Lower-Crustal Seismicity on the Eastern Border Faults of the Main Ethiopian Rift

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Abstract

Lower-crustal seismicity is commonly observed in continental rift zones despite the crust at such depths being ductile enough to prohibit brittle failure. The source of such deep seismicity across the East African Rift remains an outstanding question. Here we present analysis of an isolated cluster of lower-crustal earthquakes located on the eastern border faults of the Main Ethiopian Rift, near the Corbetti caldera. Lower-crustal earthquakes have not previously been observed in this area. Phase arrival times were determined using an automated picking approach based on continuous wavelet transform and statistical changepoint detection methods. We overcome misinterpretations from large hypocenter depth errors by considering mixture distributions for all events and their associated uncertainties. These mixture distributions represent probability density functions of any event occurring at a given depth. The mixture distribution mode for a variety of different velocity models and error parameters remained stable at a depth of 28–32 km, with the vast majority of maximum likelihood estimates for individual hypocenters located at depths of 25–35 km. Most events occur over a 2 month period, with 90% of cumulative seismic moment occurring during March and April 2012. The ephemeral and localized nature of this seismicity, combined with low event magnitudes and regional hydrothermal/magmatic activity, suggests that these lower-crustal events are likely related to fluid or magmatic processes. Plausible mechanisms include the movement of magma and/or exsolution of volatiles at depth causing transient high strain rates and pore fluid pressures that induce seismicity.

1. Introduction

Seismicity associated with continental rift zones can reach depths of more than 30 km (e.g., Albaric et al., 2014; Keir et al., 2009; Reyners et al., 2007; Yang & Chen, 2010) despite the general conception that the crust at such depths is too ductile for brittle failure to occur (Chen & Molnar, 1983). In the East African Rift (EAR; Figure 1), deep crustal, and even upper mantle, seismicity has been observed along the younger eastern and western branches of the EAR (Albaric et al., 2014; Lavayssière, Droof, et al., 2019; Lindenfeld & Rümpker, 2011; Yang & Chen, 2010), as well as beneath the western margins of the more well-developed, magmatic Main Ethiopian Rift (MER) at the northern end of the of the EAR (Keir et al., 2009; Figure 1). One explanation is that these earthquakes are caused by slip on border faults into relatively cold or strengthened lower crust (Doser & Yarwood, 1994; Jackson & Blenkinsop, 1993; Zhao et al., 1997). Such an explanation requires the underlying mantle to be cool and/or strong enough for earthquakes to occur, for which there may be compelling evidence beneath less magmatic segments on the western branch of the EAR (Lindenfeld & Rümpker, 2011; Yang & Chen, 2010). However, beneath the considerably more magmatic MER, where the lithospheric mantle has high enough temperatures to sustain partial melt (Hammond et al., 2014; Kendall et al., 2005), the lower crust and upper mantle are deemed too ductile for brittle failure; here deep seismicity is thought to be induced by magmatic processes and fluids (Keir et al., 2009; Reyners et al., 2007; Soosalu et al., 2010). In other oceanic and continental rift settings, such as Iceland and the Taupo rift in New Zealand, deep seismicity has been associated with high strain rates/pore pressures (e.g., Greenfield & White, 2015; Soosalu et al., 2010; White et al., 2011), magma/fluid emplacement (Smith et al., 2016), and the weakening of border faults (e.g., Reyners et al., 2007) arising from melt movement and/or the exsolution of volatiles. However, the mechanism for deep seismicity in the EAR, where the level of magmatism varies from section to section, is still debated (e.g., Weinstein et al., 2017).
In this paper, we examine an isolated cluster of lower-crustal seismicity approximately 30 km east of the Corbetti caldera on the eastern border faults of the MER, where no previous lower-crustal seismicity has been observed. The distribution mode of this seismicity lies at a depth of 28–32 km, directly beneath the Wondo Genet scarp (Figure 2). This scarp is thought to relate to an intersection between the rift border and the easternmost end of a cross-cutting fault structure that predates the rift (Lloyd, Biggs, Wilks, et al., 2018; Figure 2), with the interaction of these two faults potentially influencing seismicity in the area. We determine event phase arrival times with high precision using wavelet transform and statistical changepoint detection methods. We attempt to reduce the variability of individual event locations, largely caused by poor seismic array coverage relative to the region of seismicity, by considering the mixture distribution of all events observed for a given velocity model and set of location algorithm parameters. The temporal and spatial distribution of these events suggests that these isolated events may be caused by a single, ephemeral intrusion at lower-crustal depths that causes overpressure, inducing seismicity, or hot fluids reducing the effective normal stress on the faults.

