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## Depth Migration of Seismovolcanic Tremor Sources below the Klyuchevskoy Volcanic Group (Kamchatka) determined from a Network-Based Analysis

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### 15 Key Points:

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16	• A seismic network-based method for the automatic 3D location of volcanic tremors
17	is developed
18	• The migration of deep volcanic tremor sources to the surface is tracked in time
19	beneath Klyuchevskoy volcano
20	• The developed location method is fully automatic and can be updated continu-
21	ously with new data

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#### 22 Abstract

We present a method for automatic location of dominant sources of seismovolcanic 23 tremor in 3D, based on the spatial coherence of the continuously recorded wavefield at 24 a seismic network. We analyze 4.5 years of records from the seismic network at the Klyuchevskoy 25 volcanic group in Kamchatka, Russia, when 4 volcanoes experienced tremor episodes. 26 After enhancing the tremor signal with spectral whitening, we compute the daily cross-27 correlation functions related to the dominant tremor sources from the first eigenvector 28 of the spectral covariance matrix, and infer their daily positions in 3D. We apply our tech-29 nique to the tremors beneath Shiveluch, Klyuchevskoy, Tolbachik and Kizimen volca-30 noes, and observe the year-long pre-eruptive volcanic tremor beneath Klyuchevskoy from 31 deep to shallow parts of the plumbing system. This observation of deep volcanic tremor 32 sources demonstrates that the cross-correlation based method is a very powerful tool for 33 volcano monitoring. 34

#### <sup>35</sup> Plain Language Summary

Volcanic tremors are the seismic signature of magmatic and hydrothermal fluids 36 passing through volcanic conduits. Locating them in 3D is of real interest because it could 37 allow us to monitor in more details movements of magma inside volcanic edifices and in 38 some cases, to forecast eruptive episodes. The location of tremor in 3D nevertheless re-39 mains challenging because signals generated by tremors do not present any clear onset 40 that could be used for picking arrival times and for determining source location. We de-41 sign a method based on cross-correlations that recovers the differential travel times be-42 tween receivers of a seismic network from the analysis of the statistically dominating waves 43 in the wavefield. We present an application of the proposed method to volcanoes in Kam-44 chatka, Russia, and show that we can track the pre-eruptive tremor episode in depth be-45 fore the main eruption of the Klyushevskoy volcano. 46

#### 47 **1 Introduction**

The long-period volcanic seismicity, composed of long-period events and volcanic tremors, is mainly thought to be related to processes induced by fluid movements within magmatic or hydrothermal volcanic systems (B. A. Chouet et al., 1996; Iverson et al., 2006). As it constitutes an important attribute of volcanic unrest, its detection and char-

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52	acterization is a key aspect of volcano monitoring and eruption forecasting (McNutt, 1992;
53	Sparks et al., 2012; B. A. Chouet & Matoza, 2013). Long-period earthquakes and vol-
54	canic tremors derive from similar processes but they differ in nature. Long-period earth-
55	quakes are discrete emergent signals of high-frequency onset followed by a low frequency
56	harmonic waveform of typical frequency range $0.5 - 5$ Hz (B. A. Chouet et al., 1996),
57	that can sometimes become very frequent in episodes called "drumbeats" . Volcanic tremors
58	are highly irregular continuous signals that can last from a few minutes to several months,
59	of which the spectral content can be harmonic, monochromatic, banded, spasmodic, etc.
60	(Konstantinou & Schlindwein, 2003). However, both phenomena are believed to be gen-
61	erated by similar mechanisms such as magma moving through narrow cracks, fragmen-
62	tation and pulsation of pressurized fluids within the volcano, or escape of pressurized steam
63	and gases from fumaroles (B. A. Chouet & Matoza, 2013).
64	Long-period earthquakes are observed both in shallow parts of volcanoes (B. A. Chouet
65	et al., 1996; Ciaramella et al., 2011; Firstov & Shakirova, 2014; Jolly et al., 2017; N. Shapiro
66	et al., 2017; W. B. Frank et al., 2018; Woods et al., 2018, e.g.,) as well as deep parts (Ukawa
67	& Ohtake, 1987; Hasegawa et al., 1991; R. White et al., 1992; Pitt & Hill, 1994; R. A. White,
68	1996; N. Shapiro et al., 2017; W. B. Frank et al., 2018; Han et al., 2018; Hotovec-Ellis
69	et al., 2018, e.g.,), whereas most of reported volcanic tremor sources are superficial ( $<$
70	5 km). Indeed, except the deep volcanic tremors $(30-60 \text{ km})$ observed at Kilauea vol-
71	cano between 1962 and 1983 (Aki & Koyanagi, 1981; Shaw & Chouet, 1991), the ma-
72	jority of volcanic tremor studies deal with superficial sources. Thus, shallow volcanic tremor
73	sources are reported on eruptive craters and conduits (Goldstein & Chouet, 1994; Métaxian
74	et al., 2002; Battaglia et al., 2005; Di Grazia et al., 2006; Almendros et al., 2014; Ogiso
75	et al., 2015; Ichimura et al., 2018; Moschella et al., 2018, e.g.,), building cones (Haney,
76	2010), gas slug bursts at surface vents (B. Chouet et al., 1997; Ripepe et al., 2001; Fee
77	et al., 2010; Patrick et al., 2011; Yukutake et al., 2017; Barriere et al., 2017; Eibl, Bean,
78	Jónsdóttir, et al., 2017, e.g.,), lava lakes (Métaxian et al., 1997; Barriere et al., 2017; Don-
79	aldson et al., 2017, e.g.,), lava flows (Caudron et al., 2015), shallow dikes (Eibl, Bean,
80	Vogfjörd, et al., 2017; Eibl, Bean, Jónsdóttir, et al., 2017; Woods et al., 2018, e.g.,), magma-
81	water aquifer interfaces (Almendros et al., 1997), etc.
82	Methods currently used for volcanic tremor source location can be divided into fol-

