This manuscript is a preprint and was submitted for publication in *Geophysical Research Letters* in January 2021. Please note that this manuscript has not undergone formal peer review, nor has it been formally accepted for publication. Subsequent versions of this manuscript may have slightly different content. Please feel free to contact Oliver Lamb with your feedback or comments using the email address

7 below.

8

9

10

11

22

1

Acoustics	from low-magnitude fluid-induced
	earthquakes in Finland

Oliver D. Lamb<sup>1</sup>, Jonathan M. Lees<sup>1</sup>, Peter E. Malin<sup>2,3</sup>, Tero Saarno<sup>4</sup>

<sup>1</sup>Department of Geological Sciences, University of North Carolina at Chapel Hill, Chapel Hill, NC, USA
 <sup>2</sup>Earth and Ocean Sciences, Nicholas School of the Environment, Duke University, Durham, NC, USA
 <sup>3</sup>ASIR Advanced Seismic Instrumentation and Research, Dallas, TX, USA
 <sup>4</sup>St1 Deep Heat Oy, Helsinki, Finland

# <sup>16</sup> Key Points:

17	•	Audible noises were reported during induced earthquake sequence in Helsinki Metropoli-
18		tan area in 2018
19	•	Two microphone arrays were deployed and captured signals from 39 earthquakes
20		with moment magnitudes ranging from $-0.07$ to $1.87$ .
21	•	Acoustics were likely generated by ground reverberation during the arrival of seis-

mic body waves at the surface.

Corresponding author: Oliver Lamb, olamb@email.unc.edu

#### 23 Abstract

Earthquakes are frequently accompanied by public reports of audible low-frequency noises. 24 In 2018, public reports of booms or thunder-like noises were linked to induced earthquakes 25 during a Engineered Geothermal System project in the Helsinki Metropolitan area. In 26 response, two microphone arrays were deployed to study these acoustics while stimula-27 tion at the drill site continued. During the 11 day deployment, we find 39 earthquakes 28 accompanied by recognizable atmospheric acoustics. Moment magnitudes of these events 29 ranged from -0.07 to 1.87 with located depths of 4.8 to 6.5 km. Analysis of the largest 30 event revealed a broadband frequency content, including in the audible range, and high 31 apparent velocities across the arrays. We conclude that the audible noises were gener-32 ated by local ground reverberation during the arrival of seismic body waves. The inclu-33 sion of acoustic monitoring at future geothermal development projects will be beneficial 34 for studying seismic-to-acoustic coupling during sequences of induced earthquakes. 35

### <sup>36</sup> Plain Language Summary

Earthquakes are often accompanied by low thunder-like or booming noises. This 37 was the case during geothermal pilot project in the Helsinki Metropolitan area in the 38 summer of 2018, where dozens of local residents reported noises while small earthquakes 39 were occurring below. To investigate how these noises might be generated, we deployed 40 two clusters of microphones in the area to record the noises. Over 11 days, we found 39 41 earthquakes that also generated noises loud enough to be recorded by the microphones. 42 The timing of noises arriving at each cluster of sensors led us to conclude that these noises 43 were being generated by shaking of the ground around the microphones. This incident 44 demonstrated how noises from induced earthquakes might generate major public con-45 cern and that future geothermal projects can benefit from deploying microphones to help 46 with their response. 47

