Imaging the Deep Structures of Los Humeros Geothermal Field, Mexico, Using Three-Component Seismic Noise Beamforming

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Abstract

We present a 1D shear-velocity model for Los Humeros geothermal field (Mexico) obtained from three-component beamforming of ambient seismic noise, imaging for the first time the bottom of the sedimentary basement ~5 km below the volcanic caldera, as well as the brittle-ductile transition at ~10 km depth. Rayleigh-wave dispersion curves are extracted from ambient seismic noise measurements and inverted using a Markov chain Monte Carlo scheme. The resulting probability density function provides the shear-velocity distribution down to 15 km depth, hence, much deeper than other techniques applied in the area. In the upper 4 km, our model conforms to a profile from local seismicity analysis and matches geological structure inferred from well logs, which validates the methodology. Complementing information from well logs and outcrops at the near surface, discontinuities in the seismic profile can be linked to geological transitions allowing us to infer structural information of the deeper subsurface. By constraining the extent of rocks with brittle behavior and permeability conditions at greater depths, our results are of paramount importance for the future exploitation of the reservoir and provide a basis for the geological and thermodynamic modeling of active superhot geothermal systems, in general.

Introduction

Los Humeros volcanic complex (LHVC; Fig. 1), located in the eastern part of the Trans-Mexican volcanic belt (TMVB), hosts a conventional geothermal field (Ferrari et al., 2012; Gutiérrez-Negrín, 2019). On-going hydrothermal activity makes the LHVC a favorable area for geothermal exploitation, and a geothermal power plant has been operating since the 1990s. The LHVC has been identified as an important natural laboratory for the development of general models of superhot geothermal systems (SHGSs) in volcanic calderas (e.g., Jolie et al., 2018).

Although extensive geological field studies and well log analyses have provided many constraints on the near-surface geology of the caldera complex and conventional geothermal reservoir, conditions at depths greater than 2–3 km are largely unknown and currently being studied intensively (Jolie et al., 2018). It is assumed that superhot fluids could exist in the carbonate rock basement underlying the caldera (Jolie et al., 2018). These rocks might exhibit secondary permeability related to the damage zone of active resurgence faults and inherited pervasive basement structures (Lorenzo-Pulido, 2008; Rocha-López et al., 2010; Norini et al., 2015, 2019; Jolie et al., 2018). The maximum depth of these brittle structures is defined by the brittle-ductile (BD) transition zone, which thus plays an important role in geothermal exploration because upper crustal faults and fractures behave as hydraulic channels for the circulation of geothermal fluids (e.g., Ranalli and Rybach, 2005). In SHGSs that exhibit a positive thermal anomaly, the depth of the BD transition may differ from areas with a normal thermal gradient, as rocks become progressively more ductile with increasing temperature. Thus, a positive

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thermal anomaly could potentially limit the volume of rocks in which secondary permeability may exist.

We use three-component (3C) beamforming to extract structural information from ambient seismic noise. 3C beamforming is an array technique, which, like standard beamforming, not only estimates the dominant propagation direction and wavenumber of a recorded wavefield, but in addition determines the polarization of the wavefield by comparing phase shifts across different components (Riahi et al., 2013). As a result, different wave types can be distinguished and their propagation parameters analyzed separately. This allows us, for example, to estimate wavefield composition and surface-wave anisotropy, which is, however, beyond the scope of this study. Here, we consider fundamental mode Rayleigh waves only and extract dispersion curves from frequency–wavenumber (f–k) histograms; these are inverted for a shear-velocity depth profile using a reversible-jump Markov chain Monte Carlo (rj-McMC) algorithm. Although this algorithm is computationally expensive, it has the advantage of providing uncertainties for the velocity profile by finding the distribution of models that are consistent with data.

3C beamforming does not require impulsive (man-made or natural) seismic sources and is thus cheap, flexible, and applicable also in aseismic areas. Whereas cross-correlation-based ambient noise methods typically rely on month-long recordings, from beamforming, we extract stable dispersion curves from only 1 day of seismic noise data. Depending on the array geometry and seismic noise spectrum, the depth sensitivity of 3C beamforming can exceed that of other seismic methods by several kilometers, as we will show in this study. The analysis of four reflection seismic lines recorded across the LHVC, for example, provided 2D velocity maps and seismic sections down to 6 km at the most (Jousset, Ágústsson, et al., 2019). Ambient noise cross-correlation methods applied in the same area, but using a larger array, produce 3D tomographic...
images down to a maximum of 10 km depth (Granados Chavarria et al., 2020; Martins et al., 2020). In a similar manner, a recent local earthquake tomography study provides information only of the upper 3–4 km (Toledo et al., 2020). We show that 3C beamforming provides information to greater than 10 km depth.