- <sup>83</sup> lowing categories: small aperture array methods (Goldstein & Chouet, 1994; Almendros
- et al., 1997; B. Chouet et al., 1997; Métaxian et al., 2002; Almendros et al., 2007; Haney,

2010; Almendros et al., 2014; Eibl, Bean, Vogfjörd, et al., 2017; Eibl, Bean, Jónsdóttir, 85 et al., 2017, e.g.,), network-based amplitude decay methods (Aki & Koyanagi, 1981; Ripepe 86 et al., 2001; Battaglia et al., 2005; Di Grazia et al., 2006; Taisne et al., 2011; Ogiso et 87 al., 2015; Ichimura et al., 2018; Moschella et al., 2018, e.g.,), and network-based inter-88 stations cross-correlations methods (Ballmer et al., 2013; Droznin et al., 2015; Li et al., 89 2016, 2017; Donaldson et al., 2017; Barriere et al., 2017; Yukutake et al., 2017; Woods 90 et al., 2018, e.g.,). It is interesting to mention that network-migration-based methods 91 have been also used to locate tectonic tremor and low-frequency earthquakes (Kao & Shan, 92 2004; W. Frank & Shapiro, 2014, e.g.,). Dense small aperture arrays are not always avail-93 able because of their cost and their temporal nature, but large scale seismic networks 94 are nowadays permanently deployed in many places. Because of asumptions about ge-95 ometrical spreading and attenuation, network-based amplitude decay methods are adapted 96 to locate volcanic tremor sources relatively close to the used network, i.e. sources that 97 are not so far or not so deep. On the other hand, inter-stations cross-correlations meth-98 ods can observe tremors at larger scales (Ballmer et al., 2013) and might be able to lo-99 cate deeper sources. Previous volcanic tremor location studies based on cross-correlations 100 have located shallow tremor sources at Kilauea (Ballmer et al., 2013; Donaldson et al., 101 2017), Katla (Li et al., 2016, 2017), Nyiragongo and Nyamulagira (Barriere et al., 2017), 102 Hakone (Yukutake et al., 2017) and Bardarbunga (Woods et al., 2018) volcanoes. 103

To our knowledge, no deep volcanic tremor source has been located with this method 104 so far. This is certainly due to the fact that a surface wave propagation is often assumed 105 for the analysis, inherently implying the source to be located in the shallow part. The 106 automatic cross-correlations based location method proposed in our article is a 3D ap-107 proach based on the assumption of dominant S-wave emission from tremor sources and 108 associated S-wave structures, enabling to look for both shallow and deep tremors. By 109 using the daily first eigenvector of the network covariance matrix, we are able to track 110 the temporal evolution on a daily time scale of the 3D location of the daily dominating 111 volcanic tremor sources. 112