# 48 1 Introduction

Earthquakes of a wide range of magnitudes are commonly accompanied by reports 49 and/or measurements of atmospheric acoustic waves at various epicentral distances. These 50 waves may have frequencies ranging from infrasonic (< 20 Hz) up to and beyond the min-51 imum limit of human hearing ability (20 - 70 Hz). Cases of the latter have been described 52 as low rumbling sounds or booms (Michael, 2019), and have been reported for shallow 53 (<2 km) earthquakes in the USA (Ebel et al., 1982) and France (Sylvander & Mogos, 54 2005; Sylvander et al., 2007; Thouvenot et al., 2009). The event magnitudes associated 55 with these sounds have been stated to be as low as -2 and -0.7, respectively. Audible noises 56 are also frequently reported for larger magnitude earthquakes, and accompanied by the 57 frequent detection of infrasonic acoustic waves at global distances (e.g. Mikumo, 1968; 58 Young & Greene, 1982; Olson et al., 2003; Le Pichon et al., 2003; Mutschlecner & Whitaker, 59 2005; Le Pichon et al., 2006; Arrowsmith et al., 2012). Mapping of acoustic sources dur-60 ing and immediately after earthquakes has identified three sources of earthquake acous-61 tics (Arrowsmith et al., 2010): i) 'epicentral' (i.e. seismic-to-acoustic coupling directly 62 above or near the earthquake epicenter; Mikumo, 1968; Young & Greene, 1982), ii) 'lo-63 cal' (i.e. generated by the passage of seismic waves near sensor located at distance from 64 epicenter; Cook, 1971; Kim et al., 2004) and iii) 'secondary' (i.e. generated by interac-65 tion of seismic waves with topographic features; Young & Greene, 1982; Mutschlecner 66 & Whitaker, 2005; Shani-Kadmiel et al., 2018; Johnson et al., 2020). 'Epicentral' acous-67 tics have been attributed primarily to vertically propagating body waves (particularly 68 P- and SV-waves) coupling directly into the atmosphere through ground motion at the 69 Earth's surface (Hill et al., 1976). Seismo-acoustic recordings of earthquake acoustics at 70 local or epicentral distances are limited to only a few studies (e.g. Hill et al., 1976; Syl-71 vander et al., 2007; Johnson et al., 2020). Here we describe a case study of epicentral 72

<sup>73</sup> acoustic waves generated by earthquakes during a hydraulic stimulation project in Fin-

<sup>74</sup> land, one of the first documented recordings of acoustics from an induced earthquake se-

<sup>75</sup> quence and are amongst the lowest magnitude events to be recorded.

## <sup>76</sup> 2 St1 Deep Heat Oy Venture

The St1 Deep Heat Oy energy-company Engineered Geothermal System (EGS) pi-77 lot project was located in the Helsinki Metropolitan area within the campus of Aalto Uni-78 versity (Fig. 1). The aim of the project was to develop an EGS facility in order to pro-79 duce a sustainable baseload for the local district heating system (Kwiatek et al., 2019). 80 In 2018, a 6.1 km deep stimulation well was drilled into crystalline Precambrian Svecofen-81 nian basement rocks consisting of granites, pegmatites, gneisses, and amphibolites (Kwiatek 82 et al., 2019); this bedrock is only locally covered by a thin (<10 m) layer of glacial till 83 or soil (Hillers et al., 2020). From 4 June to 22 July 2018, a total of  $18,160 \text{ m}^3$  of wa-84 ter was pumped into the stimulation well at depths of 5.7 to 6.1 km; this included mov-85 ing injection intervals and multiple stoppages for a few days (Kwiatek et al., 2019; Hillers 86 et al., 2020). Induced seismicity was monitored by an extensive seismic network, includ-87 ing 3-component borehole seismometers installed in 0.3 to 1.15 km deep wells at distances 88 up to 8.2 km from the drill site (Fig. 1). The purpose of the seismic network was to pro-89 vide accurate hypocenter locations and magnitudes of induced earthquakes for both in-90 dustrial and regulatory purposes (i.e. Traffic Light System; Kwiatek et al., 2019; Ader 91 et al., 2020). 92



Figure 1. (a) Topographic map of the region around the St1 drill site (cyan cross) showing locations and names of borehole seismic stations (blue circles) and temporary acoustic arrays (red triangles). Also plotted are locations of earthquakes recorded during the acoustic deployment, colored by depth. Red star indicates the location of the  $M_w$  1.87 event. Inset: Map of Finland showing location of the Helsinki Metropolitan area. Panels (b) and (c) show the infrasound sensor distribution for arrays FIN1 and FIN2, respectively, with back azimuth direction to the ST1 drill site indicated by the blue arrow.

