station pairs according to period (left axis) and corresponding proportion of the initial dataset (right axis). The selection is performed jointly for Rayleigh and Love group velocity measurements. b) Ensemble of selected Rayleigh and Love dispersion measurements. The number of data is colour coded. The mean and median values are represented as well as one standard deviation from the mean.

classical linearised inversion within the framework of ray theory. Here again, we summarize the method but the reader is referred to the original paper for more details and discussion.

The continuous regionalization approach of Debayle & Sambridge (2004) uses the least squares solution for linearised inverse problems originally described in Tarantola & Valette (1982). Within this framework, the problem is linearised around a reference homogeneous velocity model \mathbf{m}_0 which is defined in our case as the mean of the observed group velocity measurements at a given period. The regularization is entirely controlled by the choice of C_m , the *a priori* model covariance matrix. C_m is defined by two parameters: σ_m , the model *a priori* standard deviation, and L_{corr} , the horizontal correlation length between two points of the model. σ_m is the damping parameter which controls the maximum amplitude of velocity perturbations allowed in the final 2-D group velocity model with respect to the reference model \mathbf{m}_0 . It is set to 0.05 km/s. This value allows to recover large velocity variations in our model, reaching up to 20% at 8 s period, as can be seen in Figure 4.

The choice of L_{corr} defines the lateral degree of smoothing in the model. That choice depends on the ray coverage and on the wavelength. After several trials, and taking advantage of the dense coverage available in this study (Figure A.1 in Appendix A), we choose to decrease L_{corr} with decreasing period according to Table 1. Using larger values of L_{corr} would provide smoother maps leading to a poorer fit to the data but would not change the overall result of this study.

For each wave type, a diagonal matrix C_d is defined at each period. Its diagonal terms are σ_{di}^2 , with σ_{di} being the standard deviation of measurement i , i.e. the data error. This error σ_{di} is defined at each period as the average of the absolute value of the velocity difference between all causal (positive time of the correlations) and acausal (negative time of the correlation) measurements. Since σ_{di} controls the relative contribution of station pair i to the final model, all station pairs at a given period contribute equally to the model. This is a reasonable choice, as we cannot estimate robust error on the group velocity for each individual station pair.

Within this standard regularized linear inversion scheme, the estimated level of data error and the level of posterior uncertainties are linked to the arbitrarily chosen level of damping (Aster et al. 2018). For this reason, we do not discuss any further errors on the data. In the same way, we do not discuss the uncertainties associated with the 2-D group velocity maps as they are biased by the regularization (see also section 4.4).

The 2-D inverted group velocity model is discretised on a regular grid ($0.15^\circ \times 0.15^\circ$), smaller than the horizontal smoothing imposed on the model, which can be approximated as $\sim 3L_{corr}$ (Debayle et al. 2016). This parameterization allows to retrieve lateral heterogeneities of the size of a few tens of kilometers, especially at short periods, while keeping a reasonable computational cost of few hours maximum per period.

The group velocity maps obtained via our regionalization approach result from an inversion that does not account for Rayleigh and Love wave azimuthal anisotropies. The azimuthal variation for Rayleigh waves depends mostly on the sine and cosine of 2ψ , where ψ is the azimuth, and can readily be distinguished from the isotropic component. However, the azimuthal variation of Love waves shows a stronger dependence on 4ψ terms, which need a much better path coverage to be resolved (Smith & Dahlen 1973). The counterpart is that this faster azimuthal variation should be averaged out more easily by the propagation paths in an isotropic inversion. This is important, because if the propagation paths are sufficient to average over the azimuthal dependency, the inverted velocities in an isotropic tomography should be representative of the average true velocities. If a bias exists, it should be greater for Rayleigh waves because their slower 2ψ variation (with a π -periodicity) is more difficult to average out than the faster 4ψ variation (with a $\pi/2$ periodicity) of Love waves (e.g. Debayle & Kennett 2000; Sieminski et al. 2003; Debayle & Ricard 2012). We carried out Rayleigh wave inversion with and without azimuthal anisotropy and we found that both inversions yielded very similar isotropic parts. We concluded that our ray coverage is sufficient to average out properly the azimuthal variations of both Rayleigh and Love waves. We therefore neglect azimuthal anisotropy and only perform isotropic 2-D inversions.

Period T (s)	$T \leq 20$	$20 < T \leq 50$	$50 < T \leq 100$	$100 < T$
L_{corr} (km)	20	25	50	100

Table 1. Correlation lengths used in the tomographic inversion depending on period. This parameter controls the smoothing of the final model.

To assess the parts of the model that are well resolved given the ray coverage, parameterization and the regularization used, we carried out a number of synthetic tests shown in Appendix B.

3.2 Group velocity maps

Group velocity maps for Love and Rayleigh waves are shown at 8 s, 30 s and 55 s periods in Figure 4. The corresponding ray densities are shown in Figure A.1 in Appendix A. Additional group velocity maps are shown in Figure C.1 in Appendix C for periods 60, 80 and 110 s. Here we only discuss results for periods 8, 30 and 55 s since some smearing affects a large part of the study area at longer periods.

At 8 s, both Love and Rayleigh waves are sensitive to the upper crust. In agreement with previous studies conducted from ambient noise in the area (e.g. Stehly et al. 2009; Molinari et al. 2015; Kästle et al. 2018; Lu et al. 2018, 2020), the low velocity zones in our model highlight the main sedimentary basins such as the Southeast basin in France, the Alpine molasse basin, the Po plain and to a lesser extent the Vienna, Pannonian and Aquitanian basins that are close to the boundaries of the study area. High velocity anomalies are in contrast associated with mountain belts such as the Alps, Apennines, Dinarides or Pyrenees. Variscan massifs are also associated with high velocities: Armorican Massif and Massif Central in France, and even more clearly the Bohemian Massif which is clearly visible on both Rayleigh and Love maps.

At 30 s, surface waves are mostly sensitive to the lower crust and the uppermost mantle. Low velocity anomalies observed in the Alps, Apennines and Dinarides are therefore attributed to the crustal roots of these orogens. Interestingly, the Love wave low velocity anomaly is stronger beneath the Apennines than beneath the Alps, while these differences are not observed for Rayleigh waves. A possible explanation is based on the studies of Chiarabba & Amato (1996); Chiarabba et al. (2009); Di Stefano et al. (2009) who conclude that very high temperatures, and so very low seismic velocities, can be observed in the lower crust of the Apennines. The lower crust of the Apennines would therefore be associated with a stronger velocity reduction than the lower crust of the Alps. With this in mind, we can argue that

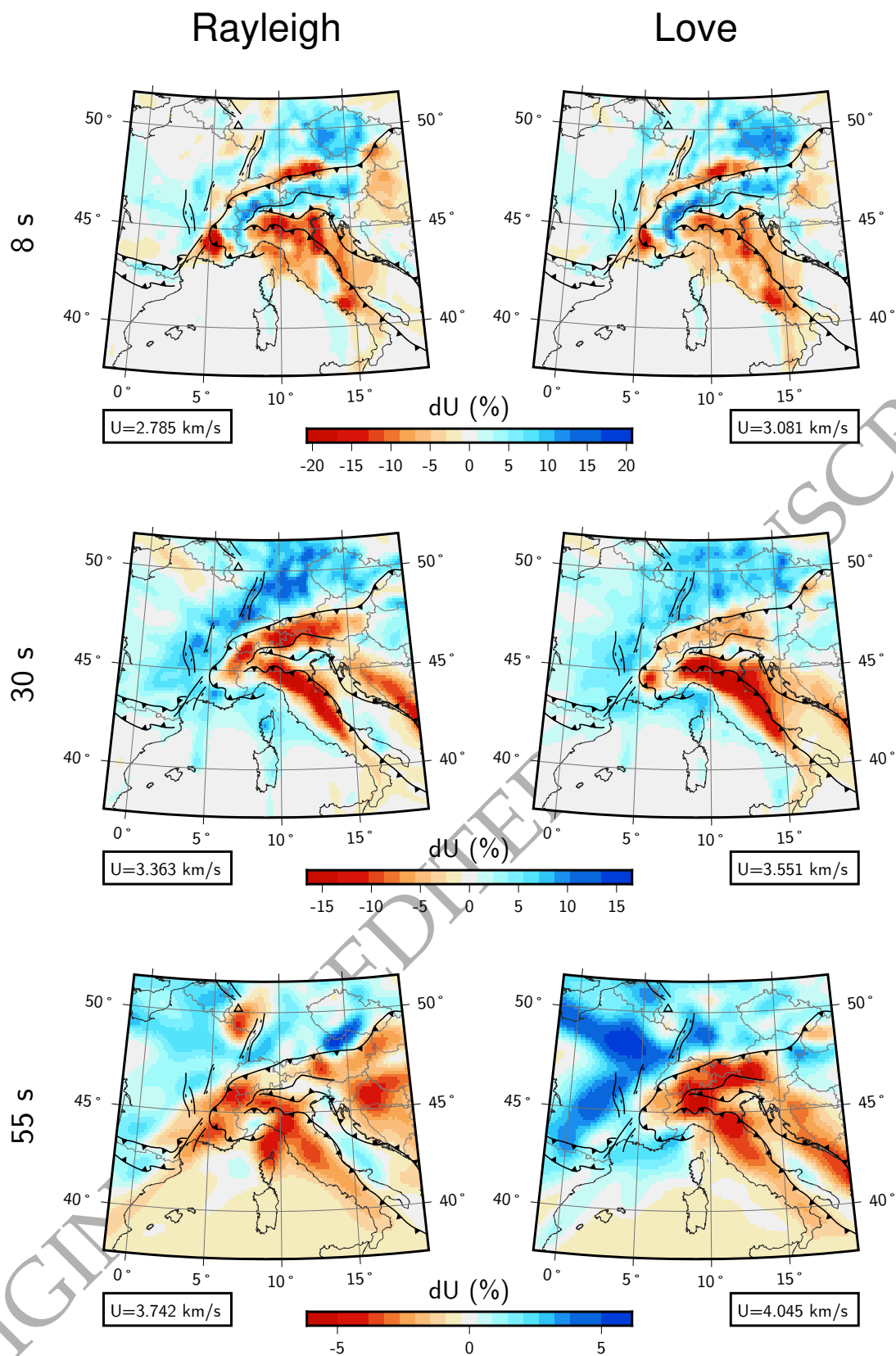


Figure 4. Group velocity perturbation maps for Rayleigh and Love waves at periods 8, 30 and 55 s. The reference group velocity U is indicated below each map. Major sedimentary basins, mountain belts, and Variscan massifs are clearly visible.

Love waves at 30 s period are sensitive to the hot and very low-velocity lower crust above the shallower Moho beneath the Apennines and to mid-crustal velocities above the deeper Moho beneath the Alps (e.g. Spada et al. 2013) thus leading to a stronger low velocity anomaly in the Apennines. However, due to their slightly deeper sensitivity at the same period, Rayleigh waves would pick up the very low velocity above the Apenninic Moho but also the higher mantle velocities at slightly greater depths, resulting in an average low velocity anomaly similar in strength to the ones observed above the deeper Moho of the Alps where the waves are primarily sensitive to the lower crust.

The pattern of velocity anomalies observed at 55 s suffers from some smearing of the high velocity anomalies, mostly in France (see also ray density maps in Figure A.1). Several features are nonetheless clearly observable. A strong Rayleigh wave low-velocity anomaly and a weaker low-velocity Love wave anomaly roughly coincide with the Eifel volcanic region in Western Germany. In this region, the lithosphere-asthenosphere boundary (LAB) is approximately at 40-50 km depth (e.g. Mathar et al. 2006; Seiberlich et al. 2013). The relative shallowness of the LAB is suggested to be linked to the presence of a thermal anomaly that spreads at the bottom of the lithosphere (e.g. Ritter et al. 2001; Keyser et al. 2002; Pilidou et al. 2005), affecting to a lesser extent the thermal state of the lithosphere itself (Schintgen et al. 2015). The difference in velocity anomalies seen between Rayleigh and Love waves maps may be due to the deeper penetration of Rayleigh waves, sampling at this period the top of the low velocity asthenosphere, while Love waves would mostly be sensitive to the colder lithosphere, which is still slightly hotter than its surroundings. The differences between the Love and Rayleigh patterns beneath the Alps, Apennines and Dinarides clearly suggest the presence of strong velocity and radial anisotropy variations in the lower crust and uppermost mantle.