In the following, we describe geology and available datasets, introduce both 3C beamforming and the rj-McMC inversion algorithm, and summarize our findings in Los Humeros and their implications for SHGSs, in general.

**Geology of LHVC**

The LHVC basement is composed of Mesozoic sedimentary rocks involved in the Late Cretaceous–Eocene compressive orogenic phase that generated the Mexican fold and thrust belt (sedimentary basement unit in Fig. 1) (Fitz-Diaz et al., 2017; references therein). The sedimentary basement rests above the Precambrian–Paleozoic crystalline basement of the Teziutlan Massif unit, made of greenschists, granodiorites, and granites (e.g., Suter, 1987; Suter et al., 1997; Ortuño-Arzate et al., 2003; Ángeles-Moreno, 2012; Fitz-Diaz et al., 2017) (Fig. 1a,b). Since the Eocene, the area underwent a limited extensional tectonic phase, associated with northeast-striking normal faults and the emplacement of Eocene–Miocene granite and granodiorite magmatic intrusions (Fig. 1a). The TMVB volcanic activity occurred from 10.5 to 1.55 Ma with the emplacement of basaltic andesite–rhyolite system of two nested calderas, namely the outer Los Humeros caldera and the inner Los Potreros caldera (Carrasco-Núñez, 2017; Carrasco-Núñez et al., 2018). Volcanic activity resumed ~700 ka ago with the emplacement of the Pleistocene–Holocene LHVC (LHVC unit in Fig. 1) (e.g., Carrasco-Núñez, Hernandez, et al., 2017; Carrasco-Núñez et al., 2018). This volcanic complex represents a basaltic andesite–rhyolite system of two nested calderas, namely the outer Los Humeros caldera and the inner Los Potreros caldera (Carrasco-Núñez, Hernandez, et al., 2017; Calcagno et al., 2018) (Fig. 1a). The LHVC caldera stage occurred between ~165 and ~69 ka and consisted of two major caldera-forming events that emplaced more than 100 km$^3$ of ignimbrite deposits (Yanez and Garcia, 1982; Ferriz and Mahood, 1984; López-Hernández, 1995; Cedillo-Rodríguez, 1997; Carrasco-Núñez, Hernandez, et al., 2017; Carrasco-Núñez et al., 2018). Widespread postcaldera monogenetic volcanic activity and resurgence of Los Potreros caldera floor occurred since 50 ka (Norini et al., 2015, 2019) (Fig. 1a).

The conventional geothermal field under exploitation is located in Los Potreros caldera, in the area deformed by resurgence faults (Fig. 1a), and is hosted by the Miocene–Pleistocene andesites, sealed by the LHVC ignimbrite deposits and/or the upper part of the pre-LHVC volcanic units (e.g., Cedillo-Rodríguez, 1997; Arellano et al., 2003; Norini et al., 2015, 2019; Carrasco-Núñez, Lopez-Martinez, et al., 2017). The permeability in the conventional reservoir is mainly secondary, generated by volcanotectonic faults and inherited tectonic structures (e.g., Cedillo-Rodríguez, 1997; Arellano et al., 2003; Norini et al., 2015, 2019). The maximum temperature of the hydrothermal fluids is around 400°C, measured at approximately 2.5 km depth (Arellano et al., 2003; Lorenzo-Pulido, 2008; Rocha-López et al., 2010).

**Data and Methods**

**Seismic array and data processing**

A multipurpose, temporary seismic array was installed across the LHVC, which recorded continuously from September 2017 to 2018 (see Data and Resources; Toledo et al., 2019). It consisted of 45 3C stations, 25 broadband and 20 short period stations, centered around previously located microseismic events within the inner caldera. For ambient noise beamforming, we use up to 17 stations of the dense broadband (DB) array (triangles in Fig. 1a) with a frequency sensitivity down to below 0.01 Hz and a sampling rate of 200 Hz. The analysis was restricted to 10 days in October and November 2017, when at least 14 stations were operating and dispersion curves had good quality. The corresponding array response function gives the beam response for a wave coming from directly below the array (with wavenumber $k = 0$) and is shown in Figure 2.