We apply the developed method to data continuously recorded by the permanent monitoring network of the Klyuchevskoy group of volcanoes (Kamchatka, Russia, see Fig. 1) during a four and a half years-long time period between January 2009 and June 2013. During this period of time, Klyuchevskoy, Tolbachik, Kizimen and Shiveluch volcanoes experienced episodes of volcanic tremor (Droznin et al., 2015) and abundant amount of

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long-period earthquakes (N. Shapiro et al., 2017). The two most important tremors led
to a summit eruption at the end of 2010 at Klyuchevskoy (Senyukov, 2013, e.g.,) and a
fissure eruption beginning at the end of 2012 at Tolbachik (Gordeev et al., 2013, e.g.,).
Short explosive eruptions also occurred at Bezymianny. The seismic monitoring network
and volcanic tremors acting in the studied volcanic group are described in Sec. 2. Sec 3
details the method developed to automatically locate tremor sources in 3D. The temporal evolution of those 3D locations is described and discussed in Sec. 3.3 and 4.

#### 125

#### 2 Volcanic Tremors at the Klyuchevskoy Volcanic Group

The Klyuchevskov volcanic group (KVG) is one of the most active subduction-zone 126 volcanic group in the world. It consists of 13 strato-volcanoes located in an approxima-127 tively 70 km diameter zone on the Russian Kamchatka peninsula (N. M. Shapiro et al., 128 2017). Three volcanoes of the KVG (Klyuchevskoy, Bezymianny and Tolbachik) were 129 active during recent decades (N. M. Shapiro et al., 2017; Gordeev et al., 1989; Ozerov 130 et al., 2007; Ivanov, 2008; Senyukov et al., 2009; Senyukov, 2013, e.g.,). In addition, two 131 other very active volcanoes, Shiveluch and Kizimen, are respectively located North and 132 of KVG (Fig. 1). 133

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#### 2.1 Single-station based monitoring of volcanic tremors

The Kamchatka Branch of the Geophysical Survey (KBGS) of the Russian Academy 135 of Sciences operates the permanent seismic network monitoring the KVG and surround-136 ing volcanoes (Gordeev et al., 2006; Chebrov et al., 2013). In this paper, we use four and 137 a half years of seismic data continuously recorded between January 2009 and June 2013 138 by this permanent network composed of 18 stations in 2009 and 19 stations from 2010 139 onwards. The stations and the five monitored volcanoes are shown on Fig. 1. The seis-140 mic stations have three-components, each one being equipped by a CM-3 short period 141 seismometer with a corner frequency of around 0.8 Hz. Continuous records are sampled 142 at 128 Hz and only the vertical component is analysed in this article, studied tremors be-143 ing more energetic in the vertical direction (but horizontal components could also be used 144 as mentioned in Sec. 2.2.1). 145

As many volcanological observatories do all over the world, the KBGS monitors volcanic tremors using a simple single-station approach. For example, stations LGN and

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KMN are used to monitor Klyuchevskoy and Tolbachik tremors respectively (Fig. 1). This
 simple approach suffers from limitations of robustness if the reference station breaks down,
 precision if electronic noise is confused with tremor, and impossibility to spatially locate
 the tremor.

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#### 2.2 Network-based monitoring of volcanic tremors

To overcome the limitations of single-station based approaches, we propose to use a network-based strategy that exploits the coherence of signals simultaneously recorded by several receivers. It is based on the network covariance matrix that can be used to extract coherent signals propagating across the network by reducing local noise (Seydoux, Shapiro, de Rosny, et al., 2016). A detailed explanation of the covariance matrix estimation from real data is presented in Sec. 2.2.1.

Being interested on long-duration tremors and not on short-transient events, all inferences are obtained from day-long network covariance matrices. Once detected through the analysis of the daily network covariance matrix eigenvalues (please refer to Sec. 2.2.2 for more information), different volcanic tremors can be clustered and located by using the daily network covariance matrix first eigenvector (Sec. 2.2.3).

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#### 2.2.1 Estimation of the daily network covariance matrix

We here narrow the focus on the vertical-component records in our analysis, though an extension of this strategy to 3-component data is possible (Wagner & Owens, 1995). We form the spectral network data vector  $\mathbf{u}(f,t)$  from the records of the N seismic stations as

$$\mathbf{u}(f,t) = \left[u_1(f,t), u_2(f,t), \dots, u_N(f,t)\right]^T$$

(1)

where  $u_i(f,t)$  is the Fourier transform of the signal recorded at the station i = 1...Ncalculated over a time window of duration  $\delta t$  and starting at t.

