Supporting information for

The unevenness of the North Iberian crustal root, a snapshot of an elusive stage in margin reactivation.

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Text S1. Crustal thickness from ambient seismic noise autocorrelation

Autocorrelations of ambient seismic noise can be used to image lithospheric discontinuities below single seismic stations. The strategies to calculate this autocorrelations are quite diverse and different processing of data is applied in accordance with the dataset characteristics. The main differences are the length of the time windows to autocorrelate, the pre- and/or post-correlation time or frequency normalization and the limits of the applied band-pass filter (Becker and Knapmeyer-Endrun 2018; Andrés et al. 2020). Our processing workflow follows the ambient noise cross-correlation procedure of Bensen *et al.*, (2007) with several modifications based on the tests performed by Romero and Schimmel, (2018). Overall, it can be divided in four stages: preprocessing, construction of the stacked autocorrelograms (ACs), quality control and crustal depth determination. Here we show an example of AC calculation for station LIEG (Fig. S1).

1) Pre-processing

Vertical component data were downsampled to 25 Hz and cut into non-overlapping 3-hour fragments. Then the mean, the trend and the instrument response were removed. A crucial step is the application of a zero-phase band-pass filter to enhance the frequency band at which we expect to found the target information. In general, higher frequencies are used to extract the shallow subsurface response (e.g. 3-12 Hz in Romero and Schimmel, 2018), whereas lower frequency bands are employed to resolve deeper discontinuities. After several tests with different frequency bands, we found that the 2-4 Hz range

recovered the main reflections and showed a good daily consistency (Fig. S2A). This frequency band has also been used in other autocorrelation studies (e.g. Gorvatov *et al.*, 2013; Kennett *et al.*, 2015; Romero and Schimmel, 2018).

2) Construction of the stacked autocorrelograms

We calculated the ACs using the phase cross-correlation (PCC) technique of Schimmel, (1999). This method is amplitude unbiased, which effectively avoids the necessity of applying a time or spectral normalization that may deteriorate the waveforms. The phase AC were computed for each 3-hour fragments with 25 s time lags and stacked using the time-frequency phase weighted stack (Schimmel and Gallart, 2007), a linear stacking procedure that enhances coherent arrivals and attenuates incoherent signals. The final ACs are obtained from the stacking of all the individual 3-hour autocorrelograms. 3-hour segments from the same day have also been stacked to create daily Acs, which are useful for testing the performance of the method.

3) Quality control

The quality of the final ACs has been tested in three steps. First, the stability at different frequency bands was investigated and coherent reflectors have been observed (Fig. S2B). Next, stacks of randomly selected sets of 60, 90, 120 and all the available daily ACs have been computed in order to assess their temporal consistency and check if coherent arrivals can be identified. As shown in Fig. S2B, Moho related reflectors can be retrieved using as less as ~2 month of data, a much lower number than the one used in this work. Finally, daily autocorrelation plots (Fig. S2C) facilitate the recognition of anomalous signals that can be removed from the calculations.

4) Crustal depth determination

In order to determine the Moho depth, the lag time of the maximum amplitude of the crust-mantle boundary originated reflection (Tibuleac and von Seggern,2012) is picked in the autocorrelograms. However, due to the presence of several reflections, the unique identification of this specific signal can be problematic. To reduce this uncertainty, we applied the automatic approach of Becker and Knapmeyer-Endrun (2018) with a few modifications. This method begins with the filtering of the stacked autocorrelograms between 2 and 4 Hz with a four-order Butterworth filter (Fig. S2D). Next, the second derivative of the envelope of the resulting AC signal is calculated. Finally, an amplitude normalization using an automatic gain control is performed to emphasize the amplitude variations of the signal (Fig. S2D). The greatest changes are linked to the largest fluctuations in the slope of the envelope. At this

point, Becker and Knapmeyer-Endrun (2018) select the local maxima of the second derivative within a time window for expected Moho reflections based on prior information from the study area. However, we have observed that picking the local minima instead of the local maxima yields better results for our datasets, because it tends to correlate with the maximum amplitude of an identifiable reflectivity change in the autocorrelogram's waveform (Fig. S2D).

Our lag time windows for retrieving Moho reflections are based on the crustal thickness maps derived from deep seismic sounding studies and receiver function determinations (Table S1, Diaz *et al.*, 2016) plus a ± 1.5 s lag time uncertainty (Fig. S2D). Considering the large differences between the crust-mantle boundary topography obtained from each of this techniques, up to 15 km below the Cantabrian Mountains, time windows for each station are long enough to ensure that the selected Moho lag time is picked with a sufficient degree of freedom. Lag times, which are expressed in two-way-time, are converted to depth using an average crustal P-wave velocity of 6 km/s based on Vp velocities obtained in refraction profiles through the study area by Fernández-Viejo *et al.* (2000). The P-wave velocity for depth conversion can be a source of error in the measurements, so it is recommended to assume a 5% velocity uncertainty. In practice, this results in an average error of ± 2.1 km for the Moho depth determinations (Table S1).