4 BAYESIAN INVERSION FOR VELOCITY AND ANISOTROPY AT DEPTH

In this section we describe how we invert group velocity maps for a 3-D model of shear wave velocity and radial anisotropy. At each geographical location, we use both Love and Rayleigh group velocities from 5 s to 149 s with a step of 5 s, and perform a 1-D inversion within a transdimensional Bayesian framework (Bodin et al. 2012a,b; Yuan & Bodin 2018).

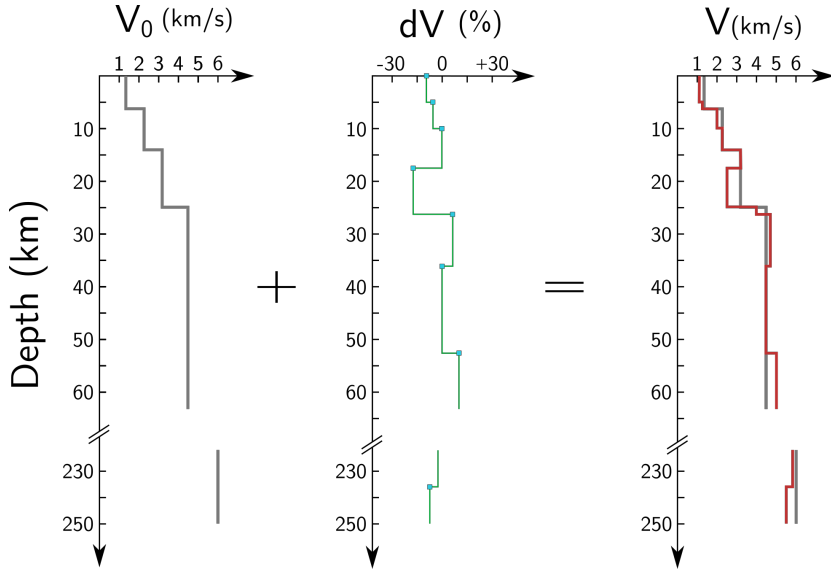


Figure 5. Illustration of the model parameterization: (left) layered reference model $V_0(z)$ that remains fixed during the inversion; (middle) model of velocity variations $dV(z)$ defined by layers whose number, thickness, and velocity change dV are unknowns of the inversion; (right) final model constructed by adding the perturbations to the background model.

4.1 Model parameterization

Since Love waves are primarily sensitive to V_{SH} and to a lesser extent to V_{SV} , and Rayleigh waves are primarily sensitive to V_{SV} , and to a lesser extent to V_P , we invert for the V_{SV} structure, the V_P/V_{SV} ratio, and radial anisotropy defined by the ratio V_{SH}/V_{SV} .

Because sharp discontinuities cannot be resolved by surface waves alone, we impose the depth of discontinuities from an isotropic reference velocity structure based on the V_{SV} model of Lu et al. (2018), which is defined at each geographical point as a stack of layers and depicts Moho topography as well as isotropic smooth lateral V_{SV} variations for the crust and uppermost mantle, down to 80 km depth. At greater depths, Lu et al. (2018)'s model is combined with PREM (Dziewoński & Anderson 1981) down to 670 km depth. As described in Yuan & Bodin (2018), we invert for perturbations around this reference model, which remains unchanged during the procedure. The velocity $V_{SV}(z)$ at depth z is then a combination of the reference model V_0 and the inverted model dV :

$$V_{SV}(z) = V_0(z)(1 + dV(z)) \quad (1)$$

Figure 5 illustrates the different elements of this parameterization in a schematic way.

The inverted model $dV(z)$ is parameterized as a stack of layers, where the thickness and the V_{SV} perturbation in each layer are unknown parameters. Note that these perturbations

do not need to be small, as there is no linearisation involved in the forward calculation of dispersion curves. The ratios V_{SH}/V_{SV} and V_P/V_{SV} in each layer of the inverted model are also free parameters. The density ρ is simply scaled on V_P following an empirical Birch's law (Birch 1961) for the lithosphere in the form $\rho = 2.35 + 0.036 (V_P - 3)^2$ as in Tkalčić et al. (2006). To account for the sensitivity of the longest-period surface waves in our dataset to upper-mantle shear velocity, the inverted model is parameterized from the surface to 250 km depth. However, we only show results from our final 3-D model down to maximum 120 km depth, as we estimate the resolution to be too low at greater depths.

4.2 A transdimensional approach

As mentioned in the introduction, a strong trade-off exists between radial anisotropy, created by LPO of anisotropic minerals, and vertical heterogeneities, i.e. horizontal layering (see e.g. Backus 1962; Bodin et al. 2015; Alder et al. 2017). Our inverse problem is then highly non-unique since a model with many isotropic layers explains the data as well as a model with fewer anisotropic layers. To address this non-uniqueness, we place ourselves in a transdimensional Bayesian framework where the number of model parameters is unknown. This means that the total number of layers in the inverted model as well as the presence of anisotropy in each layer are not fixed parameters but they are adjusted by the inversion to fit the data to the degree required by their estimated noise. Each layer in the inverted model can be either isotropic and described solely by its shear wave velocity perturbation dV , and V_P/V_S ratio (in this case, V_{SH}/V_{SV} is fixed to unity), or radially anisotropic and described by three parameters: dV , V_P/V_S , and V_{SH}/V_{SV} .

4.3 Bayesian inference

Within a Bayesian framework, each piece of information is treated in a probabilistic way. The solution of our inverse problem is a large ensemble of layered models distributed according to the posterior probability distribution $p(\mathbf{m}|\mathbf{d}_{obs})$, which represents the probability density of model \mathbf{m} given the observed dispersion curves \mathbf{d}_{obs} (and given the reference model V_0). According to Bayes' theorem,

$$p(\mathbf{m}|\mathbf{d}_{obs}) \propto p(\mathbf{d}_{obs}|\mathbf{m}) \times p(\mathbf{m}) \quad (2)$$

$$\text{posterior} \propto \text{likelihood} \times \text{prior}$$

the posterior probability distribution $p(\mathbf{m}|\mathbf{d}_{obs})$ is a function of the prior $p(\mathbf{m})$ and the likelihood function $p(\mathbf{d}_{obs}|\mathbf{m})$. The prior represents the state of knowledge of the model before taking into account the data, whereas the likelihood function is the probability of observing the measured dispersion curves given a particular model.

In the context of a Bayesian inversion, the level of data noise (both observational and theoretical errors) is an important parameter as it controls the required level of fit to the observed dispersion curves and hence the required level of complexity in the solution (e.g. Sambridge et al. 2013). This is achieved through the form of the likelihood function $p(\mathbf{d}_{obs}|\mathbf{m})$ which is based on a statistical model of data errors. Here, we assume a Gaussian distribution for data errors, and hence the likelihood is based on a least squares misfit function in the form

$$p(\mathbf{d}_{obs}|\mathbf{m}) = \frac{1}{(\sqrt{2\pi})^{2n} \times (\sigma_R \sigma_L)^n} \times \exp \left\{ -\frac{1}{2} \left(\frac{\|\mathbf{U}_R - \mathbf{u}_R(\mathbf{m})\|^2}{\sigma_R^2} + \frac{\|\mathbf{U}_L - \mathbf{u}_L(\mathbf{m})\|^2}{\sigma_L^2} \right) \right\} \quad (3)$$

where σ_R^2 and σ_L^2 respectively refer to the variance of the estimated errors for Rayleigh and Love data, n is the number of points in the dispersion curves (which is the same for Love and Rayleigh waves), \mathbf{U}_R and \mathbf{U}_L are the observed dispersion curves for Rayleigh and Love waves respectively, and $\mathbf{u}_R(\mathbf{m})$ and $\mathbf{u}_L(\mathbf{m})$ are Rayleigh and Love dispersion curves estimated for model \mathbf{m} (Bodin et al. 2016). Regarding the forward modelling step, Love and Rayleigh dispersion curves are computed with a normal mode formalism in a spherical Earth (Smith & Dahlen 1973). The estimation of fundamental mode group velocities is made in a fully non-linear fashion with a Runge-Kutta matrix integration (Saito 1967; Takeuchi & Saito 1972; Saito 1988).

Here we use a transdimensional framework, where the number of parameters is unknown. Therefore, one might think that adding parameters in the model can help fitting the data at the level required by their quality. However, transdimensional Bayesian inversions are parsimonious, which means that, for a given level of required data fit, simpler models with fewer parameters are preferred over complex ones (e.g. Sambridge et al. 2013). In our case, models with fewer and isotropic layers will then be preferred. Any anisotropy recovered by a model is therefore only that required by the data.

4.4 Hierarchical Bayes

The estimation of data errors is a well-known issue as both theoretical and observational errors are often difficult to estimate. This is particularly crucial here as underestimating noise would introduce more structure in the solution model to fit the data at a non-necessary level, whereas overestimating errors would lead to smooth models that miss resolvable structure.

Layer thickness (km)	dV (%)	V_{SH}/V_{SV}	$\frac{V_P}{V_{SV}}$	σ_R and σ_L (%)	Nb. of layers
2 - + ∞	± 30	0.8 - 1.2	1.6 - 1.9	0.2 - 3	3 - 300

Table 2. A priori information used in the Bayesian inversion.

Here, the data are the group velocities at each geographical point of the tomographic maps obtained after the first inversion in 2-D. The estimation of errors on these maps is computationally too expensive for a tomographic problem of this size (Debayle & Sambridge 2004). In addition, uncertainties on the maps are bounded by the arbitrary choice of regularization and data errors estimated before the 2-D inversion and may not be seen as an objective estimation of uncertainties (Tarantola & Valette 1982). This is because regularization or damping helps to stabilize the inversion of linear systems of equations and suppresses propagation of data noise into the solution, but this is at the cost of biasing the solution in a statistical sense (Aster et al. 2018). In this way, uncertainties in the 2-D group velocity maps cannot be propagated and used as 'data errors' in the 1-D depth inversion.

To overcome this lack of knowledge of data noise, we extend the Bayesian formulation of the problem and use a hierarchical Bayes model (Malinverno & Briggs 2004; Bodin et al. 2012a) where the data errors σ_R and σ_L are not fixed but are treated as unknown parameters to be inverted for. Although we ensured that Rayleigh and Love group velocity maps are produced with the same ray coverage, Rayleigh and Love measurements are made on different components and contain different levels of error. For this reason, we use two separate unknown variables σ_R and σ_L for the level of errors associated with each wave type.

4.5 Prior distribution

For each of the parameters involved in our inversion, the prior knowledge is represented by a uniform *a priori* distribution. The bounds of the distributions are summarized in Table 2. For example, the prior on the vector of velocity perturbations \mathbf{dV} in a model with k independent layers is, in this case

$$p(\mathbf{dV}|k) = \left(\frac{1}{dV_{\max} - dV_{\min}} \right)^k \quad (4)$$

with dV_{\min} and dV_{\max} respectively the lower and upper bounds of the prior. The uniform distributions of the total number of layers and of the number of anisotropic layers are taken as broad as possible, typically from 3 layers to several hundreds. These choices of the prior knowledge allow to explore a broad range of model variations, with the idea to allow large velocity amplitude variations and to explore trade-offs between parameters.

4.6 Sampling the posterior distribution

To deal with the variable dimension of the model space imposed by the transdimensional framework, we sample the posterior probability distribution using a reversible jump Markov chain Monte Carlo (rj-McMC) algorithm (Green 1995). Starting from the isotropic reference model, the method consists of an iterative random walk in the model space. At each iteration of the chain, a specific move from the current model $dV(z)$ is selected at random and with equal probability from the following possibilities:

(1) select a random layer, and perturb the value of V_{SV} , (2) select a random layer, and perturb the value of V_P/V_{SV} , (3) select a random anisotropic layer, and perturb the value of V_{SH}/V_{SV} , (4) modify the thickness of a random layer, (5) add or remove an isotropic layer at a random depth, (6) add or remove anisotropy in an existing random layer, (7) randomly modify the level of data noise. These perturbations only affect the inverted layered model and not the reference model which remains fixed during the whole procedure (see Section 4.1). At each iteration, a new model is proposed that only depends on the previous one. The new model is either accepted or rejected in the ensemble solution, depending on acceptance criteria based on the comparison of the likelihoods of the current and proposed models. The ensemble of accepted models asymptotically converges towards the posterior distribution (for details, see Bodin et al. 2012a).