The general workflow of covariance matrix estimation is summarized with the next steps:

1. Seismic data are downsampled from 128 Hz down to 25.6 Hz after bandpass-filtering between 0.1 and 10 Hz in order to allow for faster computation. We consider the

- raw data without correcting for the instrument response, because all the sensorsare identical.
- 2. We apply spectral whitening to the seismic records (independently for every sta-178 tion) in order to enhance the coherence of the tremors signals (Bensen et al., 2007). 179 Spectral whitening is often combined with temporal normalization in order to at-180 tenuate the amplitude of strong impulsive sources (earthquakes). But applied alone, 181 as done here, the spectral whitening has shown to significantly improve the de-182 tection of volcanic tremors at Kamchatka (Soubestre et al., 2018). Indeed, it di-183 minishes the influence of strong earthquakes and local noise and preserves the am-184 plitude modulation of tremor signals, that is a useful information when tremors 185 are not fully stationary in time and rather represent sequences of many impulsive 186 long-period events. 187
- 3. We divide the seismic data into time windows of duration  $\delta t$  from which the Fourier transform is calculated in order to obtain the spectral data vector (Eq. (1)). We also avoid artefacts caused by sharp windows by applying a tapering Hann window; we thus overlap the signal segments with a factor of 50% in order not to loose any information located in consecutive windows.

4. We calculate the covariance matrix from sets of *M* consecutive overlapping segments of data, such as

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$$\mathbf{C}(f,t) = \left\langle \mathbf{u}(f,t)\mathbf{u}^{\dagger}(f,t)\right\rangle_{\Delta t} = \frac{1}{M} \sum_{m=0}^{M-1} \mathbf{u}(f,t+m\delta t/2)\mathbf{u}^{\dagger}(f,t+m\delta t/2).$$
(2)

- 5. The covariance matrix  $\mathbf{C}(f,t)$  is then obtained for each time t and incorporates information about the data collected in the interval  $[t, t + \Delta t]$  where  $\Delta t = (1 + M)\delta t/2$ . We decide to calculate the covariance matrix with a sliding step of  $\Delta t/4$ in order to have an appreciable final temporal resolution in our analysis.
- 6. The daily covariance matrix is finally obtained by averaging consecutive covari-200 ance matrices obtained within one day of data. Note that this windowing and av-201 eraging is necessary because a matrix built from day-long spectra (without win-202 dowing) would be generated from a single data vector. Inherently, such a matrix 203 being of rank one, only a single eigenvector could be obtained. In order to observe 204 the splitting of different sources onto different eigenvectors, we must average the 205 wavefield observations onto smaller windows, allowing to increase the rank of the 206 covariance matrix and to better separate sources. 207

-7-

This daily network covariance matrix being inherently Hermitian and positive semidefinite (Seydoux, Shapiro, de Rosny, et al., 2016), it can be decomposed on the basis of its complex eigenvectors  $\mathbf{v}^{(i)}$  associated with real positive eigenvalues  $\lambda_i$ :

$$\mathbf{C}(f,t) = \sum_{i=1}^{N} \lambda_i(f,t) \mathbf{v}^{(i)}(f,t) \mathbf{v}^{(i)\dagger}(f,t)$$
(3)

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#### 2.2.2 Detection of tremors from the covariance matrix eigenvalues

According to Seydoux, Shapiro, de Rosny, et al. (2016), the number of non-zero eigen-213 values is related to the number of statistically independent signals composing the wave-214 field and then of independent seismic sources. Real seismic data is composed by a large 215 amount of seismic sources ranging from earthquakes, tremors, and more complex sources 216 usually designated as noise sources (hum, oceanic microseisms, wind, anthropic activ-217 ity, scatterers acting like secondary sources, self-sensor electronic noise...). This large num-218 ber of sources are not absolutely independent and may be located onto similar eigenvec-219 tors. Also, the estimation of the covariance matrix from a finite number of windows may 220 also reduce our ability to capture the independent seismic signals. For these reasons, de-221 riving the exact number of independent sources is a challenge (Wax & Kailath, 1985, e.g.,) 222 that is out of scope in the present study. Nevertheless, an estimate of the "level of co-223 herence" of the seismic wavefield can be obtained by calculating the width  $\sigma(t, f)$  of the 224 covariance matrix eigenspectrum (the covariance matrix *spectral width*), a proxy for the 225 presence of coherent signal within the seismic wavefield: 226

$$\sigma(f,t) = \frac{\sum_{i=1}^{N} (i-1)\lambda_i(f,t)}{\sum_{i=1}^{N} \lambda_i(f,t)}$$

$$(4)$$

This spectral width can be seen as a proxy for the number of independent seismic sources. Thus, the spectral width corresponding to ambient seismic noise produced by distributed noise sources is high, whereas the spectral width of a signal spatially coherent at the network scale generated by a single localized source is low.