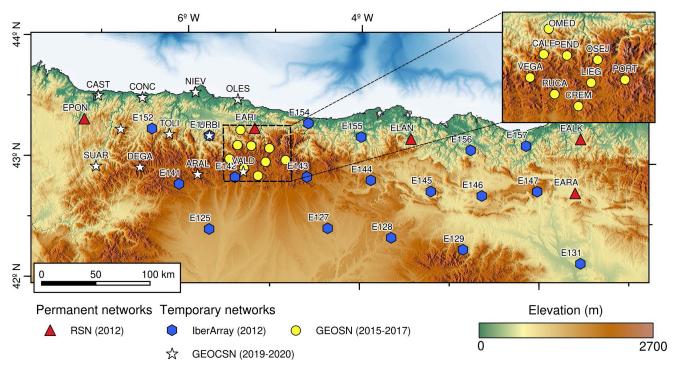


Figure S1. Topographic map showing the location of the permanent seismic stations of the Spanish Seismological Network (RSN) and the portable stations belonging to the IberArray, GEOSN (Geocantábrica Seismic Network) and GEOCSN (Geocantábrica-Costa Seismic Network) temporary experiments.

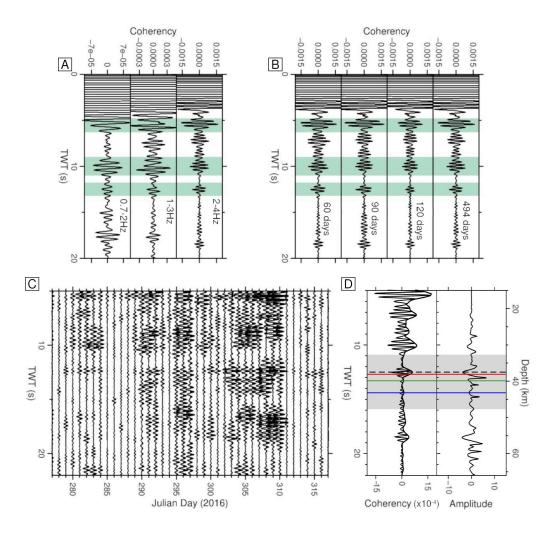


Figure S2. Quality tests and crustal depth determination for station LIEG (Location in Figure S1). A) Example of autocorrelogram stacks computed at frequency bands between 0.7-2Hz (left), 1-3Hz (center) and 2-4Hz (right). Green rectangles highlight coherent identified reflections. TWT: two-way-time. B) Secuential stacks of 60, 90 120 and 494 daily autocorrelograms. Note how the same reflections of Fig. S1a (green surfaces) are retrieved. C) Daily autocorrelation section showing a consistent reflection at ~12.5 s. D) Automatic picking of the Moho lag time by selecting the local minima of the second derivative of the envelope (right) of the total autocorrelogram stack (left). Note how this minima coincides with the maximum amplitude of a coherent reflection that can also be observed in the daily autocorrelation section. The grey surface represents the expected time window for Moho reflections based on the calculated lag times from previous receiver function (RF, red) and seismic data (R/WAR, blue) determinations (Table S1, Díaz et al, 2016) extended with a ±1.5s lag time uncertainty.

Table S1. Overview of Moho depths from ACs and R/WAR. Moho depths obtained by Díaz *et al.* (2016) from receiver functions (RF) and R/WAR used to determine AC lag time windows. Lag times

corresponding to Moho depths are shown in brackets. They are expressed in two-way-time and calculated using Vp=6~km/s.