Finally, the ensemble solution at each point is a probabilistic distribution of 70,000 layered models of V_{SV} , radial anisotropy and V_P/V_{SV} as a function of depth. We will see in Section 4.7 that although the inversion for V_P/V_{SV} is useful to avoid introducing biases on other parameters, this parameter is poorly constrained due to the weak and null sensitivity to V_P of Rayleigh and Love waves respectively. We therefore do not interpret the 3-D variations in V_P/V_{SV} and focus on V_{SV} and radial anisotropy. 1-D profiles of V_{SV} and radial anisotropy that we consider representative of the ensemble solution are then extracted from the posterior distributions. They are finally assembled in a single 3-D model of the crust and uppermost mantle of Europe.

4.7 Synthetic tests

To discuss the resolving power of our method, we carry out several tests by jointly inverting Rayleigh and Love synthetic dispersion curves to which a white Gaussian noise of 0.3% has been added. For each test, whose results are shown in Figures 6, 7 and 8, the solution is a large ensemble of layered models whose distribution is displayed to 120 km depth. In all cases, the median of each posterior distribution, i.e. the median over the large ensemble of solution

models, becomes a smooth function of depth because the depth of discontinuities in individual models is variable. Only the discontinuities present in the reference model are preserved in the median model.

It is important to note that at a given depth, each sampled model can either be isotropic or anisotropic. In this way, the distribution of the V_{SH}/V_{SV} parameter contains a large number of values equal to unity, which explains why the median of the anisotropic parameter is often exactly equal to unity. We therefore consider a median value different from unity as a reliable indication for the presence of anisotropy.

Panels a), c), e) in Figure 6 show that the general V_{SV} structure is relatively well retrieved below Moho depth, not only in its tendency but also in its absolute value. However, at depths shallower than 30 km on panels a) and c), we are only able to capture the general velocity increase instead of the sharp variations imposed in the true models. The Moho in the ensemble solution is imposed by the reference model, as implied by our particular method.

As expected, radial anisotropy is less constrained than V_{SV} : its posterior distribution is quite wide. In the case where the true model is isotropic (panel a) in Figure 6, the median value for V_{SH}/V_{SV} is strictly equal to unity, which is the true value. As shown in Figure 7, this would not be the case if anisotropy was imposed at all depths as an unknown parameter to be inverted for. Spatial variations of anisotropy are well retrieved (panels c) and e) in Figure 6), but the absolute level of anisotropy, both in positive (>1) and negative (<1) values, is always underestimated by the median of the distribution.

As surface waves are not sensitive enough to V_P , the V_P/V_{SV} ratio is not constrained at all and we do not interpret it. Of course, fixing V_P/V_{SV} in the inversion to the correct value would allow a better recovery of the amplitude of radial anisotropy. However, we do not know the actual 1-D profile of V_P/V_{SV} , and there is a risk to bias the results if we fix the V_P/V_{SV} ratio to a wrong value. In the context of a Bayesian formalism, it is therefore more sensible to use a wide *a priori* distribution for V_P/V_{SV} and to treat it as an unknown parameter, although we do not interpret its wide and poorly informative *a posteriori* distribution. Since V_P and density are scaled on one another, we could have inverted for ρ instead of V_P/V_{SV} . However, the density would not have been better constrained because surface waves have no sensitivity to this parameter.

From the tests shown in Figure 6, we conclude that the best resolved parameters are V_{SV} and radial anisotropy. We cannot resolve sharp discontinuities in these parameters but we recover well their variations, although radial anisotropy amplitudes are somewhat underestimated.

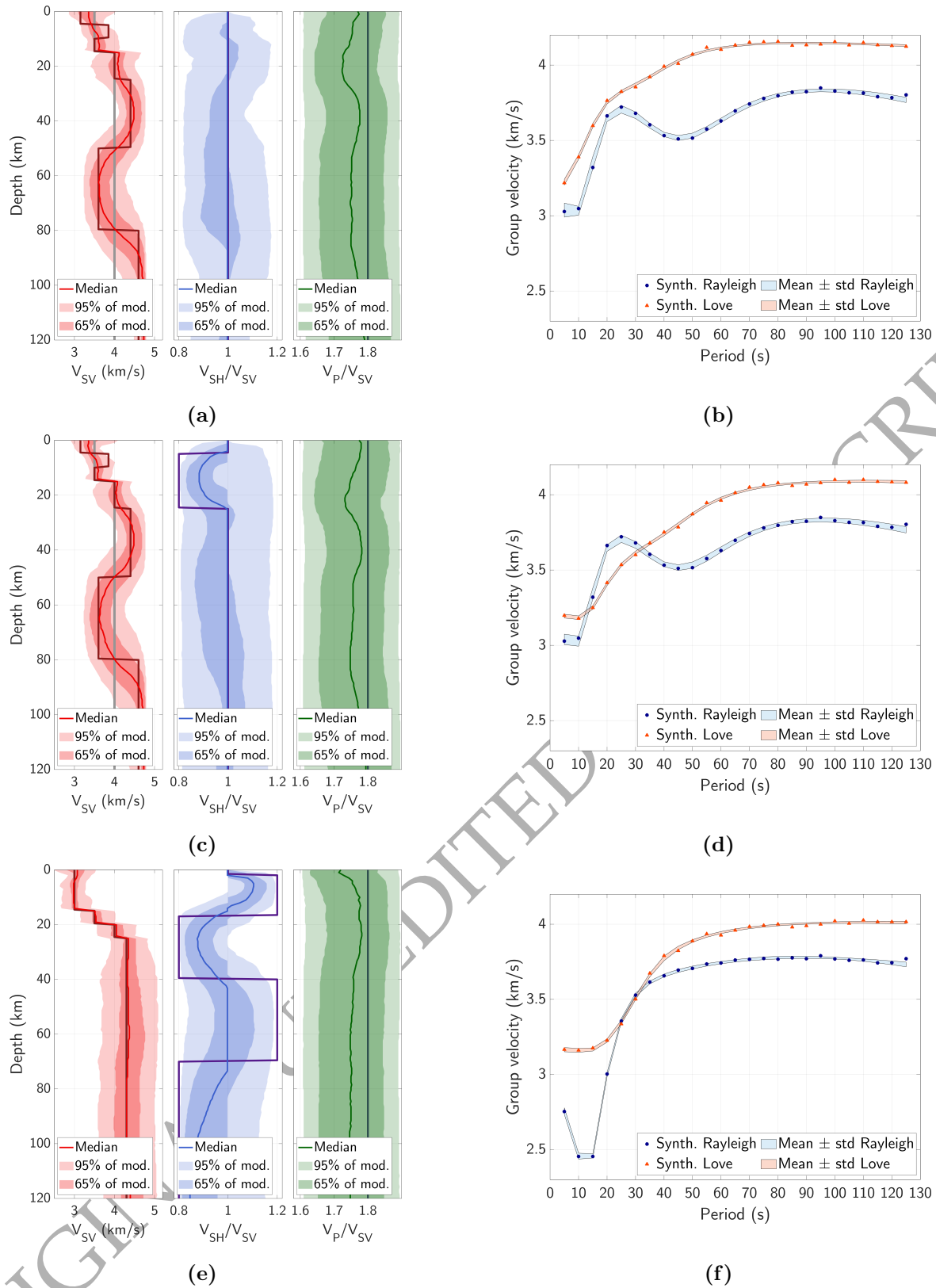


Figure 6. Synthetic tests. Left panels: posterior distributions of V_{SV} (red), V_{SH}/V_{SV} (blue) and V_P/V_{SV} (green). For each case (a, c, e), the true model is the thick solid line in every panel and the reference velocity model used in the inversion is the grey line in V_{SV} panel. Posterior distributions are depicted with their median (thin solid line) and likelihood intervals: for each parameter, the dark surface includes 65% of the models in the ensemble solution while the light area includes 95% of the models. Right: synthetic data (coloured dots) calculated from the true model and interval of mean \pm standard deviation of group velocity data calculated from the inverted models (coloured area).

In addition, while the ensemble solution is broadly spread in the model space, the associated data distribution in Figure 6 is relatively narrow and close to synthetic data. This clearly shows the non-uniqueness of the inverse problem given the flexible parameterization used here.

In Figure 7, we illustrate the benefit of this flexible parameterization where each layer can be either isotropic or anisotropic. A true model is designed with an anisotropic layer in the crust and an isotropic upper crust and mantle. Noisy synthetic data are created and inverted with two procedures. We compare our procedure (left panels) with a more standard approach where anisotropy is imposed and inverted at all depths (right panels). Everything else is equal in the two inversions, e.g. the number of layers is variable in both cases. Although V_{SV} is equally well resolved in both cases, the anisotropic layer in the crust is better recovered with our flexible scheme. In the mantle, where the true model is isotropic, the ensemble solution produced by our scheme includes a large number of isotropic models, resulting in a narrower distribution for V_{SH}/V_{SV} and a median value equal to unity. Conversely, the more standard approach documented in the right panels leads to a wider distribution of anisotropy and gives a median $V_{SH}/V_{SV} \approx 0.95$. This comparison shows that in the case of a true isotropic model, our flexible scheme fits data with simpler isotropic models described by fewer parameters, resulting in a distribution closer to the true model.

In Figure 8, we test our scheme in the case of isotropic models including a large number of horizontal interfaces, where the layer thickness in the true model is much smaller than what can be resolved by surface waves at these periods. The goal here is to test whether thin horizontal layering is going to be mapped into radial anisotropy. Layering in the continental lower crust of the Variscan provinces of Western Europe is known to exist from deep seismic reflection profiles, and it is interpreted as resulting from intrusions of mafic/ultramafic rocks derived from the upper mantle alternating with felsic granulitic lower crustal rocks (Singh & McKenzie 1993). Here, we design a model with a 30-km thick crust where the lower half of the crust (i.e. between 15 km and 30 km depth) is composed of a stack of 150 lamellae of 100 m thickness with V_{SV} velocities alternating between 3.33 km/s and 4.07 km/s, i.e. $\pm 10\%$ around 3.7 km/s, resulting in a peak-to-peak level of heterogeneity of 20%. This corresponds to the model proposed by Paul & Nicolin (1989) to explain the frequency dependent attenuation of the $P_M P$ phase observed in wide-angle data collected along the North of France deep seismic reflection ECORS profile. Left panels in Figure 8 show that in this case, no artificial anisotropy is recovered in the inverted model, as a majority of sampled profiles can explain both Love and Rayleigh waves data without the need for anisotropy. Similar results are observed for velocity

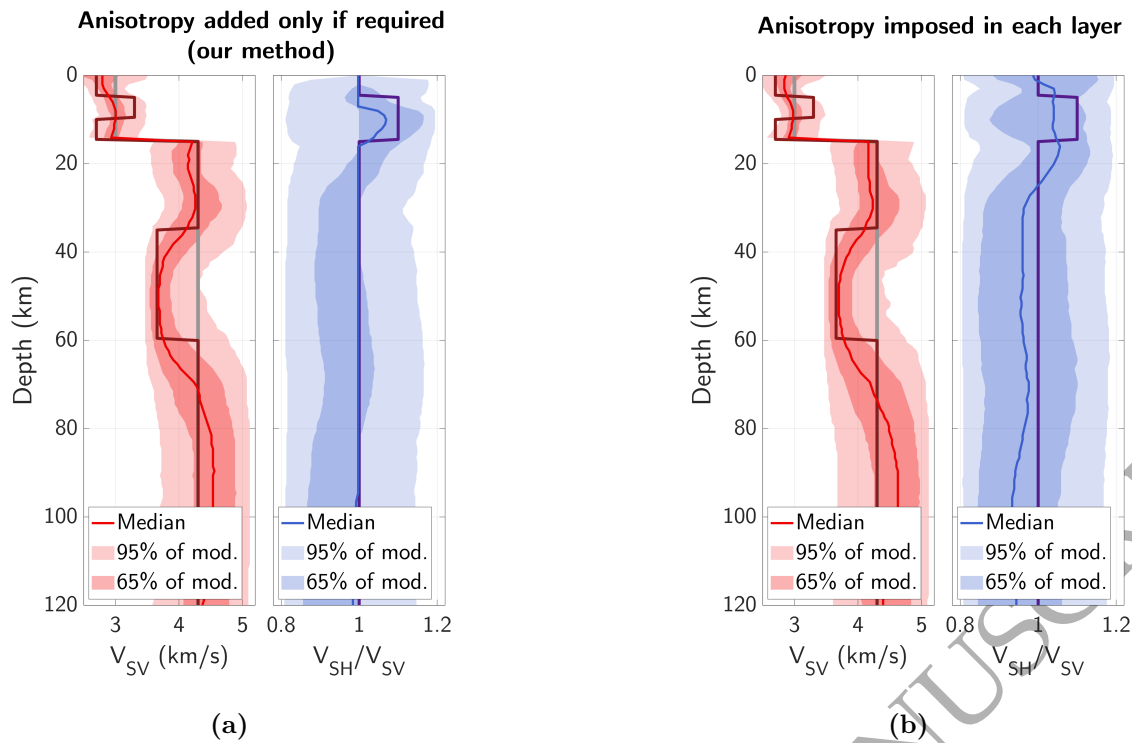


Figure 7. Synthetic tests where the same data produced by the true model are inverted with two different procedures. Left panels: inversion used in this study, where each layer can either be isotropic or anisotropic. Right panels: inversion where anisotropy is imposed as an unknown parameters at all depths. See legend of Figure 6 for the meaning of the different lines and colours.

contrasts reaching up to 30%. The second model tested in the right panels of Figure 8 shows a stronger level of heterogeneity with V_{SV} alternating between 2.96 km/s and 4.44 km/s, i.e. $\pm 20\%$ around 3.7 km/s, resulting in a peak-to-peak level of heterogeneity of 40%. This model is much less realistic as heterogeneities are stronger than what can be expected in the lower crust (Paul & Nicollin 1989; Singh & McKenzie 1993). In this case, the algorithm adds radial anisotropy to fit data, resulting in a 10-km thick anisotropic layer at the base of the crust in the median model. From these tests, we conclude that thin hectometric layering with up to 30% velocity contrasts will not map into anisotropy in our median posterior model, even though a large number of individual models in the ensemble solution depict anisotropy. This will be useful to interpret our 3-D anisotropic model of Europe.