2. The MER and Corbetti Caldera

The MER is an area of continental rifting that forms the major divergent plate boundary between the Nubian and Somalian tectonic plates at the northern end of the EAR (Figure 1). It is thought to have initiated and developed asynchronously along its length (e.g., Wolfenden et al., 2004), with the development of different sectors influencing magmatism, strain, and crustal thickness across the region (Keir et al., 2015; Muluneh et al., 2017). Most regional seismicity occurs along the axis of the MER (Wilks, 2016), where the seismogenic layer is constrained to upper crustal depths (< ~16 km; Muluneh et al., 2018). However, earthquakes as deep as ~35 km (the approximate thickness of the crust in this region; Dugda et al., 2005; Ebinger et al., 2017; Stuart et al., 2006) have been identified on the western margin of the MER, associated with areas of recent volcanism along the Debre-Zeit and Yerer-Tullu Wel Wel Volcanotectonic Lineaments near Addis Ababa (Keir et al., 2009). Additional lower-crustal seismicity at depths of ~22 km, which is still below the brittle-ductile transition (BDT) zone for the region (16 km depth; Muluneh et al., 2018), has been observed beneath the rift-adjacent northwestern (NW) Ethiopian Plateau (Keir et al., 2009), north of Addis Ababa. The location of this seismicity in volcanic areas supports the idea of magma emplacement below/into the lower crust as a source of lower-crustal seismicity in the MER (e.g., Keir et al., 2009; Soosalu et al., 2010), rather than fault slip within a relatively cold or strengthened lower crust.

The Corbetti caldera is the southernmost silicic center along the rift. It lies toward the eastern side of the MER and formed at 182 ± 18 ka (Lloyd, Biggs, Wilks, et al., 2018). Its magmatic and hydrothermal processes are thought to be influenced by a cross-cutting fault structure that predates the rift and may extend as far as the rift border, ~30 km to the east (Lloyd, Biggs, Wilks, et al., 2018). Seismicity beneath the caldera appears to be constrained to the uppermost crust (Lavayssière, Greenfield, et al., 2019; Lloyd, Biggs, Birhanu, et al., 2018), although previous seismic monitoring in the area has been limited.

Prior to the events analyzed in this paper, no other sequence of lower-crustal seismicity had been observed beneath the rift itself, the Corbetti caldera, the eastern border faults of the MER, or the seismically quiet southeastern (SE) Somalian plateau to the east.

3. Seismic Array Geometry

Two local seismic arrays were deployed at Aluto (Wilks et al., 2017) and Corbetti (Wilks, 2016) volcanoes throughout 2012 and 2013, with 12 and 7 broadband seismometers, respectively, deployed at each location.
volcano, although only subsets of around 15 of these instruments were operational at any given time throughout the study period (Figure 2). The relative location of these two local arrays to the observed seismicity leaves a substantial array gap (~270°), with no other regional seismic networks operating at the time.