From 4 June to 1 August 2018, a total of 8412 earthquakes were automatically recorded by the network out of which 1977 were suitable for relocations and magnitude calculations (Kwiatek et al., 2019). These events were located across three distinct clusters rang-

 $_{96}$  ing in depths of 4.8 – 6.6 km and moment magnitudes (M<sub>w</sub>) of -0.76 to 1.86 (Fig. S1 in

Supporting Information). Fault plane solutions for a set of selected events indicated re-97 verse faulting along pre-existing fractures associated with NW-SE trending fault zones 98 reactivated by the hydraulic injection (Hillers et al., 2020). The Institute of Seismology 99 at the University of Helsinki (ISUH) collected 220 public reports of felt earthquakes, which 100 unexpectedly also included dozens of audible disturbances, typically described as thunder-101 or blast-like (Ader et al., 2020; Hillers et al., 2020). The largest and most reported event 102 was a  $M_w$  1.87 event on 8 July 2018 located at 6.3 km depth (Fig. 1). This event gen-103 erated 78 public reports and was apparently heard up to 9 km away from the epicenter 104 (Hillers et al., 2020). Notably, spatial distributions of the reports were strongly corre-105 lated with the SH radiation pattern of the reverse faulting mechanism in the event (Hillers 106 et al., 2020). 107

### <sup>108</sup> **3** Data and Methods

In response to the reports of audible earthquake events, we deployed two tempo-109 rary arrays of infrasound microphones in the area from 7 - 18 July to study the nature 110 of these atmospheric acoustics. The arrays were deployed at distances of  $\sim 2.5$  and  $\sim 2.2$ 111 km from the St1 drill site. Each deployment consisted of three microphones extended 112 on cables up to 35 m from a central data recorder, where a fourth microphone was lo-113 cated (Fig. 1b, c). The data recorder was a REFTEK RT 130 data logger which pro-114 vided a 24-bit, GPS-time synchronized recording set to 100 samples per second, result-115 ing in an anti-aliasing Finite Impulse Response (FIR) filter cut off of 40 Hz. The micro-116 phones were identical infraBSU (vers1) microphones, which incorporate a MEMS sen-117 sor and capillary filters to provide a flat response at >0.1 Hz (Marcillo et al., 2012). To 118 aid analysis and interpretation of acoustic data in this study, we also included seismic 119 data from borehole seismometers located near each array (TAGC and MURA; Fig. 1a). 120 Each seismometer was composed of a three-component Sunfull PSH geophone sensor ( $f_N$ 121 = 4.5 Hz) recording at 500 samples per second and located  $\sim 1.15$  km below the surface 122 (For more information, see Kwiatek et al., 2019). 123

For this study, all data were filtered with a 2 Hz high-pass Butterworth filter to 124 reduce continuous background noise (unless otherwise indicated). Data were manually 125 inspected for consistent arrivals across at least two microphones in each array to assess 126 if earthquake-generated atmospheric acoustic waves were detected following an induced 127 earthquake. To estimate the arrival times for different body wave phases at each array, 128 we use P- and S-wave velocities of 6.25 and 3.75 km.s<sup>-1</sup> respectively, as estimated from 129 130 borehole logs at the St1 drill site (see supplementary materials in Kwiatek et al., 2019). One of the key advantages of deploying acoustic microphones in an array configuration 131 is it permits the calculation of back azimuth direction and slowness of acoustic waves 132 propagating across the deployment. Here we estimated back azimuths and slowness val-133 ues for 0.1 s windows with 90% overlap within the first 3 s after the initiation time of 134 the earthquake. We used waveform envelopes, determined from the square root of the 135 Hilbert Transform, which were then smoothed using the average of an 8 sample moving 136 window (Fig. 4a, b). All analysis presented here was carried out within the ObsPy python 137 package (Krischer et al., 2015). 138

# 139 4 Observations

During 7 – 18 July, 266 earthquakes were detected and relocated within a few hundred metres of the stimulation interval. These events occurred at depths of 4.8 to 6.5 km below sea level and had moment magnitudes ranging from -0.19 to 1.87 (Fig. 1a, 2a, b). Of the 266 earthquakes, 39 were followed shortly by atmospheric disturbances across at least one array that may be interpreted as earthquake associated acoustic waves (Fig. 2). Atmospheric disturbances were more commonly seen at FIN2 (n=36) than FIN1 (n=9), with only 3 events seen exclusively at the latter. The smallest event was a M<sub>w</sub> -0.07 on <sup>147</sup> 8 July, and the largest was the widely heard M<sub>w</sub> 1.87 on the same day (Fig. 2c). As the
<sup>148</sup> latter earthquake produced the highest signal-to-noise ratios at both microphone arrays,
<sup>149</sup> the remainder of this section will focus on the analysis of acoustic data from this par<sup>150</sup> ticular event.