5 RADIALLY ANISOTROPIC 3-D MODEL

We apply the 1-D inversion presented above at each geographical location and obtain a 3-D probabilistic model of the study area. Figure 9 and Figure 10 present 4 depth slices at 8,

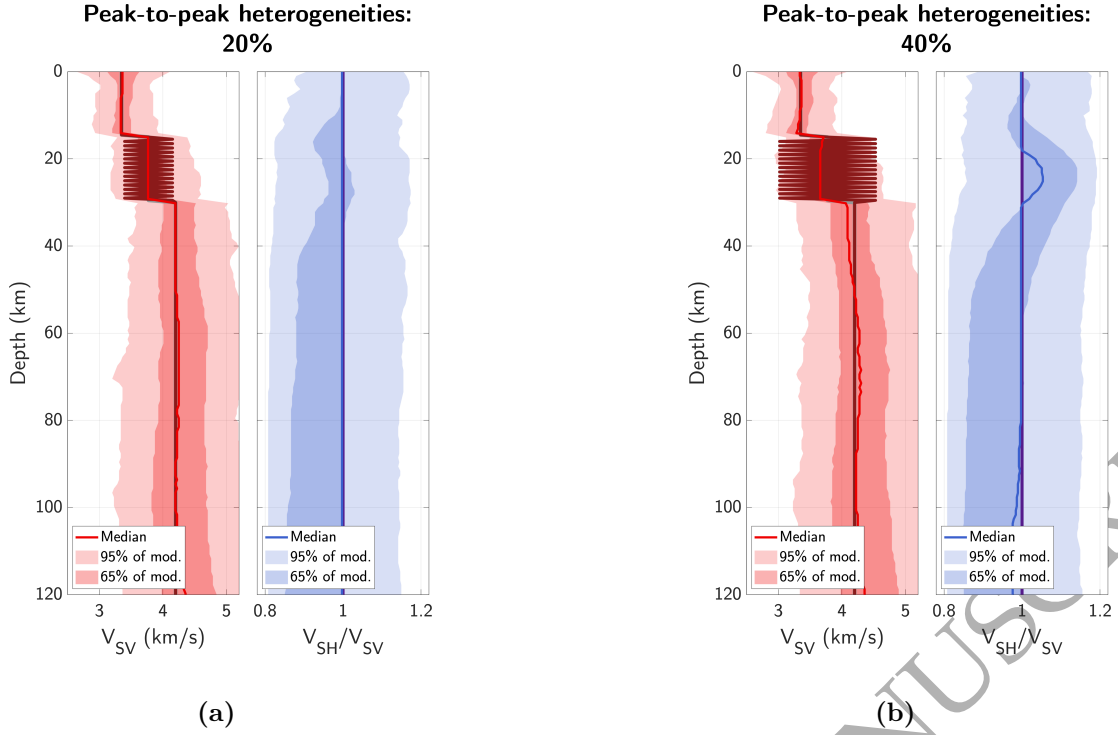


Figure 8. Synthetic tests where the lower crust in the true model is made of a stack of 150 identical lamellae of thickness 100 m, alternating between high and low velocities. Left: results where the lower crustal velocities in the true (target) model alternate between $V_{SV}=3.33$ km/s and $V_{SV}=4.07$ km/s (i.e. $\pm 10\%$ around 3.7 km/s, resulting in a peak-to-peak level of heterogeneity of 20%). Right: results where the lower crustal velocities alternate between $V_{SV}=2.96$ km/s and $V_{SV}=4.44$ km/s (i.e. $\pm 20\%$ around 3.7 km/s, resulting in a peak-to-peak level of heterogeneity of 40%). See legend of Figure 6 for the meaning of the different lines and colours.

32, 50 and 70 km in our final 3-D model. We show the posterior median dV_{SV} and radial anisotropy models in the upper and middle parts of each figure, respectively. The lower panels of the figures show the probability that anisotropy takes a positive ($V_{SH} > V_{SV}$) or negative ($V_{SH} < V_{SV}$) value. This probability is the number of models with positive or negative anisotropy expressed as a percentage of the total number of models. Figure 11 and Figure 12 show cross-sections in our model. The level of data fit of our 3-D probabilistic model is shown in Figure D.1.

5.1 Velocity structure

As explained before, the rejection of a large number of Rayleigh wave dispersion data in the selection procedure (see Section 2.2) implies that the resolution of our V_{SV} model is lower than that of models built from larger Rayleigh waves dispersion datasets (e.g. Lu et al. 2018).

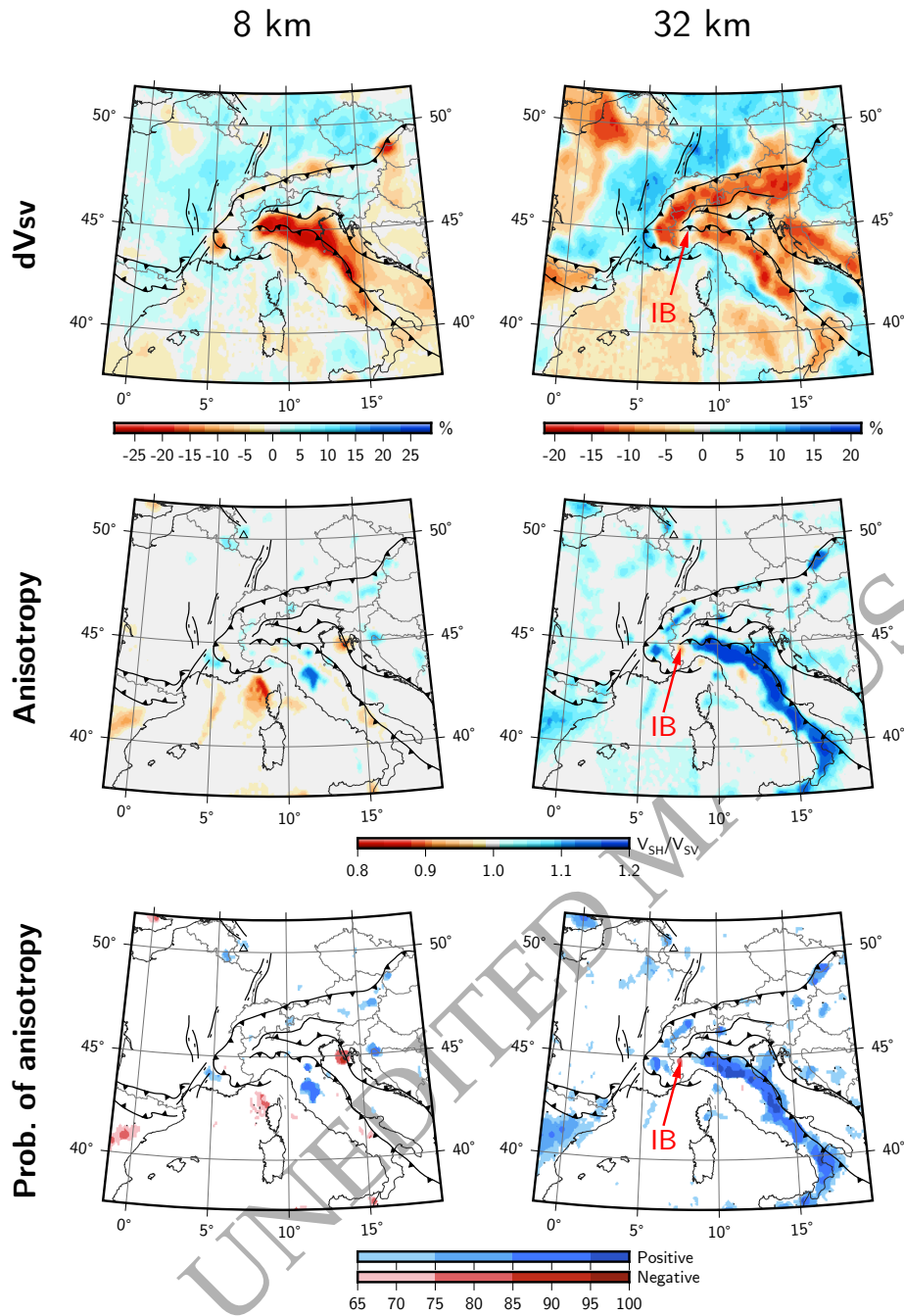


Figure 9. Depth slices at 8 and 32 km in our 3-D model displaying the posterior median model of velocity as perturbations around the mean of the map (top panels), the median radial anisotropy as the inverted parameter V_{SH}/V_{SV} (middle panels) and the probability of positive (blue) and negative (red) radial anisotropy (bottom panels). IB: Ivrea Body.

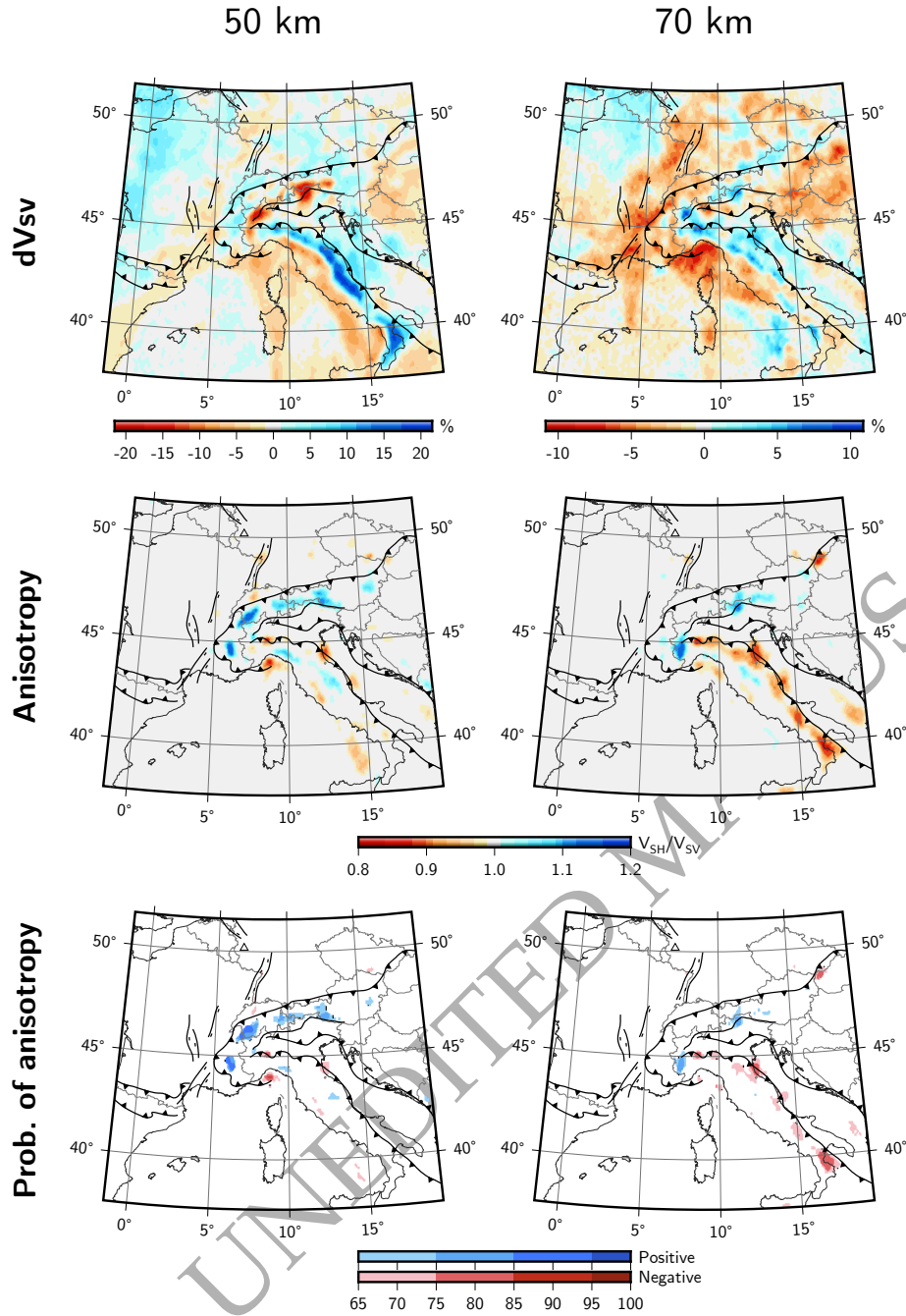


Figure 10. Same as Figure 9 but for depths 50 and 70 km.