4. Phase Arrival Picking

Phase arrival times were identified manually across the full deployment (>3,000 events over 2 years continuous data; Wilks, 2016). The anomalous subset of apparent lower-crustal seismicity was identified by subsequent event location (number of events = 134). Initial phase arrival picks for this subset had large pick errors and large predicted traveltimes (TT) residuals, sometimes on the order of several seconds; for this reason, the first step in this study was to reduce this source of error in hypocenter estimation through a time-frequency-based automated picking approach (Figures 3 and 4). This approach was based on continuous wavelet transform (CWT) spectral analysis (e.g., Lapins et al., 2020) and statistical changepoint detection (Fryzlewicz, 2014).

P-wave arrivals were characterized by producing CWT scalograms between 2 and 12 Hz (e.g., Lapins et al., 2020; Figure 3b) for the raw vertical component traces at each station (Figure 3a). We then determined the average wavelet energy between these frequency bounds for each time sample point (black line in Figure 3c) using both low- and standard-frequency Morlet wavelets (Morlet central frequency parameter $\omega_0 = 1$ and $\omega_0 = 6$, respectively). Changes in average wavelet energy with time were detected using wild binary segmentation (WBS; Fryzlewicz, 2014), an a posteriori changepoint detection method that segments the signal by a given statistical property, which in this case is mean energy. The WBS “threshold” parameter for determining a change in mean energy was set very low (threshold constant $c = 0.5$), which led to a near-uniform rate of “false positive” changepoint detections during pre-event noise (one to two detections per second) and a much higher rate of changepoint detections at the P-wave arrival and during the coda ($\geq$1 detection per second; vertical red lines in Figure 3c). The cumulative number of detected changepoints within a signal window (90 s) was then used to identify P-wave arrival through second-order differencing (lag = 4 s) and a simple trigger algorithm (Figure 3d).

S-wave arrivals were identified in a similar manner to P-wave arrivals but using the cross-wavelet transform (XWT) derived from the two horizontal component CWT scalograms at each station (Figure 4). This had the effect of reducing incoherent background noise while enhancing coherent signal across the two horizontal components (Figure 4b). Again, the average wavelet energy was determined at each time sample point for low- and standard-frequency Morlet wavelets. Finally, WBS and second-order differencing was used to identify S-wave arrivals common to both horizontal components (Figures 4c and d), this time with a greater WBS threshold constant ($c = 250$ for Aluto stations and $c = 5,000$ for nearby Corbetti stations) to avoid a higher rate of changepoint detections around P-wave arrivals. We use different threshold constants across the two arrays due to higher amplitude values in the raw signal at the nearer Corbetti stations.

All P and S wave picks were manually checked against raw and filtered traces, CWT scalogram images, and additional STA/LTA (ratio of short-term average to long-term average amplitude; e.g., Withers et al., 1998) traces to assess quality and uncertainty. Arrival times for highly emergent signals were sometimes picked late by the CWT-WBS approach outlined above, likely a consequence of the lower-frequency onset of these signals, which is outside of the frequency range encompassed by the CWT scalograms (2–12 Hz). These arrival times were adjusted manually using visual determination on wider frequency band CWT scalograms and raw traces. Arrival time picks that were difficult to confirm manually (e.g., in the presence of very low signal-to-noise ratio [SNR]) were removed.

Picks made using the low-frequency Morlet wavelet ($\omega_0 = 1$), which yields improved time resolution at the cost of frequency resolution (Addison et al., 2002; Lapins et al., 2020), were more accurate for signals with medium to high SNR, while picks based on the standard Morlet wavelet ($\omega_0 = 6$) were more accurate for...
very noisy signals because of greater frequency localization of pertinent signal features among ambient noise (Lapins et al., 2020).

5. Absolute Event Locations

Events were located using NonLinLoc (Lomax et al., 2000) and three different regional velocity models (Daly et al., 2008; Mackenzie et al., 2005; details in Figure 5 caption); TT error was varied to assess the stability of event locations and determine potential bias in absolute hypocenter locations (e.g., artificially deep locations). Only events with a total of 7 or more phase arrival times, including at least one S phase, were located (58 events in total). We include at least one S phase to improve spatial constraints on hypocenter locations (e.g., Lomax et al., 2009), which can have large uncertainties when using stations with a large array gap or that are far from the event.