Data processing is carried out following Riahi et al. (2013) and Løet et al. (2018). Data are downsampled to 10 Hz, band-pass filtered between 0.01 and 1 Hz, and cleared from linear trends. We apply spectral whitening and one-bit normalization in the time domain to suppress large amplitude signals from

![Array response function](image-url)
earthquakes and to equalize amplitudes across different stations. Although time-domain normalization removes absolute amplitude information, we found that it significantly improved the extraction of Rayleigh-wave dispersion curves compared to other methods tested. Time series are divided into short windows, which are then beamformed separately. The length of a single time window corresponds to four times the minimum period, rounded up to the next power of two to speed up Fourier transformation. To this end, data are processed in three frequency bins, 0.08–0.155, 0.160–0.310, and 0.315–0.430 Hz, with corresponding time-window lengths of 51.2, 25.6, and 12.8 s, respectively. This way, the relative time-window length with respect to frequency is kept constant and, compared to a constant absolute time-window length, the number of superimposed waves, especially for larger frequencies, is reduced. Afterward, an average beam response of 10 consecutive time windows is computed.

**3C beamforming**

In standard, vertical-component beamforming, the horizontal wavenumber $k$, and the azimuth $\theta$ of a wavefield recorded at an array are estimated by analyzing phase shifts of the signal recorded at different stations within a small time and frequency window (Rost and Thomas, 2002). When 3C data are available, one can also analyze the phase shift between different components of each station and thereby estimate the orientation of particle motion of the dominant wave, hence its polarization (e.g., Riahi et al., 2013; Löer et al., 2018). In this case, the beam response is computed as a function of three parameters, namely the horizontal wavenumber $k$ and the azimuth $\theta$, combined in the horizontal wavenumber vector $k = \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix}$. 

(1)

and the so-called polarization state $\xi$, according to

$$R^{3C}(k, \xi) = w(k, \xi) \times S^{3C} \times w(k, \xi)^\dagger.$$  

(2)

All operations are performed in the frequency domain for a single frequency $\omega$; the frequency dependence is dropped in all equations for brevity. In equation (2), $S^{3C}$ represents the cross-spectral density matrix of 3C array data in the frequency domain, and

$$w(k, \xi) = c(\xi) \otimes a,$$  

(3)

is the Kronecker product of phase shifts $c(\xi)$ and $a(k)$, caused by polarization and wavenumber vector, respectively. Quantity

$$a(k) = \frac{1}{\sqrt{M}} \begin{bmatrix} \exp(i k \times r_1) \\ \vdots \\ \exp(i k \times r_M) \end{bmatrix},$$  

(4)

is an $M$-dimensional vector (Riahi et al., 2013), in which $[r_1, r_2, ..., r_M]$ denote the $M$ different station locations of the array. It is related to the array response vector

$$A(k) = \frac{1}{M} \sum_{m=1}^{M} \exp(i k \times r_m),$$  

(5)

known from standard beamforming. $c(\xi)$ is a complex $3 \times 1$ vector representing the polarization ellipse parameterized by $\xi = \xi(\theta, \phi, \epsilon, \gamma)$. The four polarization parameters denote azimuth ($\theta$), dip ($\phi$), ellipticity ($\epsilon$), and tilt ($\gamma$) of a wave’s particle motion. For more details, see Riahi et al. (2013) and Löer et al. (2018). The maximum of the beam response $R^{3C}(k, \xi)$ gives the parameter combination $(k, \theta, \xi)$ that matches the parameters of the actual wavefield best.

The resolvable wavenumber range is restricted by the geometry of the array (the minimum and maximum interstation distances $d_{\text{Min}}$ and $d_{\text{Max}}$, and the location of stations with respect to one another). A first estimate is obtained using an approximation for the resolvable wavelength $\lambda$,

$$2d_{\text{Min}} < \lambda < 3d_{\text{Max}},$$  

(6)

(Tokimatsu, 1997), and transferring it to wavenumber:

$$\frac{1}{3d_{\text{Max}}} < k < \frac{1}{2d_{\text{Min}}},$$  

(7)

For Los Humeros DB array with $d_{\text{Min}} = 1.6$ km and $d_{\text{Max}} = 13.9$ km, this results in $0.02 \text{ km}^{-1} < k < 0.31 \text{ km}^{-1}$. We note, however, that the wavenumber sensitivity is slightly azimuthally dependent due to the irregular aperture of the array, as can be seen in the array response function (Fig. 2).