The main focus of this work is therefore to image radial anisotropy in the crust and upper mantle of Europe. However, this work benefits from a good data coverage (see Figure A.1 in Appendix) which allows to image a few features that deserve to be briefly described in the following.

In the upper crust (8 km maps in Figure 9), the shear velocity pattern is coherent with

the group velocity maps at 8 s (Figure 4). Thick sedimentary basins such as the Po plain, the Adriatic basin, the Alpine foreland or the Vienna basin, are associated with the lowest velocity anomalies.

At 32 km (Figure 9), our shear velocity map highlights variations in Moho depth (Figure 9). The low velocity anomalies in the Alps, Apennines and Dinarides correspond to the crustal roots of these orogens. Slow anomalies are also visible in the Anglo-Paris basin in the Northern part of France, in relation with the thick crust observed in this area (e.g. Laske et al. 2013). The low velocity signature of the Alpine crustal root is limited to the East by the Pannonian basin, where fast anomalies are observed. This is coherent with the presence of a rather thin crust in the Pannonian basin (e.g. Balázs et al. 2016). Stronger high velocity perturbations are associated with the Cenozoic rift system, from the Rhine graben to South-Eastern France, where the crust is thinner (e.g. Molinari & Morelli 2011).

At 50 km depth, our model shows high velocity perturbations related to the Adriatic slab subducting under the Apennines (Figure 10). Beneath the Alps, the deep crustal root associated with low velocities is still visible, although it is interrupted by a fast anomaly under the Central Alps. This difference is due to the deeper Moho beneath the Alps than beneath the Apennines (Spada et al. 2013). The slab beneath the Apennines is delimited to the West by a sharp transition to slow anomalies associated with the shallow mantle under the stretched crusts of the Liguro-Provencal basin and the Tyrrhenian sea. A slab gap, which was documented by teleseismic travelttime tomography (Lucente et al. 1999; Giacomuzzi et al. 2011), is observed in the Southern part of the Apennines. It is also visible on profile EE' (Figure 11) where the fast anomaly associated with the slab is weaker between 500 and 600 km on the horizontal scale.

At 70 km depth (Figure 10), the high velocity anomaly associated with the slab shifts to the West, in agreement with the southwestward dip of the Adriatic subduction. On cross-section BB' (Figure 11), high velocity anomalies are observed in the mantle below the Apenninic front that clearly display the Adriatic slab plunging towards the South-West down to 100 km depth. The slab is however not visible on profiles CC' and DD' at depths larger than ~ 60 km.

Another noticeable feature of the velocity maps is the low velocity anomaly associated with the Eifel volcanic region. This anomaly is observed at 50 and 70 km (Figure 10) but not at shallower depths (Figure 9), in agreement with our group velocity observations. This is coherent with the presence of a shallow LAB reaching 40-50 km depth (e.g. Mathar et al. 2006; Seiberlich et al. 2013) due to an asthenospheric plume (e.g. Ritter et al. 2001; Keyser et al.

2002; Pilidou et al. 2005) spreading at the bottom of an unusually hot lithosphere (Schintgen et al. 2015).

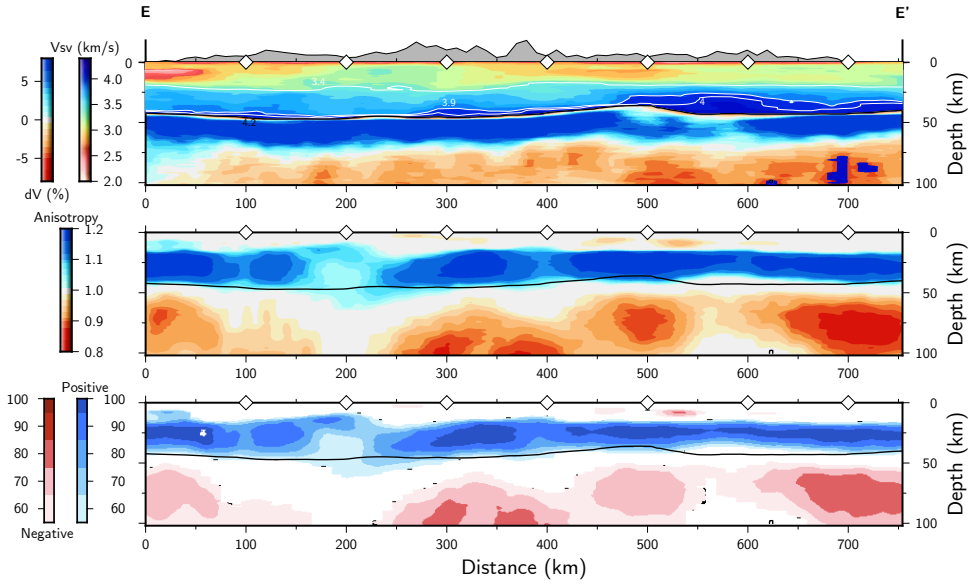
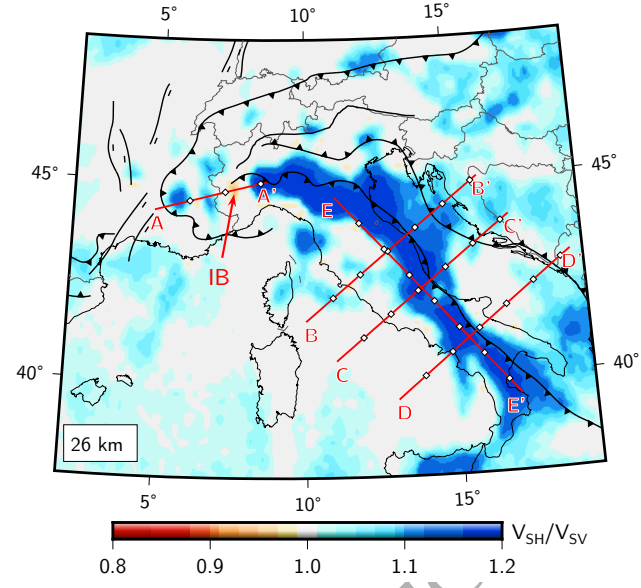
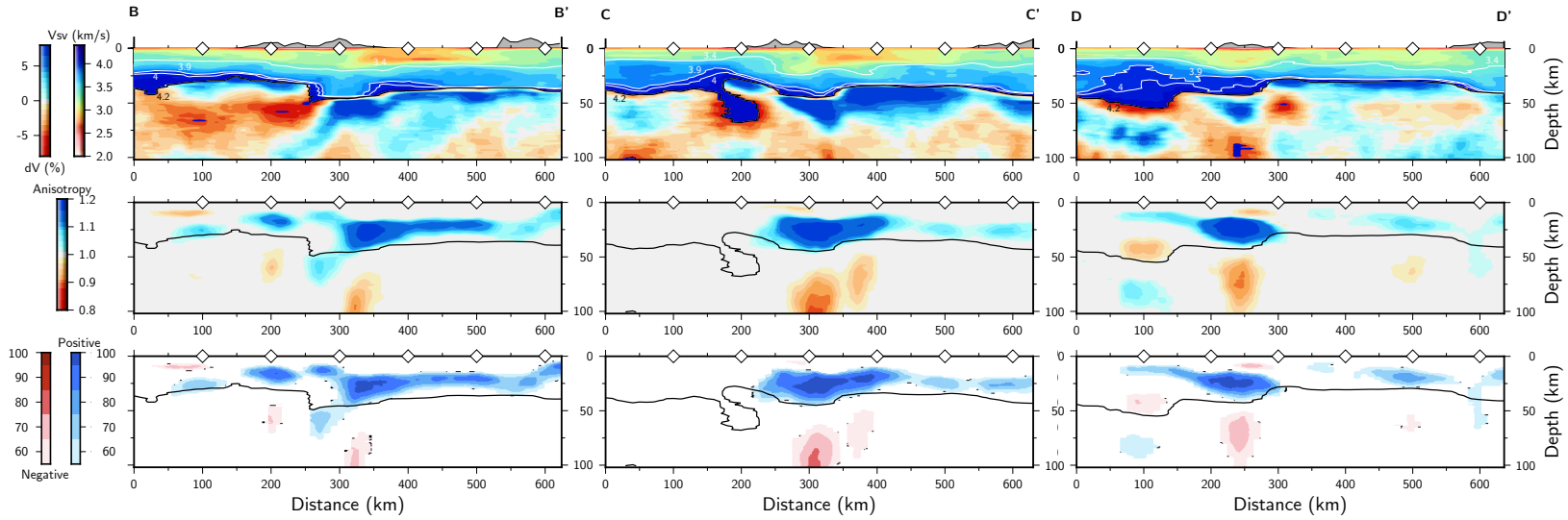
5.2 Anisotropic structure

As opposed to previous studies where radial anisotropy is observed in all parts of the study area (e.g. Zhu et al. 2015), the major part of our posterior median model is isotropic. It only exhibits radial anisotropy in some localized areas of the crust and uppermost mantle. The most striking pattern of anisotropy is observed from about 20 to 40 km depth along the Apennines, from the Northern part of the chain to the Calabrian arc (see maps at 26 and 32 km depth in Figure 11 and Figure 9 respectively). The observation of positive radial anisotropy with $V_{SH}/V_{SV} \approx 1.2$ is associated with a probability larger than 85% and is therefore considered as robust (i.e. 85% of the sampled models in the ensemble solution have positive radial anisotropy). The cross-sections across and along the chain (Figure 11) clearly suggest that this positive radial anisotropy anomaly is restricted to the lower crust. Given the synthetic tests shown in Figure 8, such a high value of V_{SH}/V_{SV} cannot result from isotropic hectometric layering in the lower crust. In the underlying mantle, negative ($V_{SH} < V_{SV}$) anisotropy is observed with a high probability from ~ 60 to ~ 100 km depth (Figure 11). This anomaly coincides with the high velocity perturbation interpreted as the Adriatic slab.

Another robust anisotropy pattern, although corresponding to a much smaller area, is observed on the 32-km depth slices beneath the westernmost Po basin (IB in Figure 9) and on the CIFALPS profile (at ~ 220 km distance in Figure 12). On these figures, a negative anisotropy anomaly ($V_{SV} > V_{SH}$) is associated with the high V_{SV} anomaly attributed to the Ivrea Body. The Ivrea body (IB) is a high-velocity high-density body located at depths ≥ 10 km in the crust of the Western Alps, which is interpreted as a flake of Adriatic upper mantle (Closs & Labrouste 1963; Nicolas et al. 1990). To our knowledge, this is the first time that a strong anisotropic signature is associated with the Ivrea body.