The significant array gap in this case (approximately 270°) meant that individual event locations did indeed have large error distributions, although deeper events were generally better constrained than shallower events (Figure 6). For this reason, we also examine mixture distributions (a weighted combination of probability distributions for each individual event location) for all events, with a given set of location parameters (Figure 7), to determine stability and likelihood of event depth estimates using the whole population of events in the study. More formally, a (finite) mixture distribution, $f(x)$, is a convex combination of $n$ component probability distributions, $p_1(x), \ldots, p_n(x)$,

$$f(x) = \sum_{i=1}^{n} w_i p_i(x), \quad (1)$$

with weights $w_1, \ldots, w_n$ such that $w_i \geq 0$ and $\sum w_i = 1$. Here, the component distributions, $p_i(x)$, are the individual posterior probability density functions (PDFs) for each event’s hypocenter location, estimated by NonLinLoc, where the vector $x$ represents the three-dimensional spatial $x, y, z$ hypocenter location and $i$ is event index 1, $\ldots$, $n$. A mixture distribution preserves the required properties of probability distributions (nonnegativity and integrating to 1) and is therefore itself a probability distribution.

We use mixture distributions to estimate the overall population distribution for all event locations and assess the likelihood of events occurring at a given depth (from here on we use the terms overall population distribution and mixture distribution interchangeably). In practice, probability distributions for individual
Earthquake locations are nonlinear and may be mathematically intractable. For this reason, earthquake location distributions are often estimated through probabilistic sampling of their complete posterior distribution (e.g., Gesret et al., 2015; Lomax et al., 2000). The number of samples used for estimating each component distribution is determined by its corresponding weight with regard to the mixture distribution (Equation 1). As each component distribution in our study represents a single event location, all with equal weight, all component distributions are weighted equally, with the same number of samples drawn from each event location PDF in NonLinLoc ($s_i = 5,000$, where $s_i$ is the number of samples drawn for event index $i = 1, \ldots, n$).

Residuals between observed and expected TTs were greatly improved over initial manual picking through the CWT-WBS picking approach outlined in section 4 (original manual-picking mean absolute TT residuals = 0.89 s; CWT-WBS mean absolute TT residuals = 0.23 s). However, reducing pick error did not markedly reduce the size of individual event location uncertainties, and, in some cases, using a smaller number of higher-confidence phase arrival times produced greater hypocenter location uncertainty (in terms of spatial spread) than using a larger number of lower-confidence arrival times. This suggests that arrival time pick error has a smaller contribution to absolute location uncertainty in NonLinLoc than array geometry and estimates of TT and velocity model error (Lomax et al., 2009).

Despite the large uncertainty in individual event locations, the overall population (mixture) distribution for all event locations produces a clear, stable mode between 28 and 32 km depth for all velocity models used for locating events ($V_p = P$ wave velocity; $V_s = S$ wave velocity). Model 1 is from Table 1 in Daly et al. (2008); Model 2 is approximated from Figure 5 in Daly et al. (2008); Model 3 is approximated from Figure 6 in Mackenzie et al. (2005). $V_p/V_s$ ratio of 1.75 (average from Table 1 in Daly et al., 2008) used for Models 2 and 3 as only $V_p$ given in original publications.
models and model error parameters (shaded density curves in Figure 7). Furthermore, the vast majority of event hypocenters are located between depths of 25 and 35 km (histograms in Figure 7) for any given level of Gaussian TT error. This suggests that most events are indeed at lower-crustal depths, although this becomes understandably less clear with very large TT error levels (e.g., ≥10% TT error; Figure 7 bottom). The maximum likelihood estimates (MLEs) for hypocenter locations across all model runs also fall into two distinct clusters: a small, shallower cluster above the BDT zone (Muluneh et al., 2018) and a larger, deeper cluster between depths of 20 and 35 km (Figure 7), with very few events located around the BDT zone itself.