Figure 2. Moment magnitudes (a) and depths (b) of the 266 relocated seismic events recorded during the infrasound array deployment near the St1 Deep Heat Oy EGS project. Red 'x' indicate the events which were detected by at least one acoustic array. (c) 6 s of normalised acoustic data (highpass filtered at 5 Hz) recorded by sensor 2 at FIN2 after the initiation of five example earthquakes, including the lowest and highest magnitude events. Calculated  $M_w$  and depth (Z) of each event is indicated on the right. (See figures S2 to S11 in Supporting Information for waveforms and frequency spectrograms from all microphones for each event.)

For the  $M_w$  1.87 event the acoustic data recorded at FIN2 have peak amplitudes 151 an order of magnitude larger than those recorded at FIN1 (Fig. 3c, g). Frequency spec-152 tra highlight the broadband nature of the atmospheric acoustics, with frequencies rang-153 ing from 2 to 50 Hz (Fig. 3d, h). The acoustic waves and their spectra at each array ap-154 pear to show distinct multi-phase arrivals that correlate with seismic waves recorded at 155 the nearby borehole seismometers (Fig. 3a, b, e, f). The different arrival phases at each 156 array appear to be coincident with the predicted arrivals of P- and S-waves (dotted and 157 dashed red lines in Fig. 3). The highest acoustic amplitudes are correlated with the ar-158 rival of the S-waves at each array. Calculated values of back azimuth and slowness are 159 generally well scattered across the analysed 3 s time window (Fig. 4b-d). However, at 160 or near the estimated time of arrivals for P- and S-waves (red lines in Fig. 4a, b), the 161 back azimuths indicate arrivals from the direction of the  $M_w$  1.87 event epicenter (Fig. 162 4c, d). Slowness values at these times indicate relatively high propagation velocities across 163 the array (Fig. 4e, f). 164

#### 165 5 Discussion

Here we have presented evidence for infrasonic and audible atmospheric acoustics generated by low magnitude fluid-induced earthquakes. These observations are notable for two reasons: i) these are the first recorded earthquake-generated acoustics from induced earthquakes, and ii) they represent the lowest magnitude events to be recorded



Figure 3. Waveforms (left column) and their respective frequency spectrograms (right column) of the  $M_w 1.86$  event as recorded by seismic station MURA (a, b), acoustic array FIN2 (c,d), seismic station TAGC (e, f) and acoustic array FIN1 (g, h). Note that the seismic waveforms are from the north component of the station. Also plotted is the time of the event (solid red line), as well as predicted arrival times for P- and S-wave phases (dotted and dashed red lines, respectively).

by acoustic microphones. (There are reports of audible noises from earthquakes with mag-170 nitudes as low as -2 (Thouvenot et al., 2009) but these events were not recorded with 171 microphones.) Manual inspection of acoustic data identified at least 39 events where acous-172 tic waves were recorded propagating across at least one array of sensors (Fig. 2). This 173 represents only 15% of all earthquakes relocated during the deployment, but the loca-174 tion of the arrays within a large metropolitan area with a large number of noise sources 175 may have acted to reduce this proportion. The acoustic waves contained broadband fre-176 quency ranges from 2 up to 40 Hz, and possibly higher but is limited by the anti-alias 177 FIR filter of the sample recording rate (Fig. 3d, h). This frequency range overlaps with 178 the lower range of human hearing (down to 20 Hz), therefore confirming that thunder-179 or blast-like sounds heard by the public were generated by the earthquakes (Ader et al., 180 2020; Hillers et al., 2020). These frequency ranges also match previously reported val-181 ues from audible natural earthquakes (Hill et al., 1976; Sylvander et al., 2007). 182

The significant scattering of back azimuth and slowness values before the arrival 183 of atmospheric acoustics (Fig. 4) is interpreted to be a result of the large number of noise 184 sources found in a metropolitan area. However, during the expected arrival of the P- and 185 S-waves the back azimuth values align at or around the direction of the earthquake epi-186 center (Fig. 4c, d). Simultaneously, the slowness values indicate relatively high propa-187 gation values across the array (Fig. 4e, f). These values correlate with waves of either 188 high velocities  $(>1 \text{ kms}^{-1})$  or near-vertical wave arrival directions at the array. Consid-189 ering the ratio between earthquake depths (4.8 - 6.5 km) and epicenter-array distances 190 (<2.5 km), it is reasonable to expect near vertical arrival angles of seismic waves at each 191 array. Therefore, the atmospheric acoustics recorded during the largest earthquake, and 192 all other recorded events, were generated by ground motion during and immediately af-193