Ambient noise tomography (ANT) based on cross correlations requires interstation distances larger than one typical wavelength, that is, $\lambda \ll d$ (e.g., Luo et al., 2015). That means, using the same seismic array, larger wavelengths and hence deeper structures can be investigated using ambient noise beamforming compared to ANT.

From horizontal wavenumber $k$ surface-wave phase velocity $c$ can be computed from $c = f/k$. Otherwise, by estimating minimum and maximum local phase velocities $c_{\text{Min}}$ and $c_{\text{Max}}$, wavenumber limits can be transferred to frequency limits, according to

$$c_{\text{Max}} \cdot k_{\text{Min}} < f < c_{\text{Min}} \cdot k_{\text{Max}}.$$  

(8)

Taking $c_{\text{Min}} = 1.5 \text{ km} \text{s}^{-1}$ and $c_{\text{Max}} = 4.5 \text{ km} \text{s}^{-1}$ and using the wavenumber limits from equation (7), we obtain a resolvable frequency range of $0.11 \text{ Hz} < f < 0.47 \text{ Hz}$ for surface waves.
In our analysis, we use discrete frequency steps of \( \Delta f = 0.005 \text{ Hz} \). The wavenumber is limited to \( k_{\text{Max}} = 0.3 \text{ km}^{-1} \) and sampled at discrete intervals of \( \Delta k = 0.0015 \text{ km}^{-1} \) in agreement with the resolution constraints of the array. The azimuth is sampled counterclockwise from east in 5° steps. The incidence angle varies between 0° (vertical) and 90° (horizontal) in 10° steps for body waves and is constant (90°) for surface waves.

**Dispersion curve retrieval and inversion**

In a medium in which velocities change with depth, surface waves are dispersive, that is, different frequencies propagate at different velocities, since they oscillate at different depths. Using ambient noise beamforming, we measure frequency-dependent Rayleigh-wave phase velocities, which are displayed as dispersion curves representing velocity as a function of frequency. Dispersion curves are extracted from 2D \( f-k \) histograms, showing how often a specific wavenumber was measured at a specific frequency for a given time window and polarization. Histograms of consecutive time windows are stacked up to 1 day. Figure 3a shows one such histogram for Rayleigh-wave polarization on day 300 in 2017. From the daily wavenumber histograms, we extract the maximum of the distribution at each frequency, which yields a curve \( k(f) \). Using \( c = f/k \), we then transfer it from the wavenumber to the phase velocity domain. Taking the average of multiple daily curves, we compute a mean curve and its standard deviation (Fig. 3b), downsampled to \( \Delta f = 0.2 \text{ Hz} \) for a faster application of the inversion scheme.

The average dispersion curve is inverted using a Bayesian inversion scheme. Bayesian inversions are performed using the method of rj-McMC (Green, 1995; Bodin et al., 2012). These methods update a prior probability distribution that describes information about model parameters that is independent of the current data, with the new information added by that data. This process is known as Bayesian inference, and the result is a posterior distribution over parameters. The rj-McMC algorithm performs the update while allowing the number and depth extent of seismic velocity parameters of horizontally layered media (and hence the dimensionality of parameter space) to vary in the inversion. The parameterization is adapted to only the complexity required by the combination of prior information and data, which improves results on otherwise high-dimensional nonlinear inverse problems. The resulting probability density function (PDF) provides the distribution of models that are consistent with the data.

Prior information about seismic shear velocities is assigned to be a uniform distribution between 500 m s\(^{-1}\) < \( v_S \) < 7500 m s\(^{-1}\), and the number of layers in depth is allowed to vary between 5 and 25, with the maximum depth constrained to 30 km. Each run of rj-McMC then generates a chain of two million samples (example velocity models), which are distributed according to the posterior distribution as the number tends to infinity. The rj-McMC inversion scheme is run multiple times (here, 12 chains in total), each starting from a different random model. To obtain approximately independent samples, we retain only every 250th sample after the first 0.5 million samples (the “burn-in” phase) in each chain for analysis. The density of the final set of retained samples represents the PDF of shear velocity with depth.