Figure 11: 2-D cross-sections along four profiles in the Apennines. Locations of the profiles BB' to EE' are displayed on the map of radial anisotropy at 26 km depth, in the bottom left corner of the figure. AA' cross-section corresponds to the CICALPS profile (Zhao et al. 2015) and is represented in Figure 12. For each profile, the top panel displays absolute velocities in the crust and perturbations of velocity in the mantle. The middle and bottom panels display radial anisotropy as V_{SH}/V_{SV} and the probability of anisotropy respectively. Moho depth (solid black line on each profile) is defined as the iso-velocity 4.2 km/s, as in PREM (Dziewoński & Anderson 1981). IB: Ivrea body.

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Evidence for radial anisotropy in the Apennines 29

6 DISCUSSION

6.1 Comparison with previous studies

Our 3-D shear wave velocity model is in good agreement with recent high resolution ambient noise tomography studies (Kästle et al. 2018; Lu et al. 2018, 2020) that show similar large-scale structures in the crust and uppermost mantle. Even though the spatial resolution of our model does not reach that of these previous models, some features are well imaged, such as the Adriatic slab under the Apennines.

The most original contribution of this study comes from the radial anisotropy model. Our most striking result is that radial anisotropy is clearly localized and not pervasive in the crust and upper mantle. This observation contrasts with previous studies (Kustowski et al. 2008; Boschi et al. 2009; Schivardi & Morelli 2011; Zhu et al. 2015) where radial shear wave anisotropy is required everywhere at crustal and mantle depths in Europe to explain the observed Love/Rayleigh discrepancy. One of the origins for this difference may be the unequal data coverage between Rayleigh and Love waves in these studies, but it most likely arises from the fact that we use a massive dataset that includes short period measurements (down to 5 s) combined with a Bayesian inversion that efficiently accounts for possible layering, as shown by our synthetic tests in Figure 8. In previous studies based on optimization schemes, the level of vertical smoothness is fixed in advance, and the solution represents the best fitting model given the imposed parameterization. By introducing anisotropy as a free parameter, seismic data can be fitted with smoother models and fewer spatial parameters (Montagner & Jobert 1988; Trampert & Woodhouse 2003). A heterogeneous isotropic medium may then be imaged as a smooth anisotropic model. In the transdimensional scheme used here, no vertical scale is imposed, as neither the level of layering nor the presence of anisotropy is determined in advance. Although the first step of our tomographic inversion involves a standard least squares approach with a level of lateral smoothing determined by the horizontal correlation length L_{corr} , radial anisotropy is involved only in the second step, where we invert the group velocity maps for the 3-D model. As this stage, the transdimensional inversion is free to choose between layering and radial anisotropy, while respecting the parsimonious condition where simple models described by fewer parameters are favoured. For this reason, the existence and level of radial anisotropy is directly determined by the data and not by a selected degree of smoothing.

At each geographical point, the posterior distribution of layered models obtained after our 1-D transdimensional Bayesian inversion exhibits models with many isotropic layers (resulting in

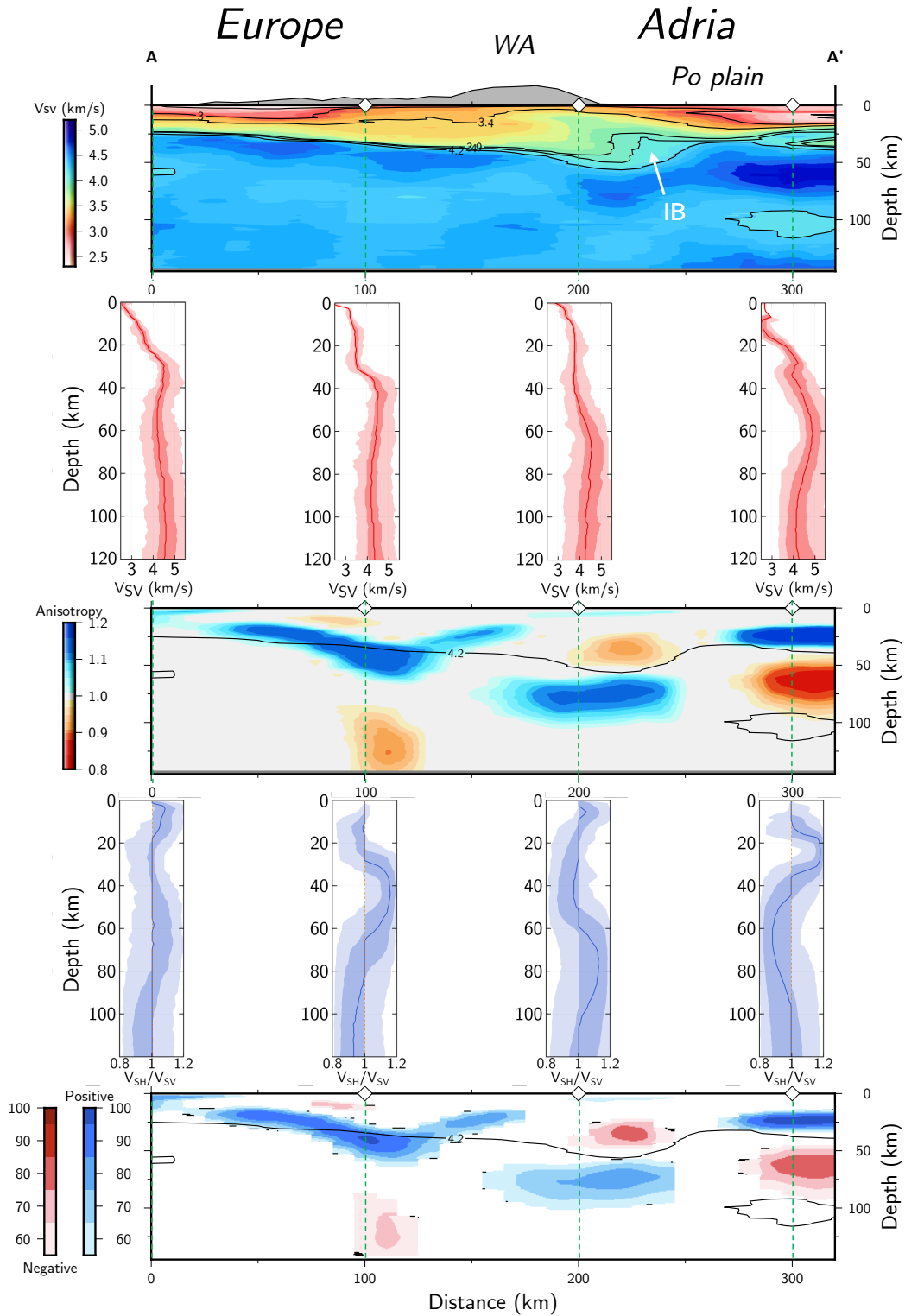


Figure 12. 2-D cross-sections along the CIFALPS profile (AA' in Figure 11) and posterior distributions of velocity (red 1-D profiles) and radial anisotropy (blue 1-D profiles) as described in Figure 6 for four points along the profile. Top panel: absolute velocity; middle panel: radial anisotropy as V_{SH}/V_{SV} ; bottom panel: probability of positive (blue) and negative (red) radial anisotropy. IB: Ivrea Body ; WA: Western Alps

a median V_{SH}/V_{SV} value of 1). Our model then suggests that the Rayleigh/Love discrepancy observed in the data can be explained by small scale (hectometric) layering in the crust and uppermost mantle in most of Europe, except in the Apennines. Such small-scale layering is likely to be present in the lower crust of Variscan Europe, as imaged by deep reflection seismics, and it could result from layering of mafic/ultramafic intrusions of partial melts derived from the upper mantle (Singh & McKenzie 1993; Rey 1993). Our synthetic tests (Figure 8) suggest that thin layering with up to 30% velocity contrasts would not be mapped as anisotropy with our inversion. This suggests that when accounting for possible layering, radial anisotropy is not necessarily required to reconcile Love and Rayleigh observations beneath most of Europe.

6.2 A focus on the Apennines

The Adriatic slab is rather well imaged in our model under the Apennines (Figure 10). Previous studies report a very steep angle of subduction towards the West in the Northern Apennines (Koulakov et al. 2009; Kästle et al. 2018; Zhao et al. 2016a). Our model rather shows a slab subducting towards the West with a moderate angle (profile BB' Figure 10). The slab gap observed in the Southern Apennines is consistent with the gap previously reported by e.g. Piromallo & Morelli (2003); Koulakov et al. (2009) from body wave tomography and by Kästle et al. (2018) from surface wave tomography. As suggested by Carminati & Doglioni (2012), its existence could be related to different convergence rates between Northern and Southern Apennines.

Strong positive radial anisotropy is observed with a high probability in the lower crust of the chain. We show in Figure 8 that this high level of anisotropy cannot be ascribed to thin layering as described e.g. in Singh & McKenzie (1993). Our results rather suggest the horizontal alignment of the fast axes of anisotropic minerals such as amphiboles and micas (see Almqvist & Mainprice 2017, for a review on anisotropy in the continental crust). We propose that this lattice preferred orientation of minerals is related to horizontal ductile flow in the lower crust of the Apennines in response to the recent and present-day extensional regime observed in the chain from structural, seismotectonic and GPS data (e.g. D'Agostino et al. 2008) and caused by the subduction of Adria both under the Apennines and Dinarides. In the upper mantle, anisotropy is commonly associated with the orientation of fast axes of olivine crystals, as olivine is the dominant mineral in the mantle. A negative radial anisotropy in the sublithospheric mantle suggests vertical flow (Montagner 1994). Negative radial anisotropy is observed at sublithospheric mantle depths beneath the Apennines, under

the slab. It suggests sub-vertical orientation of olivine fast axes, likely related to the westward dipping slab.

6.3 Anisotropy along the CIFALPS profile

Along the CIFALPS profile (AA' profile in Figure 12), the European Moho reaches 50 km depth under the Ivrea body (Figure 12), while Zhao et al. (2015) from a receiver function study, and Zhao et al. (2020) from a surface wave ambient noise study have documented a deeper Moho depth of 70-75 km in a very localized zone (<50 km in width). Zhao et al. (2020) applied a transdimensional Bayesian inversion scheme similar to ours but in the inversion of Rayleigh wave group velocity data for a 3-D isotropic V_{SV} model. The differences between our and their V_{SV} image can be explained by our selection procedure, where we imposed an equal coverage between Love and Rayleigh measurements, thus resulting in a decrease of Rayleigh wave measurements compared to recent isotropic studies. It is therefore likely that we do not resolve the small area where the Moho exceeds 50 km depth.

In the Ivrea body high V_{SV} anomaly, we observe negative radial anisotropy ($V_{SV} > V_{SH}$) at 25-50 km depth along the CIFALPS profile (Figure 12). The anisotropy map at 32-km depth (Figure 9) shows that the anisotropy anomaly does not follow the North-South extension of the Ivrea body documented by its positive Bouguer anomaly (Masson et al. 1999). It rather appears as a small anisotropic spot inside a mostly isotropic body. Despite its small size, the anisotropy anomaly is reliable as documented by its high probability (Figure 9). Therefore, we propose that it is related to the presence of highly-anisotropic serpentinites in the Ivrea body and/or in the subduction complex beneath it, with strongly-dipping fast-velocity directions (Solarino et al. 2018; Zhao et al. 2020). This interpretation should be valid for the entire Ivrea body, including the parts that appear isotropic to the North and South of the CIFALPS profile. A possible explanation for the absence of anomaly is the resolution of our tomography, which may be insufficient to image such a narrow body (~50-70 km wide). Another interpretation would be to link the negative anisotropy to the set of sub-vertical faults that produce numerous microearthquakes at 20-75 km depth beneath the Westernmost Po plain, precisely at the same location as the anisotropy spot (Malusà et al. 2017). This interpretation would explain why other parts of the Ivrea body appear isotropic, as no anomalously deep earthquakes are recorded there.

The pattern of positive radial anisotropy in the upper mantle under the Ivrea body, from 55 to 100 km depth suggests the presence of frozen-in anisotropy (e.g. Yuan & Romanowicz 2010) in the lithosphere, due to past deformation. The same interpretation can be applied to

the positive anisotropy observed at Moho depth under Europe. This would however require further investigation.

7 CONCLUSIONS

We present the first 3-D radially anisotropic probabilistic model of the crust and uppermost mantle of Western Europe with a lateral resolution of a few tens of kilometers. This model is derived from Rayleigh and Love waves group velocity measurements in the period range 5 to 149 s. The dispersion data are obtained from the correlation of ambient noise at more than 700,000 station pairs in Europe. The collection of such a large dataset, which is a difficult task in particular for Love wave data derived from horizontal components of ambient noise records, has been made possible by the recent deployment of the AlpArray temporary seismic network (Hetényi et al. 2018).