Figure 4. Beamforming results for arrays FIN1 (left column) and FIN2 (right column) for the first 3 seconds after the  $M_w 1.86$  event. (a, b) Smoothed waveform envelopes from each element in each array. Dotted and dashed lines plot the estimated arrival times of P- and S-waves, respectively. (c, d) Back azimuth calculations for 0.2 s moving windows with 90% overlap. Horizontal dotted lines plot the azimuth from each array to the  $M_w 1.86$  event epicenter. (e, f) Calculated slowness values across each array for each 0.2 s window. Points in panels c-f are colored by relative power, where lighter colors indicate higher relative power (i.e. the signal power of the mean waveform for peak slowness divided by average element power in same time window).

ter the arrival of P- and S-waves at the ground surface within close proximity of the microphone arrays.

A notable observation from the public reports compiled during the induced earth-196 quake sequence is the geographical distribution of disturbances correlated with the ra-197 diation patterns of S-waves (See Fig. 5 in Hillers et al., 2020). The FIN2 acoustic array 198 was located adjacent to the area with the greatest number of reports. This pattern cor-199 relates with the amplitude difference between the acoustic waves recorded at FIN1 and 200 FIN2 for the  $M_w$  1.86 event, with amplitudes an order of magnitude higher at the lat-201 ter than the former (Fig. 3c, d). Furthermore, a higher number of earthquake-generated 202 acoustic waves were recorded at FIN2 (N=36) than at FIN1 (N=9). Another factor to 203 consider is that the FIN1 array was deployed on the margin of an active golf course which 204 was built on top of a former municipal waste landfill, while FIN2 was deployed in an area 205 where buildings are frequently constructed directly onto outcropping bedrock. This sug-206 gests that the presence of a soft sedimentary layer above the bedrock may act as a damp-207 ener during seismic-to-acoustic coupling. 208

Given that the infrasound sensors are typically placed in direct contact with the 209 ground surface during deployments, contamination of recorded infrasound signals by phys-210 ical shaking of the sensor could be a concern. However, testing of the seismic response 211 of various acoustic sensors have consistently concluded that physical vibration does not 212 significantly influence the recorded infrasound signals (Bedard, 1971; Hill et al., 1976; 213 Sylvander et al., 2007). The MEMS-based microphones used in this study (InfraBSU vers1) 214 have low inertial mass and are similar in design to the MEMS-based transducers described 215 in Marcillo et al. (2012). These sensors were found to have minimal seismic-to-noise cou-216 pling during calibration studies at the Facility for Acceptance, Calibration and Testing 217

site at the Los Alamos National Laboratory (Johnson et al., 2020). Therefore, we do not
 consider direct seismic shaking of the sensor to be of importance in the acoustic signals
 presented here.

A common observation in previous earthquake acoustic studies is the presence of 221 secondary infrasound generated away from the earthquake epicenter (Young & Greene, 222 1982; Le Pichon et al., 2003; Mutschlecner & Whitaker, 2005; Arrowsmith et al., 2010; 223 Shani-Kadmiel et al., 2018; Johnson et al., 2020). These acoustics are confirmed to be 224 caused by the interaction of surface waves with topography or other significant crustal 225 features such as sedimentary basins (Mutschlecner & Whitaker, 2005; Arrowsmith et al.. 226 2010). These are usually manifested as a unusually long coda of secondary arrivals af-227 ter the local infrasound phases (e.g. Johnson et al., 2020). The infrasound waves described 228 here have relatively short durations with no significant coda, therefore we infer that no 229 secondary infrasound has been generated by the induced earthquakes. We interpret this 230 as a result of the low magnitudes of the events, as well as the lack of steep topograph-231 ical features around the St1 drill site (Fig. 1a). However, due to the location within an 232 metropolitan area, we cannot rule out the presence of acoustics generated by mechan-233 ical shaking of buildings or other structures (e.g. bridges) near each array. Altogether, 234 we interpret the acoustics presented here as 'epicentral' earthquake acoustics generated 235 by ground surface reverberation during the direct arrival of body waves generated by fluid-236 induced earthquakes. 237