**Results**

Figure 4a shows the PDF of shear velocity as a function of depth, in which the zero level equals 2.9 km above sea level, that is, the average topographic height of the seismic stations considered. Dimensions are 500 × 100 grid points, with a grid spacing of \( \Delta z = 62 \text{ m} \) in \( z \) direction and about \( \Delta v_S = 71 \text{ m/s} \) in \( x \) direction. For each grid point, the number of samples indicates the number of models that assign a certain shear velocity to the respective depth. The color scale has been normalized with respect to the maximum number of samples at each depth level. Light colors indicate velocities of higher probability compared to dark colors. The solid white curve denotes the maximum likelihood of the PDF, the dashed curve is the mean of
the PDF, and the dotted white curve shows the shear-velocity profile obtained from earthquake analysis for comparison (T. Toledo, personal comm., 1 June 2020). The latter was estimated using the code Velest (Kissling et al., 1994) for joint inversion for P- and S-wave models using travel-time data from 333 local seismic events.

In Figure 4b, we display a shear-velocity profile that combines the results from both methods: down to 3.2 km, it is based on earthquake data (dotted curve in Fig. 4a); below 3.2 km, it follows ambient noise beamforming (solid curve in panel (a), below 3.2 km); background colors indicate geological structure (see legend) as in Figure 1b. Transition depths are derived from well data for the two upper sections (see Table S1) and from the shear-velocity PDF for the deeper structures. Dashed and dotted horizontal lines indicate the range of transition depths found in well data. BD, brittle-ductile. The color version of this figure is available only in the electronic edition.

Figure 4b). Transition depths observed in well logs at 0.9 and 2.2 km, respectively (see Table S1 and Norini et al., 2019). Deeper section transitions could not be observed in well logs and are thus derived from discontinuities in the shear-velocity profile.

At 5.1 km depth, shear velocities increase abruptly from around 2.6 to 3.3 km s$^{-1}$, implying the transition from the sedimentary to the crystalline basement (the Teziutlan massif), which so far has only been mapped at the surface west-northwest of the central caldera (Fig. 1a). Laboratory values for S-wave velocities at respective pressure conditions confirm a range between 2.5 and 3.1 km s$^{-1}$ for limestones and between 3.2 and 3.6 km s$^{-1}$ for granites and granodiorites (Gebrande, 1982). Bär and Weydt (2019) performed ultrasonic pulse velocity measurements at rock samples from Los Humeros geothermal wells, reporting S-wave velocities between 2.2 and 3.6 km s$^{-1}$ at an average of 3.4 km s$^{-1}$ for limestones at surface temperature. With increasing temperature, they show that S-wave velocities decrease to about 1.5 km s$^{-1}$ at 400°C.

At about 7 km depth, the velocity profile shows another sudden increase of shear velocity, which could indicate a geological boundary within the generally heterogeneous Teziutlan massive (Yanez and Garcia, 1982; Ángeles-Moreno, 2012) or an intrusive body never identified before. We note, however, that the PDF exhibits a relatively broad velocity distribution at and below that depth and that also the mean curve implies a considerable uncertainty with respect to depth and magnitude of the discontinuity. Thus, a more gradual velocity change related to increasing pressure conditions seems equally likely.
The decline in shear velocity observed at about 10.5 km depth is interpreted as the onset of the BD transition, which correlates with a decrease in shear strength (e.g., Imber et al., 2008). This transition marks the maximum depth for the occurrence of brittle structures that could facilitate secondary permeability.

**Discussion and Conclusions**

Analyzing ambient seismic noise data, we have imaged for the first time the bottom depth of the sedimentary basement and the BD transition zone in the area of Los Humeros geothermal field. We used 3C beamforming to retrieve Rayleigh-wave dispersion curves from noise data, which were then inverted to give a shear-velocity profile using an rj-McMC algorithm. The profile complements results from earthquake-based methods, which are in good agreement with geological data obtained from well logs; however, these methods provide information only down to a maximum of ~4 km. The depth sensitivity of our method extends down to 15 km, covering the transition from the sedimentary to the crystalline basement in 5 km depth and the BD transition zone around 10 km depth.