6. Further Source Characteristics

There is some evidence of repeating event source(s) from high interevent cross-correlation (CC) values (e.g., Augliera et al., 1995) at Station C02E, the closest station to the study area, with 47 out of a total of 58 P wave arrivals having CC values >0.7 with at least one other event (4 s window around P wave arrival; at least four distinct multiplet groups identified). However, this is not seen at other stations across the two volcanic arrays, where interevent CC values are consistently low due to low SNR, particularly across the Aluto array (on average, <4 out of 58 events have CC value >0.7 with at least one other event at a given station). As such, it is difficult to determine whether events have similar source mechanisms or exploit any self-similarity in our analyses of event locations.

Most events occur over a 2 month period (Figure 8), with 45 of 58 events and 90% of cumulative seismic moment ($M_0$) occurring during March and April 2012. $M_0$ and moment magnitudes ($MW$) were calculated for each event at each station using spectral analysis of both P and S wave arrivals, with noise spectra subtracted, and Brune source model fitting (Abercrombie, 1995; Prejean & Ellsworth, 2001; Wilks, 2016). The values of $M_0$ and $MW$ were then averaged across all stations to attain a final, single value of $M_0$ and $MW$ for each event. To verify the quality of these predicted magnitude values, local magnitudes ($ML$) were also calculated, using a scale calibrated for the MER (Keir et al., 2006), with station corrections (determined by the average deviation of magnitudes measured at a given station) applied to account for the variability in the recording environments from station to station. Estimates of $ML$ and $MW$ were found to be in close alignment, with the 2.5% and 97.5% quantiles of $ML - MW$ equal to −0.28 and 0.29, respectively. First motion polarities, where identified, for all events within the main cluster (January–June 2012) show a consistent trend: First motions across one array were consistently opposite to those at the other. During this period, first motion was predominantly downward (dilation) across the Corbetti stations and upward (compression) across the Aluto stations, although these polarities reversed to their opposite sign at least three
occasions. However, the final four isolated events (between August 2012 and May 2013) had distinctly different behavior: The polarities were the same across both arrays, with two events having compression first motion across all stations and the other two events having dilation first motion across all stations. This overall behavior is interesting, particularly as these events are clustered in time, as it suggests they may represent different sources or a complex pattern from a single process.

Figure 7. Event locations and mixture distributions for varying levels of traveltime error (1% top, 5% middle, 10% bottom) using velocity Model 1 from Figure 5. All velocity models produced similar results. (left) Individual event locations using NonLinLoc (blue circles). Black triangles are seismic stations. Histograms show the number of events located within a 0.05° bin (latitude and longitude) or 5 km bin (depth). (right) Corresponding mixture distributions for all event locations (and estimates of their complete posterior PDFs). Red scatter and corresponding density curves represent mixture distributions of NonLinLoc event location PDFs with respect to a given plane (latitude, longitude, and depth).
7. Discussion

Mixture distributions were used to overcome potential misinterpretations arising from large depth uncertainties in individual hypocenter locations and to assess the overall likelihood of lower-crustal events. The overall mixture distribution mode and majority of MLEs for hypocenter locations lie between 25 and 35 km depth for all velocity models and levels of TT error (Figure 7). These results indicate that earthquakes beneath the magmatic MER and its border faults likely occur at lower-crustal depths and far below the generally recognized seismogenic zone along the rift (Keir et al., 2009; Muluneh et al., 2018; Yang & Chen, 2010).

Our MLE hypocenter locations suggest a possible bimodal distribution of event depths (above 15 km and below 20 km), which is consistent with depth distributions previously observed near the MER (Keir et al., 2009) and in less magmatic sections of the EAR (Yang & Chen, 2010). The ephemeral and very localized nature of this seismicity (Figures 7 and 8) combined with low event magnitudes (range: 1.9–3.6 $M_W$; median: 2.5 $M_W$), the magmatic setting associated with the Corbetti volcano and MER, and the adjacent hot springs around the Wondo Genet scarp at the surface all suggest that these lower-crustal events are likely related to fluid or magmatic processes (Keir et al., 2009; Yang & Chen, 2010) rather than slip on cold or modified crust.