The spectral width  $\sigma(f, t)$  computed from daily covariance matrices during the 4.5 years-long studied time period is analyzed by Soubestre et al. (2018). This result clearly shows three main long-duration episodes when the spectral width is reduced: (1) in the beginning of 2009, (2) from the end of 2009 to the end of 2010, and (3) from the end of 2012 to mid 2013. From *a priori* knowledge (Droznin et al., 2015), we know that the two first episodes correspond to periods of strong non-eruptive-tremors at Klyuchevskoy and the last one to a strong eruptive tremor at Tolbachik. This approach based on eigenvalues is therefore efficient to detect volcanic tremor, but it is unable to tell whether those
detected tremor sources are from the same volcano or from different ones (here we know
it from *a priori* knowledge). To overcome the limitation of this approach to distinguish
different sources of volcanic tremor, Sec. 2.2.3 focuses on the eigenvectors of the daily
network covariance matrix.

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# 2.2.3 Clustering and location of tremors from the covariance matrix first eigenvector

Instead of using directly the daily network covariance matrix to cluster or locate 246 tremor sources, it has been shown in Soubestre et al. (2018) that better results are ob-247 tained by using the covariance matrix first eigenvector. The idea is derived from prin-248 cipal component analysis, where the first component (eigenvector) is known to represent 249 the most coherent part of the wavefield (Wagner & Owens, 1995; Seydoux, Shapiro, Rosny, 250 & Landès, 2016). Therefore, for days with tremor activity, the first eigenvector charac-251 terizes the dominant tremor source, and acts as a denoising operator because the higher-252 order eigenvectors are orthogonal to the first one, and most likely span information re-253 lated to the background noise. It has to be noticed that such a filtering by the first eigen-254 vector theoretically avoids the possibility of identifying simultaneously active tremor sources 255 from different volcanoes. Being mainly interested by the daily dominating tremor source 256 in the studied region, this is not a problem for our purpose. But further work should be 257 realized to analyze in detail periods when more than one volcano are emitting tremor 258 simultaneously, what is out of scope of the present article. This distinction of simulta-259 neously acting tremors might require to include higher-order eigenvectors of the covari-260 ance matrix in the analysis. Another option could be to consider horizontal components, 261 in order to bring polarization constraint in the analysis (Vidale, 1986). 262

The frequency-dependent first eigenvector of the daily covariance matrix  $\mathbf{V}_1(day, f)$ 263 can be analysed to cluster and/or locate the daily dominant tremor source. Using cor-264 relation coefficient as a measure of their similarity, Soubestre et al. (2018) developed a 265 clustering algorithm to analyze this collection of daily first eigenvectors. The clustering 266 process identifies seven clusters associated with seismo-volcanic activity during the time 267 period considered in the present article. Each cluster's central day eigenvector (highest 268 correlation coefficient) corresponds to the most characteristic first eigenvector of the clus-269 ter. Those characteristic eigenvectors can then be used as templates for template match-270

-9-

ing detection of tremor sources, comparing them with the daily first eigenvectors from the continuous records (Droznin et al., 2015) or with new available data. This clustering approach is interesting to classify different sources of volcanic tremor, but it does not associate the obtained clusters to volcanoes of the studied zone. This location of tremors can be done using the moveout information of the daily first eigenvector, as detailed in the next section.

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### 3 Automatic 3D Location of Volcanic Tremor Sources

Methods of volcanic tremor sources location from network cross-correlations have 278 already been proposed in previous studies (Ballmer et al., 2013; Droznin et al., 2015; Li 279 et al., 2016, 2017; Donaldson et al., 2017; Barriere et al., 2017; Yukutake et al., 2017; Woods 280 et al., 2018, e.g.,), but with an *a priori* on the type of wave emerging in cross-correlations. 281 Because they study shallow tremor sources, the authors assume that cross-correlations 282 are dominated by surface waves. Conversely to those methods considering a surface wave 283 propagation from the tremor source, the cross-correlations based location method de-284 veloped in our article assumes an S-wave propagation and the knowledge of the associ-285 ated S-wave structure, allowing to locate shallow as well as deep tremor sources. Another 286 main difference between the method proposed here and existing ones, is that the con-287 sidered cross-correlations are not directly those of the inter-stations cross-correlations 288 matrix, but those derived from the first eigenvector of the network covariance matrix. 289 This makes the cross-correlations more representative of the dominant tremor, as proved 290 in the supporting information S1 where the location of a synthetic tremor (made of mul-291 tiple long-period events merging) is more precisely recovered using the filtering of the 292 network covariance matrix by its first eigenvector than the full matrix itself. 293