We acknowledge here that a number of approximations have been made to construct this 3-D model, that may result in a number of errors and biases:

(i) sources of noise are not uniformly distributed, which results in an approximate reconstruction of Green's functions from ambient noise correlation.

(ii) azimuthal variations of group velocities were ignored in the 2-D inversion. These unmapped azimuthal variations can result in biased group velocity maps.

(iii) Love and Rayleigh waves are measured on different components, and their measurements may be contaminated by different contributions. Although we have ensured that we have similar data coverage, these different levels of error may have resulted in different lateral resolutions between Love and Rayleigh group velocity maps, and thus producing some apparent isotropic or anisotropic heterogeneities.

(iv) the Bayesian inversion of dispersion curves was done assuming a radially symmetric Earth, and group velocities were computed by normal mode summation. That is, dispersion curves were assumed to be only sensitive to the 1-D structure below the considered point, and 3-D structural effects have been ignored.

(v) a discontinuous reference model was imposed as the mean of the prior distribution. Our results shown here depend on this chosen reference model.

However, the large-scale structures of our 3-D shear wave velocity model are in good agreement with studies conducted in the same region from Rayleigh wave dispersion data (e.g. Kästle et al. 2018; Lu et al. 2018, 2020; Zhao et al. 2020). The main outcome of our work concerns the pattern of radial anisotropy, which is present only in restricted areas of our model. This is

in stark contrast to previous large scale studies that showed radial anisotropy throughout the crust and upper mantle (Kustowski et al. 2008; Boschi et al. 2009; Schivardi & Morelli 2011; Zhu et al. 2015). We attribute this difference to the thin hectometric layering of the lower crust, which is ubiquitous in the Variscan crust of Western Europe and can be mapped as radial anisotropy in large-scale tomographic models. Our Bayesian inversion applied to a very large dataset, including short period measurements down to 5 s, produces radial anisotropy in a more parsimonious way, only in locations where it is required by data. Our synthetic tests demonstrate that the transdimensional inversion does not introduce artificial radial anisotropy in the model, even when strong layering is present.

Our most prominent and robust observation is positive radial anisotropy ($V_{SH} > V_{SV}$) in the lower crust of the Apennines. We attribute this to horizontal flow in the ductile lower crust in response to the recent and present-day extension observed from structural geology, earthquake focal mechanisms and geodesy.

In order to improve the model, future work may involve the use of higher frequency data such as converted or reflected body waves, in a joint inversion with surface wave measurements. Indeed, higher frequency content would allow to retrieve the depth of seismic discontinuities thus avoiding the need of a reference velocity model in the inversion at depth.

Dense temporary networks similar to AlpArray are expected to be deployed in Europe in the coming years (e.g. AdriaArray). Applying the same procedure we used to a larger surface waves dataset combining measurements obtained from those different arrays would highly improve the resolution of radially anisotropic models in Europe.

8 AVAILABILITY OF THE MODEL AND CODES

Our tomographic model and the Bayesian inversion code are available on request to Chloé Alder. The regionalization code is available on request to Eric Debayle.

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9 APPENDIX

APPENDIX A: RAY DENSITY

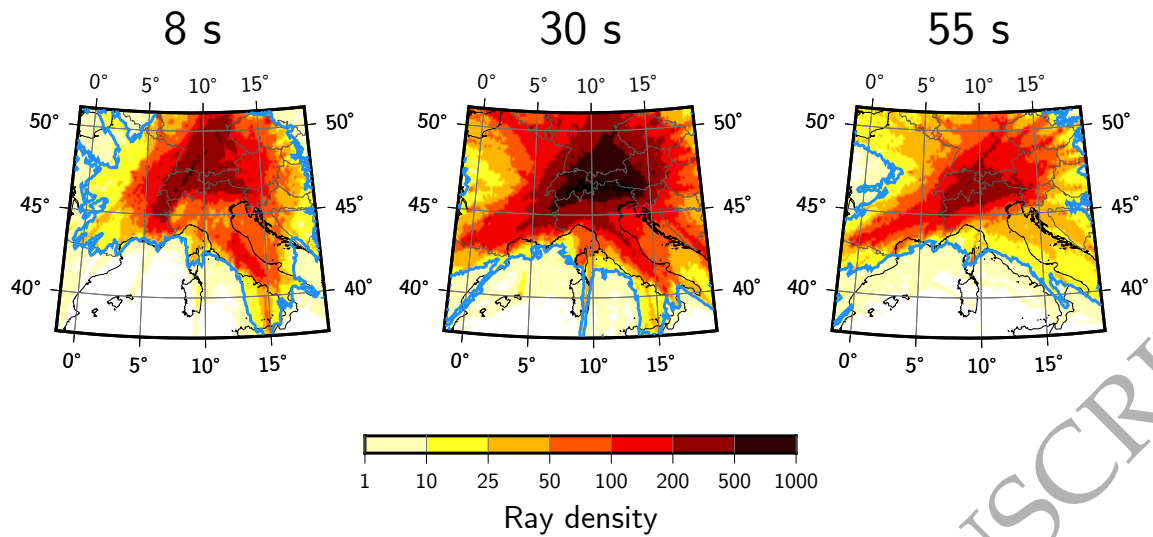


Figure A.1: Ray density maps at periods 8 s, 30 s and 55 s showing the number of paths crossing each $0.15^\circ \times 0.15^\circ$ cell. The blue line delimits the area with ray density higher than 10.

APPENDIX B: FLAT TESTS

To assess the lateral resolution of the 2-D group velocity maps, we used flat tests where we estimate how a known and uniform perturbation of velocity is recovered by the regionalization method (i.e. 2-D tomography) of Debayle & Sambridge (2004) given our data coverage, choices of parameterization and regularization. For each period, we compute synthetic group velocities between receivers of every station pair in the dataset assuming, at each grid point, a velocity perturbation of 15% in amplitude with respect to the reference velocity calculated as the mean of the real data. The 2-D tomography is then performed with the same parameters used in the data inversion and discussed in Section 3.1. Well resolved areas are the ones where the recovered perturbation is close to the one we introduced.

Figure B.1 shows the results of flat tests performed for six periods between 8 and 90 s, including periods 8, 30 and 55 s discussed in the text. From those tests, we show that our data coverage and chosen parameterization and regularization enable us to recover more than 90% of the amplitude of the initial perturbation for periods as long as 80 s in the Alps and Northern Italy. In the same period range, at least 30% of the input velocity anomaly is recovered in most of Europe, except in the Mediterranean sea. At longer periods, we lose resolution in Italy. In addition, some smearing is observed in the SW-NE direction, in Southern France, from period 55 s.

Finally, these results show that the smoothing parameters, i.e. the correlation lengths, are adapted to the data coverage at every period since no velocity oscillation is observed around the ray paths.

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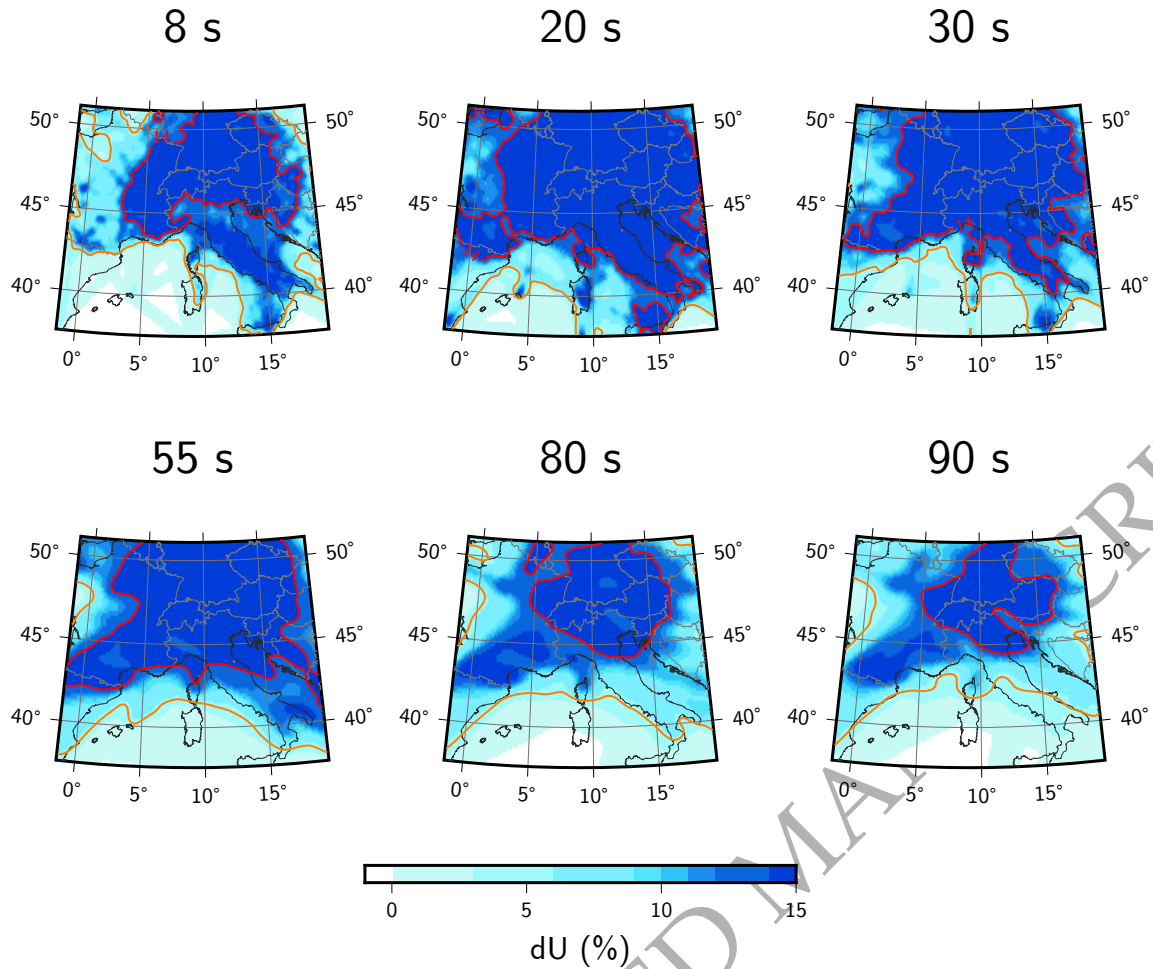


Figure B.1: Results of flat tests. The input anomaly represents a velocity perturbation of 15% at each grid point, i.e. a flat model. The recovered perturbations after regionalization are shown for periods 8, 20, 30, 55, 80 and 90 s. Red lines represent a velocity anomaly of 14%, i.e. 90% of the input perturbation. Orange lines represent an anomaly of 5%, i.e. 30% of the input model.