### 238 6 Conclusions

Acoustic monitoring can help explain human observations and may also provide 239 quantitative insights into the mechanics of ground motions responsible for generating earth-240 quake sounds. Here we have presented acoustic events recorded within the Helsinki Metropoli-241 tan area in July 2018 during hydraulic stimulation at a pilot Engineered Geothermal Sys-242 tem project. Based on the estimated timing of body wave arrivals, frequency content of 243 the waveforms, as well as estimated slowness calculations, we have interpreted these acous-244 tic events as being generated by reverberation of the ground surface during the arrival 245 of P- and S-waves from induced low magnitude earthquakes. Although only a minor pro-246 portion of induced earthquakes generated recognizable acoustic waves, events with mo-247 ment magnitudes ranging from -0.07 to 1.87 were recorded with acoustic microphones 248 at the surface. As far as we are aware, these events represent the first induced earthquakes 249 and are amongst the lowest magnitude events to be recorded with acoustic microphones. 250 Given that Traffic Light Systems are increasingly being implemented to reduce the po-251 tential seismic hazard due to induced seismicity (Ader et al., 2020), and the consider-252 able public interest generated by audible earthquakes in the Helsinki Metropolitan area 253 (Ader et al., 2020; Hillers et al., 2020), future projects for developing geothermal sys-254 tems can benefit from deploying acoustic sensors to provide more detailed information 255 in responses to public concern. 256

#### 257 Acknowledgments

The authors wish to thank Dr Peter Leary and the technicians at the St1 Deep Heat project 258 for their help and logistical support during the acoustic sensor deployment. This research 259 was performed while ODL held an NRC Research Associateship with the U.S. Army Re-260 search Laboratory/Army Research Office while based at the University of North Car-261 olina at Chapel Hill. All acoustic data presented here will be made available, without 262 undue reservation, to any qualified researcher. Data from the borehole seismometers were 263 transmitted to the Institute of Seismology at the University of Helsinki as part of a regulatory agreement with the city of Espoo and have not been released to the public. To-265 pographic data used in Fig. 1a were downloaded from the National Land Survey of Fin-266 land via the Open data file download service (last accessed December 2020). 267