Seismicity in the area around Wondo Genet, where our mixture location distribution mode lies, has not been observed in previous (e.g., Maguire et al., 2003) or subsequent (e.g., Lavayssière, Greenfield, et al., 2019) seismic deployments, with the latter study deploying a broadband seismometer directly above the area of seismicity identified in this paper. This lack of subsequent seismicity supports the interpretation that these events were, in fact, related to a single, ephemeral intrusion or transient exsolution/migration of volatiles, rather than ongoing volcanic or shallow hydrothermal activity associated with the Corbetti caldera or Wondo Genet hot springs. A reasonable interpretation from the identified pattern of first motion polarities across the two arrays (section 6) could be a stable source mechanism during the main period of an intrusion (i.e., the main cluster of events during January–April 2012) followed by a more complex process following the intrusion event. Alternatively, the transient presence of hot fluids may have increased pore pressure or reduced the effective normal stress on the border and cross-rift faults at this intersection.

Figure 8. (bottom) Cumulative moment for all located events (orange line) and histogram of number of events per month (yellow). Ninety percent of total moment occurred in March/April 2012. Largest three events and moment size indicated in blue. (top) Corresponding event depths (gray and blue circles) from Figure 7, TT (traveltime) error = 0.01. Moho depth of 38 km approximated from Dugda et al. (2005) and Stuart et al. (2006).
Unfortunately, additional assessments of source mechanism (e.g., focal mechanism determination/full waveform inversion) and relative event locations (e.g., double-differencing and coda wave interferometry) all yielded poor or unstable solutions due to the low number of events, large array gap, low number of picks/stations, unknown velocity model error, low SNR, and low cross-correlation values. As such, assessment of source must come from temporal and spatial characteristics combined with plausible physical mechanisms within the regional setting.

The weakness of our analysis lies in the relative location of instruments available during the study period, which yields large depth errors for individual hypocenter locations regardless of velocity model or error estimates (Lomax et al., 2009). As such, it is impossible to state whether any of these events lie below the Moho, assumed to be at 37–40 km depth for the southern MER (Ayale et al., 2004; Dugda et al., 2005; Stuart et al., 2006), or whether they are all constrained within the lower crust. Stable overall population (mixture) distributions, however, reveal a clear mode between 28 and 32 km depth, regardless of velocity model or parameter adjustments, and thus strongly suggest that at least some of these events are deep. While the highly heterogeneous composition beneath the rift and these volcanic centers makes it difficult, or even impossible, to know which level of error is most appropriate, an error level of no more than 10% of TT seems suitable given the error bounds published for one of the regional velocity models used in this study (approximately 1.6% difference in total TT; Daly et al., 2008) and the maximum absolute difference in TTs between all models used (approximately 8%).

One way in which our hypocenter location estimates could be placed artificially deep is through use of a velocity model that is slower than the true Earth velocity structure (e.g., Poliannikov & Malcolm, 2016). Where the velocity model used is incorrect, the direct-search approach of NonLinLoc provides a better estimate of location hypocenter than linearized methods, as negative and positive TT residuals need not be balanced to produce a complete posterior probability distribution. By contrast, linearized approaches produce a single-point estimate, often with Gaussian errors subsequently calculated (Gesret et al., 2015). It is possible to use approaches which jointly infer the velocity structure and event origin parameters to obtain PDFs of the hypocenter with the uncertainty of the velocity included (e.g., Piana Agostinetti et al., 2015), but these require either good coverage of the volume being imaged or well-constrained prior information on velocities to tightly locate events, which is not available here. Regardless, every care has been taken to use velocity models representative of the region through use of several previously published models (Daly et al., 2008; Mackenzie et al., 2005) and a range of TT error levels that are consistent with other published estimates of velocity structure (e.g., Keranen et al., 2009). Furthermore, the previously identified magmatic-hydrothermal activity beneath Aluto and Corbetti volcanoes (Lloyd, Biggs, Birhanu, et al., 2018; Wilks et al., 2017) would suggest that the true Earth velocity structure along the ray path to these volcanic centers would more likely be slower, rather than faster, than our models, which do not incorporate any adjustments for these volcanic centers.