APPENDIX C: ADDITIONAL GROUP VELOCITY MAPS

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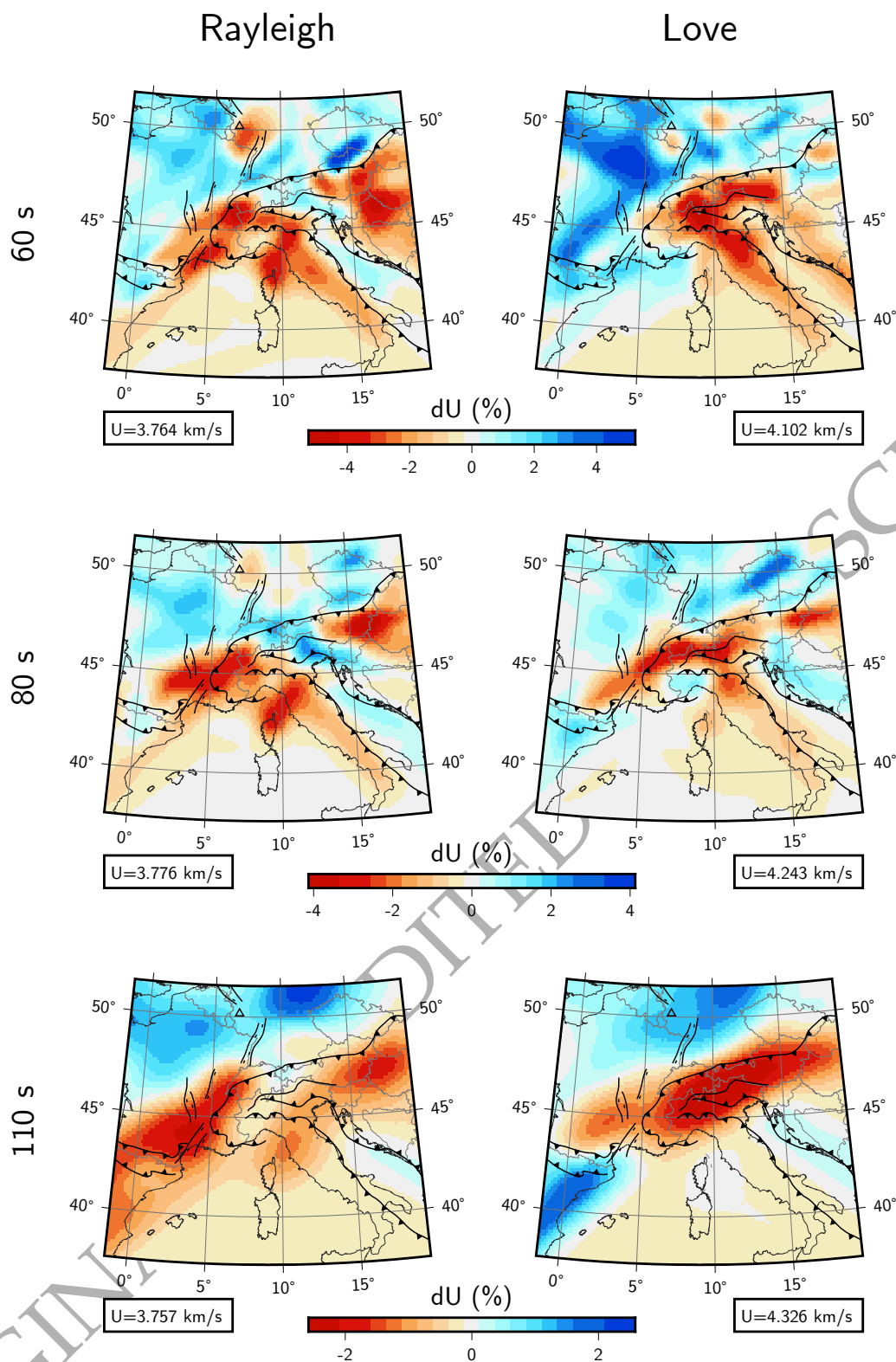


Figure C.1. Group velocity perturbation maps for Rayleigh and Love waves at periods 60, 80 and 110 s. The reference group velocity U is indicated below each map.

APPENDIX D: DATA FIT

Our solution model is not a single velocity field but instead a probability distribution in the model space, and computing a single data misfit value is not possible. We decide to compare observations to the data vector obtained after averaging over the ensemble of data estimated for the ensemble of models in the solution. We compute separately the fit to Love and Rayleigh data at each period between 5 and 149 s, with an increment of 5 s.

For each geographical point, the misfit between observed data, and data obtained from the probabilistic solution is given by:

$$\Phi_R(T) = \frac{\sqrt{(u_R(T, \mathbf{m}) - U_R(T))^2}}{U_R(T)} \quad (\text{D.1})$$

$$\Phi_L(T) = \frac{\sqrt{(u_L(T, \mathbf{m}) - U_L(T))^2}}{U_L(T)} \quad (\text{D.2})$$

where, for Rayleigh and Love waves respectively, $\Phi_R(T, \mathbf{m})$ and $\Phi_L(T, \mathbf{m})$ are the mean of the estimated group velocities at period T from each model of the posterior distribution. $U_R(T)$ and $U_L(T)$ are the velocities extracted at T from the Rayleigh and Love 2-D velocity maps. Misfit maps are shown in Figure D.1 for periods 10, 30 and 55 s. The misfit is the highest at short periods (10 s on the maps), in thick sedimentary basins such as the Po plain or the Vienna basin, and could be related to the complex structure of the basins not perfectly retrieved in our models.

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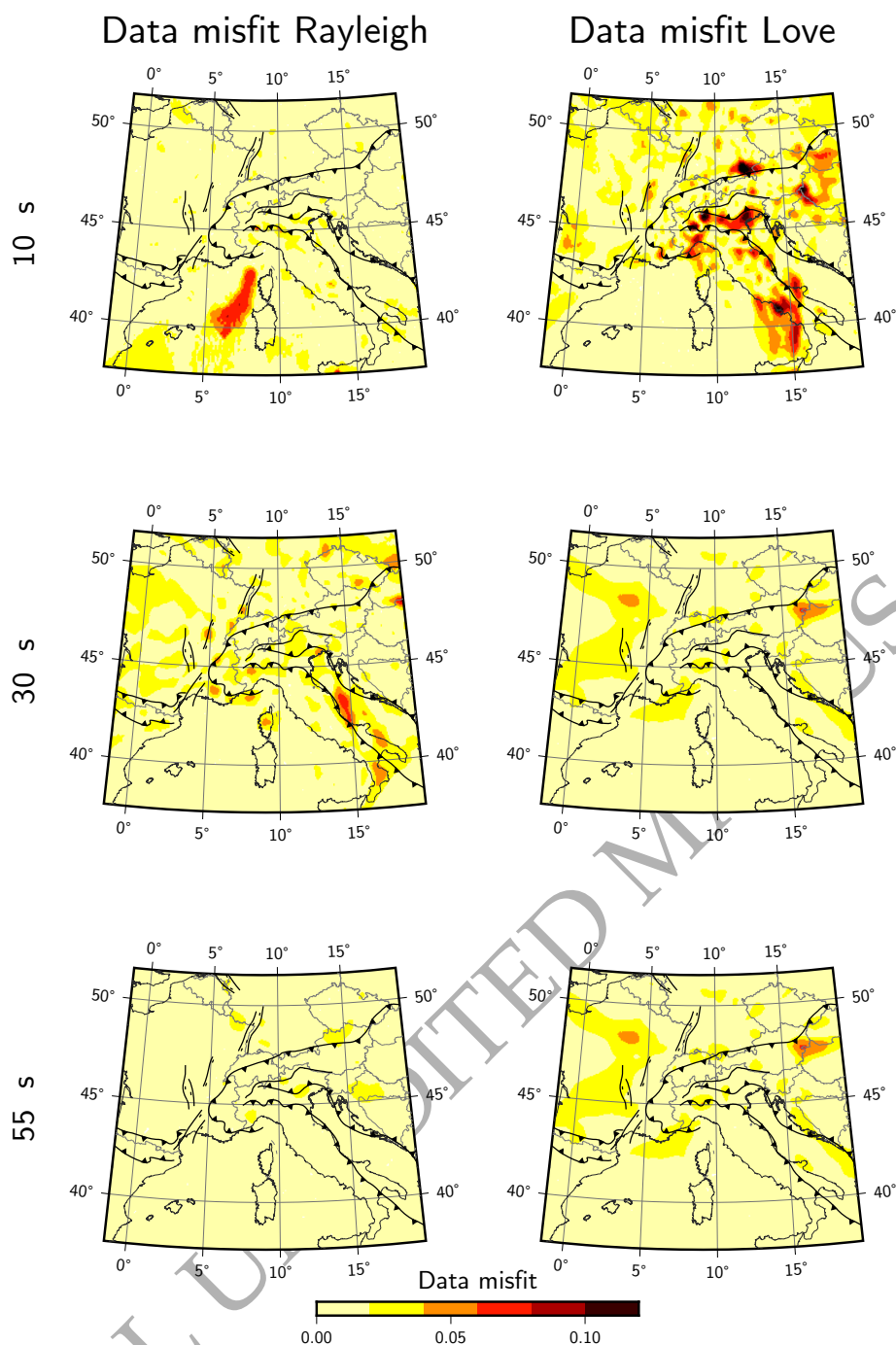


Figure D.1. Data misfit for Rayleigh and Love waves at 10, 30 and 55 s, as defined in Eq. D.2.

APPENDIX E: DATA ORIGIN

Seismic ambient noise data used in this study were collected from networks with codes 1G: Thybo et al. (2012); 2D; 4A: Passarelli et al. (2012); 4C; 6E; 6G: Sens-Schönfelder & Delatre (2011); 6H; 8A: Dias et al. (2010); 8X; 9C: Heit et al. (2010); AC: Institute Of Geosciences,

Energy, Water And Environment (2002); BA; BE: Royal Observatory Of Belgium (1985); BN; BS: National Institute Of Geophysics, Geodesy (1980); BW: Department Of Earth And Environmental Sciences, Geophysical Observatory, University Of Munchen (2001); C4; CA: Institut Cartogràfic I Geològic De Catalunya (1984); CH: Swiss Seismological Service (SED) At ETH Zurich (1983); CL: Corinth Rift Laboratory Team And RESIF Datacenter (2013); CQ: Geological Survey Department Cyprus (2013); CR: University Of Zagreb (2001); CZ: Institute Of Geophysics (1973); DK; DZ; EB; EE; EI: INSN (1993); ES: Instituto Geografico Nacional, Spain (1999); FN; FR: RESIF (1995); G : Institut De Physique Du Globe De Paris (IPGP) & (EOST); GB; GE: GEOFON Data Centre (1993); GR: Federal Institute for Geosciences and Natural Resources (BGR) (1976); GU: University Of Genova (1967); HA: University Of Athens (2008); HC: Technological Educational Institute Of Crete (2006); HE: Institute Of Seismology, University Of Helsinki (1980); HF; HL: National Observatory Of Athens, Institute Of Geodynamics (1997); HP: University Of Patras, Geology Department (2000); HT: Aristotle University Of Thessaloniki Seismological Network (1981); HU: Kövesligethy Radó Seismological Observatory (Geodetic And Geophysical Institute, Research Centre For Astronomy And Earth Sciences, Hungarian Academy Of Sciences (MTA CSFK GGI KRSZO)) (1992); IB: Institute Earth Sciences "Jaume Almera" CSIC (ICTJA Spain) (2007); II: Scripps Institution Of Oceanography (1986); IM; IP; IS; IU: Albuquerque Seismological Laboratory (ASL)/USGS (1988); IV: INGV Seismological Data Centre (1997); IX; KO: Bogazici University Kandilli Observatory And Earthquake Research Institute (2001); LC: Laboratorio Subteraneo De Canfranc (2011); LX: Instituto Dom Luiz (IDL)-Faculdade De Ciências Da Universidade De Lisboa (2003); MD: Geological And Seismological Institute Of Moldova (2007); ME: Sector For Seismology, Institute Of Hydrometeorology (1982); MN: MedNet Project Partner Institutions (1988); MT: French Landslide Observatory – Seismological Datacenter / RESIF (2006); NI: OGS (Istituto Nazionale Di Oceanografia E Di Geofisica Sperimentale) And University Of Trieste (2002); NL: Royal Netherlands Meteorological Institute (KNMI) (1993); NO; NS; OE: ZAMG-Zentralanstalt Für Meteorologie Und Geodynamik (1987); OT: University Of Bari "Aldo Moro" (2013); OX: OGS (Istituto Nazionale Di Oceanografia E Di Geofisica Sperimentale) (2016); PL; PM: Instituto Português Do Mar E Da Atmosfera, I.P. (2006); RD: RESIF (2018); RF: University Of Trieste (1993); RO: National Institute For Earth Physics (NIEP Romania) (1994); SI; SJ; SK: Earth Science Institute Of The Slovak Academy Of Sciences (ES-ISAS) (2004); SL: Slovenian Environment Agency (2001); SS; ST: Geological Survey-Provincia Autonoma Di Trento (1981); SX: Leipzig University (2001); TH: Jena, Friedrich Schiller University (2009); TT; TU: Disaster And Emergency Management Presidency (AFAD Turkey)

(1990); TV; UK; UP: SNSN (1904); WM: San Fernando Royal Naval Observatory (ROA) et al. (1996); X7: Chevrot et al. (2017); XT: AlpArray Seismic Network (2014); Y4: Roessler et al. (2014); Y7; YF; YH: DANA (2012); YI: Chaljub (2017); YP: Zhao et al. (2016b); YR: Segou et al. (2016); YW: Guéguen et al. (2017); YZ: Passarelli et al. (2017); Z3: AlpArray Seismic Network (2015); ZH: Deschamps & Beucler (2013); ZS: Heit et al. (2017); ZU: Chevrot et al. (2018); ZW.

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