Generally, rocks become progressively more ductile with depth because of the increasing temperature, until the BD transition is reached (Ranalli and Rybach, 2005). Our results demonstrate for the first time in the LHVC area that even if the geothermal gradient is higher than for normal crust (e.g., Arellano et al., 2003; Lorenzo-Pulido, 2008; Rocha-López et al., 2010), and if alteration of the rocks hosting the hydrothermal fluids occurs with an expected reduction of the total rock strength (e.g., Arzate et al., 2018; Norini et al., 2019), both sedimentary succession (sedimentary basement unit in Figs. 1 and 4) and upper part of the underlying crystalline basement (Teziutlán Massif unit in Figs. 1 and 4) exhibit brittle rheology down to 10 km below the topographic surface. The 10 km deep BD transition acts as the root zone of brittle structures responsible for secondary permeability in the geothermal reservoir and potentially hosting superhot fluids. The rheological zonation identified at 10 km below surface should thus be included in any geological model of the caldera complex and hydrothermal system, as well as in heat flow and heat transfer modeling (e.g., Calcagno et al., 2019).

One way to verify our results would be to test them against receiver functions, which are generally used to investigate the near-surface structure using body-wave energy from distant sources that have been refracted and converted at layers beneath a seismometer (Phinney, 1964; Vinnik, 1977). Galetti and Curtis (2012) showed that the theories of receiver functions and seismic interferometry are closely connected, and that the former could also be computed from ambient noise sources and across different stations. To show this, however, is beyond the scope of this work and will be subject to future research.

We provide an estimate of the BD transition depth at 10 km that is independent from local seismicity, since our shear-velocity profile has been estimated from ambient noise. Our results confirm that declining seismicity below 3.2 km depth is not caused by the transition from brittle to ductile rheology; instead, it is related to the maximum exploitation depth of the geothermal field. Seismicity studies at LHVC prior to exploitation activities have been minimum (Ponce and Rodriguez, 1977), with most studies being accomplished during the exploitation phase (e.g., Lermo et al., 2008). These studies along with a more recent survey (Gaucher et al., 2019; Jousset, Toledo, et al., 2019; Toledo et al., 2020) show minimum local seismicity mostly related to the geothermal field activities. In fact, earthquakes at LHVC occur mostly in clusters located close to injection wells. Some of these events have depths larger than the injection wells (~2.5 km depth) and are located in the limestones belonging to the sedimentary basement unit, suggesting possible local fracturing of this layer due to pressure changes caused by injection and production activities. The heterogeneous focal mechanisms (mostly strike slip with left- and right-lateral motion) (Lermo et al., 2008) and the temporal relation of the seismicity rate with injection rates (Lermo et al., 2008; Jousset, Toledo, et al., 2019) seem to confirm that parts of the local seismicity is induced by local stresses due to injection. Another part of the seismicity is of tectonic origin.

The presented shear-velocity model may also help to constrain boundary conditions for other methods. For example, it can be used as a starting model for 3D seismic tomography (Gaucher et al., 2019; Toledo et al., 2020), in earthquake localization with time-reverse imaging (Werner and Saenger, 2018) or for assessing the correlation between rock moduli and temperature (Mendrinos et al., 2019).

Overall, we demonstrate that 3C beamforming of ambient noise combined with dispersion curve inversion using the rj-McMC algorithm provides information on deeper structures of SHGSs that is typically not found by standard methods such as well log or seismicity analysis. Based on ambient seismic noise, the method is also applicable in aseismic regions where other methods are altogether infeasible. It complements ambient noise cross-correlation-based techniques, which, while providing 3D tomographic images, require a significantly larger array aperture to reach the same depth sensitivity and are computationally much more expensive. 3C beamforming provides a quick first estimate of deep geological structures from a 1D seismic profile, which can also serve as initial model for other methods. Our findings in the LHVC suggest that SHGSs can exhibit brittle rheology at large depths, despite positive thermal anomalies, increasing the rock volume available for geothermal exploitation.

**Data and Resources**

Waveform data and associated metadata are available from the GEOFON data center under network code 6G (https://geofon.gfz-potsdam.de/doi/network/6G/2017, last accessed March 2019) and...
are embargoed until January 2023. The Comisión Federal de Electricidad (CFE) of Mexico kindly provided logs of geothermal wells and access to the geothermal concession area. Cartographic data are stored in a personal geodatabase, and the final map of Figure 1 was produced in ESRI ArcMap 10.3. The supplemental material for this article includes a figure showing the depth sensitivity kernels of Rayleigh waves compared to the depth distribution of local seismicity (Fig. S1) and a table listing geothermal wells and transition depths of geological units identified in the corresponding well logs (Table S1).

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References