The source of lower-crustal seismicity and processes by which magmatism evolves within the crust in continental rift zones remains an outstanding question (e.g., Weinstein et al., 2017). The 2012–2013 seismicity east of Corbetti volcano appears to lie at a potential intersection between the rift border and a preexisting cross-rift structure beneath the Corbetti caldera (Lloyd, Biggs, Wilks, et al., 2018). However, the apparent NW-SE linear/listric distribution of events away from Corbetti is almost certainly an artifact of array geometry (Lomax et al., 2009), with the vast majority of individual location PDFs from NonLinLoc marking out this cross-rift “trend,” so it is impossible to say whether this seismicity relates to or indicates the easternmost extent of this cross-rift structure. Event locations near Corbetti do, however, fit with observations of lower-crustal seismicity in other regions of recent volcanism, both around the MER (Keir et al., 2009) and other volcanic centers (e.g., Neuberg et al., 2006; Soosalu et al., 2010). The distance of the overall distribution mode of seismicity at ~30 km from the Corbetti caldera suggests that the source of these events is exploiting a potential point of weakness along the rift border fault or cross-rift structure rather than being directly related to the magmatic storage processes beneath Corbetti (Lloyd, Biggs, Birhanu, et al., 2018; Lloyd, Biggs, Wilks, et al., 2018). Plausible mechanisms, given the temporal and spatial distribution of this seismicity, include the movement of magma and/or exsolution of volatiles causing transient high strain rates and pore fluid pressures that induce seismicity (e.g., Greenfield & White, 2015; Keir et al., 2009; Soosalu et al., 2010) or reduce the effective normal stress on the border or cross-rift faults (e.g., Reyners et al., 2007), as opposed to an
unusually strong lower crust (e.g., Craig et al., 2011). Further seismic monitoring, both in this area and of Ethiopian volcanoes in general, would provide greater opportunity to observe such lower-crustal events again in the future, constrain source process, and identify how magma migrates from mantle to crust in continental rift zones.

### Data Availability Statement

The seismic network, XM, and the waveforms used in this study are open access and available through IRIS Data Services (http://service.iris.edu/fdsnws/dataselect/1/). See https://www.fdsn.org/networks/detail/XM_2012 for further details on data access and availability. Velocity models, phase arrival pick times, a catalog of events, and an example link to download SAC waveform data through IRIS can be found in the supporting information.

### References


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**Acknowledgments**

We would like to thank SEIS-UK for the use of their equipment, for their help while in the field, and upon managing the data when back in the United Kingdom. Likewise, we thank various collaborators from the Ethiopian Electric Power Corporation (EEPCo) and the Geological Survey of Ethiopia (GSE) for their contributions to the project. The Bristol University Microseismic Projects (BUMPS) provided funding for the seismic experiment and fieldwork, and the seismic equipment was loaned from SEIS-UK with GEF loan 962. Author S. L. is supported by a GWA + Doctoral Training Partnership studentship at the Natural Environment Research Council (NERC) (NE/L002434/1). Author M. W. was funded by an EPSRC studentship, and author K. V. C. is supported by the AXA Research Fund and a Royal Society Wolfson Merit Award. Author A. N. was supported by NERC (NE/R001154/1, REMIS: Reliable Earthquake Magnitudes for Induced Seismicity). This work is a contribution to the NERC funded RiftVolc project (NE/L013932/1, Rift volcanism: past, present and future).