References

268

269	Ader, T., Chendorain, M., Free, M., Saarno, T., Heikkinen, P., Malin, P. E.,
270	et al. (2020). Design and implementation of a traffic light system for
271	deep geothermal well stimulation in Finland. Journal of Seismology, $24(5)$ ,
272	991–1014. doi: $10.1007/s10950-019-09853-y$
273	Arrowsmith, S. J., Burlacu, R., Pankow, K., Stump, B., Stead, R., Whitaker, R.,
274	& Hayward, C. (2012). A seismoacoustic study of the 2011 January 3 Cir-
275	cleville earthquake. Geophysical Journal International, $189(2)$ , $1148-1158$ . doi:
276	10.1111/j.1365-246X.2012.05420.x
277	Arrowsmith, S. J., Johnson, J. B., Drob, D. P., & Hedlin, M. A. (2010). The seis-
278	moacoustic wavefield: A new paradigm in studying geophysical phenomena.
279	Reviews of Geophysics, $48(4)$ , 1-23. doi: $10.1029/2010$ RG000335
280	Bedard, A. J. (1971). Seismic Response of Infrasonic Microphones. Journal of
281	Research of the National Bureau of Standards - C. Engineering and Instrumen-
282	$tation, \ 75C(1), \ 41-45.$
283	Cook, R. K. (1971). Infrasound radiated during the montana earthquake of 1959
284	august 18. $Geophysical Journal of the Royal Astronomical Society, 26(1-4),$
285	191–198. doi: 10.1111/j.1365-246X.1971.tb03393.x
286	Ebel, J. E., Vudler, V., & Celata, M. (1982). The 1981 microearthquake swarm near
287	Moodus, Connecticut. Geophysical Research Letters, 9(4), 397–400. doi: 10
288	.1029/GL009i004p00397
289	Hill, D. P., Fischer, F. G., Lahr, K. M., & Coakley, J. M. (1976). Earthquake sounds
290	generated by body-wave ground motion. Bulletin of the Seismological Society
291	of America, $66(4)$ , 1159–1172.
292	Hillers, G., T. Vuorinen, T. A., Uski, M. R., Kortström, J. T., Mäntyniemi, P. B.,
293	Tiira, T., Saarno, T. (2020). The 2018 Geothermal Reservoir Stimulation
294	in Espoo/Helsinki, Southern Finland: Seismic Network Anatomy and Data
295	Features. Seismological Research Letters. doi: 10.1785/0220190253
296	Johnson, J. B., Mikesell, T. D., Anderson, J. F., & Liberty, L. M. (2020). Mapping
297	the sources of proximal earthquake infrasound. <i>Geophysical Research Letters</i> ,
298	47(23), 19. doi: 10.1029/2020GL091421
299	Kim, T. S., Hayward, C., & Stump. (2004). Local infrasound signals from the
300	tokachi-oki earthquake. Geophysical Research Letters, 31(20), L20605. doi:
	10 1020 /2004 CL 021172
301	10.1029/2004GL021178
301 302	Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &
301 302 303	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp;</li> <li>Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific</li> </ul>
301 302 303 304	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp;</li> <li>Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific</li> <li>Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi:</li> </ul>
301 302 303 304 305	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp;</li> <li>Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific</li> <li>Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> </ul>
301 302 303 304 305 306	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. <i>Computational Science &amp; Discovery</i>, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain,</li> </ul>
301 302 303 304 305 306 307	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-</li> </ul>
301 302 303 304 305 306 307 308	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi:</li> </ul>
301 302 303 304 305 306 307 308 309	<ul> <li>10.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> </ul>
301 302 303 304 305 306 307 308 309 310	<ul> <li>10.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infra-</li> </ul>
301 302 303 304 305 306 307 308 309 310 311	<ul> <li>IO.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake.</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 311 312 313	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312 313 314	<ul> <li>IO.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic observations of the Chilean earthquake of 2005 June 13. Geophysical Journal</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312 313 314 315	<ul> <li>IO.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic observations of the Chilean earthquake of 2005 June 13. Geophysical Journal International, 167(2), 838-844. doi: 10.1111/j.1365-246X.2006.03190.x</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316	<ul> <li>10.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic observations of the Chilean earthquake of 2005 June 13. Geophysical Journal International, 167(2), 838-844. doi: 10.1111/j.1365-246X.2006.03190.x</li> <li>Marcillo, O., Johnson, J. B., &amp; Hart, D. (2012). Implementation, characteriza-</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317	<ul> <li>IO.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic observations of the Chilean earthquake of 2005 June 13. Geophysical Journal International, 167(2), 838-844. doi: 10.1111/j.1365-246X.2006.03190.x</li> <li>Marcillo, O., Johnson, J. B., &amp; Hart, D. (2012). Implementation, characterization, and evaluation of an inexpensive low-power low-noise infrasound sensor</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic observations of the Chilean earthquake of 2005 June 13. Geophysical Journal International, 167(2), 838-844. doi: 10.1111/j.1365-246X.2006.03190.x</li> <li>Marcillo, O., Johnson, J. B., &amp; Hart, D. (2012). Implementation, characterization, and evaluation of an inexpensive low-power low-noise infrasound sensor based on a micromachined differential pressure transducer and a mechanical</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319	<ul> <li>IO.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic observations of the Chilean earthquake of 2005 June 13. Geophysical Journal International, 167(2), 838-844. doi: 10.1111/j.1365-246X.2006.03190.x</li> <li>Marcillo, O., Johnson, J. B., &amp; Hart, D. (2012). Implementation, characterization, and evaluation of an inexpensive low-power low-noise infrasound sensor based on a micromachined differential pressure transducer and a mechanical filter. Journal of Atmospheric and Oceanic Technology, 29(9), 1275–1284. doi:</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320	<ul> <li>IO.1029/2004GL021178</li> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic observations of the Chilean earthquake of 2005 June 13. Geophysical Journal International, 167(2), 838-844. doi: 10.1111/j.1365-246X.2006.03190.x</li> <li>Marcillo, O., Johnson, J. B., &amp; Hart, D. (2012). Implementation, characterization, and evaluation of an inexpensive low-power low-noise infrasound sensor based on a micromachined differential pressure transducer and a mechanical filter. Journal of Atmospheric and Oceanic Technology, 29(9), 1275–1284. doi: 10.1175/JTECH-D-11-00101.1</li> </ul>
301 302 303 304 305 306 307 308 309 310 311 312 313 314 315 316 317 318 319 320 321	<ul> <li>Krischer, L., Megies, T., Barsch, R., Beyreuther, M., Lecocq, T., Caudron, C., &amp; Wassermann, J. (2015). ObsPy: a bridge for seismology into the scientific Python ecosystem. Computational Science &amp; Discovery, 8(1), 1–17. doi: 10.1088/1749-4699/8/1/014003</li> <li>Kwiatek, G., Saarno, T., Ader, T., Bluemle, F., Bohnhoff, M., Chendorain, M., et al. (2019). Controlling fluid-induced seismicity during a 6.1-km-deep geothermal stimulation in Finland. Science Advances, 5. doi: 10.1126/sciadv.aav7224</li> <li>Le Pichon, A., Guilbert, J., Vallée, M., Dessa, J. X., &amp; Ulziibat, M. (2003). Infrasonic imaging of the Kunlun Mountains for the great 2001 China earthquake. Geophysical Research Letters, 30(15). doi: 10.1029/2003GL017581</li> <li>Le Pichon, A., Mialle, P., Guilbert, J., &amp; Vergoz, J. (2006). Multistation infrasonic observations of the Chilean earthquake of 2005 June 13. Geophysical Journal International, 167(2), 838-844. doi: 10.1111/j.1365-246X.2006.03190.x</li> <li>Marcillo, O., Johnson, J. B., &amp; Hart, D. (2012). Implementation, characterization, and evaluation of an inexpensive low-power low-noise infrasound sensor based on a micromachined differential pressure transducer and a mechanical filter. Journal of Atmospheric and Oceanic Technology, 29(9), 1275–1284. doi: 10.1175/JTECH-D-11-00101.1</li> <li>Michael, A. J. (2019). Earthquake Sounds. In H. K. Gupta (Ed.), Encyclopedia</li> </ul>