Network Station Moho depth RF (km) Moho depth R/WAR (km) Moho depth R/WAR (km) Moho depth AC (km) GEOSN CALE 35 (11.7) 42 (14) 45.2 (15.1) 43.5 (14.5) GEOSN CREM 37 (12.3) 40 (13.3) 38 (12.7) 37.8 (12.6) GEOSN LIEG 38 (12.7) 43 (14.3) 40 (13.3) 37.6 (12.5) GEOSN OMED 35 (11.7) 48 (16) 48.5 (16.2) 44.6 (14.9) GEOSN OSEJ 39 (13) 49 (16.3) 44.2 (14.7) 39.2 (13.1) GEOSN PEND 35 (11.7) 49 (16.3) 46 (15.3) 47.5 (15.8) GEOSN PORT 39 (13) 48 (16) 43.7 (14.6) 47.6 (15.9)	AC Moho depth error ± (km) 2.18 1.89 1.88 2.23 1.96 2.38 2.38 1.91
GEOSN CREM 37 (12.3) 40 (13.3) 38 (12.7) 37.8 (12.6) GEOSN LIEG 38 (12.7) 43 (14.3) 40 (13.3) 37.6 (12.5) GEOSN OMED 35 (11.7) 48 (16) 48.5 (16.2) 44.6 (14.9) GEOSN OSEJ 39 (13) 49 (16.3) 44.2 (14.7) 39.2 (13.1) GEOSN PEND 35 (11.7) 49 (16.3) 46 (15.3) 47.5 (15.8)	1.89 1.88 2.23 1.96 2.38 2.38
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	2.38
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GEOSN VEGA 35 (11.7) 39 (13) 39.7 (13.2) 41.4 (13.8)	2.07
GEOCSN CAST 29 (9.7) 27 (9) 30.1 (10) 25 (8.3)	1.25
GEOCSN CONC 33 (11) 30 (10) 29.8 (9.9) 32.7 (10.9)	1.64
GEOCSN NIEV 33 (11) 36 (12) 36.8 (12.3) 33.3 (11.1)	1.67
GEOCSN OLES 35 (11.7) 46 (15.3) 46.5 (15.5) 39.8 (13.3)	1.99
GEOCSN LAGO 33 (11) 30 (10) 29.7 (9.9) 33.5 (11.2)	1.68
GEOCSN TOLI 35 (11.7) 34 (11.3) 32.1 (10.7) 36.6 (12.2)	1.83
GEOCSN URBI 33 (11) 38 (12.7) 39.7 (13.2) 37.8 (12.6)	1.89
GEOCSN SUAR 30 (10) 34 (11.3) 33.3 (11.1) 33.9 (11.3)	1.7
GEOCSN DEGA 34 (11.3) 35 (11.7) 33 (11) 32.1 (10.7)	1.61
GEOCSN ARAL 38 (12.7) 35 (11.7) 34.3 (11.4) 41.4 (13.8)	2.07
GEOCSN VALD 38 (12.7) 40 (13.3) 38.2 (12.7) 36.9 (12.3)	1.85
RSN EARI 36 (12) 49 (16.3) 49.4 (16.5) 48.3 (16.1)	2.42
RSN EPON 26 (8.7) 30 (10) 29.5 (9.8) 32.1 (10.7)	1.61
RSN ELAN 49 (16.3) 42 (14) 41.1 (13.7) 43.1 (14.4)	2.16
RSN EALK 44 (14.7) 30 (10) - 39.3 (13.1)	1.97
RSN EARA 41 (13.7) 46 (15.3) 42 (14) 47.2 (15.7)	2.36
IberArray E152 35 (11.7) 35 (11.7) 31.2 (10.4) 36.3 (12.1)	1.8
IberArray E153 33 (11) 38 (12.7) 39.7 (13.2) 36.2 (12.1)	1.82
IberArray E154 35 (11.7) 49 (16.3) 49.1 (16.4) 45.8 (15.3)	2.29
IberArray E155 40 (13.3) 42 (14) 41.2 (13.7) 38.7 (12.9)	1.94
IberArray E156 38 (12.7) 46 (15.3) 44 (14.7) 39.1 (13)	1.96
IberArray E157 42 (14) 39 (13) 50.6 (16.9) 38.1 (12.7)	1.91

IberArray	E141	38 (12.7)	35 (11.7)	33.9 (11.3)	40.5 (13.5)	2.03
IberArray	E142	34 (11.3)	38 (12.7)	36.4 (12.1)	38.2 (12.7)	1.91
IberArray	E143	31 (10.3)	40 (13.3)	37.7 (12.6)	41.4 (13.8)	2.07
IberArray	E144	37 (12.3)	38 (12.7)	34.5 (11.5)	37.2 (12.4)	1.86
IberArray	E145	38 (12.7)	40 (13.3)	35.7 (11.9)	41 (13.7)	2.05
IberArray	E146	40 (13.3)	43 (14.3)	41.8 (13.9)	44.5 (14.8)	2.23
IberArray	E147	43 (14.3)	45 (15)	42.1 (14)	44.7 (14.9)	2.24
IberArray	E125	32 (10.7)	38 (12.7)	33 (11)	29.5 (9.8)	1.48
IberArray	E127	34 (11.3)	38 (12.7)	33.9 (11.3)	34.2 (11.4)	1.71
IberArray	E128	37 (12.3)	33 (11)	36.3 (12.1)	31.5 (10.5)	1.58
IberArray	E129	39 (13)	40 (13.3)	40.6 (13.5)	36.9 (12.3)	1.85
IberArray	E131	33 (11)	37 (12.3)	-	31.8 (10.6)	1.59

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