323	$.1007/978$ - $3$ - $030$ - $10475$ - $7$ _ $201$ - $1$
324	Mikumo, T. (1968). Atmospheric pressure waves and tectonic deformation asso-
325	ciated with the Alaskan earthquake of March 28, 1964. Journal of Geophysical
326	Research, 73(6), 2009–2025. doi: 10.1029/JB073i006p02009
327	Mutschlecner, J. P., & Whitaker, R. W. (2005). Infrasound from earthquakes. Jour-
328	nal of Geophysical Research, 110(1), 1–11. doi: 10.1029/2004JD005067
329	Olson, J. V., Wilson, C. R., & Hansen, R. A. (2003). Infrasound associated with the
330	2002 Denali fault earthquake, Alaska. Geophysical Research Letters, $30(23)$ .
331	doi: 10.1029/2003GL018568
332	Shani-Kadmiel, S., Assink, J. D., Smets, P. S. M., & Evers, L. G. (2018). Seis-
333	moacoustic Coupled Signals From Earthquakes in Central Italy: Epicentral
334	and Secondary Sources of Infrasound. $Geophysical Research Letters, 45(1),$
335	427–435. doi: 10.1002/2017GL076125
336	Sylvander, M., & Mogos, D. G. (2005). The sounds of small earthquakes: Quan-
337	titative results from a study of regional macroseismic bulletins. Bulletin of the
338	Seismological Society of America, 95(4), 1510–1515. doi: 10.1785/0120040197
339	Sylvander, M., Ponsolles, C., Benahmed, S., & Fels, J. F. (2007). Seismoacous-
340	tic recordings of small earthquakes in the Pyrenees: Experimental results. Bul-
341	letin of the Seismological Society of America, 97(1 B), 294-304. doi: 10.1785/
342	0120060009
343	Thouvenot, F., Jenatton, L., & Gratier, JP. (2009). 200-m-deep earthquake
344	swarm in Tricastin (lower Rhône Valley, France) accounts for noisy seis-
345	micity over past centuries. Terra Nova, 21(3), 203–210. doi: 10.1111/
346	j.1365-3121.2009.00875.x
347	Young, J., & Greene, G. (1982). Anomalous infrasound generated by the Alaskan
348	earthquake of 28 March 1964. The Journal of the Acoustical Society of Amer-
349	ica, 71, 334-339. doi: $10.1121/1.387457$