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#### 3.1 Method: 3D location from cross-correlations envelopes

Volcanic tremor sources can be located using the moveout information contained in the daily network covariance matrix first eigenvector, that contains an information about the location and the mechanism of the dominating tremor source. We extract the filtered covariance matrix  $\tilde{\mathbf{C}}(f,t)$  from the complex outer product of the first eigenvector  $\mathbf{v}^{(1)}(f,t)$  with its Hermitian transpose

$$\tilde{\mathbf{C}}(f,t) = \mathbf{v}^{(1)}(f,t)\mathbf{v}^{(1)\dagger}(f,t).$$
(5)

This operation can be seen as a low-rank denoising of the covariance matrix, a strategy widely used in image processing to remove the noise spanned by higher-order eigenvectors (Orchard et al., 2008). The inverse Fourier transform of this matrix retrieves the time-domain filtered cross-correlations:

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$$\mathbf{R}(\tau, t) = \mathcal{F}^{-1}\mathbf{C}(f, t), \tag{6}$$

where  $\mathcal{F}^{-1}$  is the inverse Fourier transform operated to the frequency dimension, and  $\tau$  is the cross-correlation time lag.

To locate the position of the dominant tremor source we use a grid search on a 3D 308 space grid. We compute covariance matrices (5) at frequencies between  $0.5 \,\mathrm{Hz}$  and  $2.0 \,\mathrm{Hz}$ 309 (spectral band where tremors are more energetic in Kamchatka according to Droznin et 310 al. (2015) and Soubestre et al. (2018)). The resulting time-domain cross-correlations are 311 naturally filtered in the same range. We then compute the smooth envelope of the cross-312 correlations  $\mathbf{E}(\tau, t)$  by computing the absolute value of the analytic signal derived from 313 the Hilbert transform h, and performing a convolution with a Gaussian filter q(t) with 314 a 100 seconds width: 315

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$$\mathbf{E}(\tau, t) = g(t) \otimes |\tilde{\mathbf{R}}(\tau, t) + ih(\tilde{\mathbf{R}})(\tau, t)|$$
(7)

where,  $g(t) = \frac{1}{\beta\sqrt{2\pi}} \exp\left(-\frac{1}{2}\frac{t^2}{\beta^2}\right)$  is a Gaussian window with typical width  $\beta = 10$  seconds (larger values have been tested with no significant difference and the  $\beta = 10$  s value was found to be the best lower bound). The smoothing operation allows to reduce the alteration of the location quality from scattering due to local heterogeneities and from the imprecision of the location method itself, particularly the imprecision of the velocity model described hereafter.

Then, for every point  $\mathbf{r} = x\mathbf{e}_x + y\mathbf{e}_y + z\mathbf{e}_z$  of the 3D grid, each cross-correlation envelope  $E_{ij}(\tau)$  is shifted by the time difference  $d\tau_{ij}(\mathbf{r})$  between travel times needed for an S-wave to travel from the tested point to the two stations *i* and *j*:

$$d\tau_{ij}(\mathbf{r}) = \tau_i(\mathbf{r}) - \tau_j(\mathbf{r}) \tag{8}$$

Travel times  $\tau_i(\mathbf{r})$  and  $\tau_j(\mathbf{r})$  are calculated based on the 1D velocity model for S-waves (dominating in volcanic tremors) accounting for velocity variation with depth derived by Droznina et al. (2017), described in supporting information S2. The value at zero lagtime of shifted cross-correlations envelopes is finally stacked for all stations pairs to obtain the network response function and this computation is repeated for every day start $_{332}$  ing at time t:

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$$b(\mathbf{r},t) = \sum_{i=1}^{N} \sum_{j>i}^{N} E_{ij}(-d\tau_{ij}(\mathbf{r}),t)$$
(9)

Note that cross-correlations envelopes being not relatively normalized, the weight in the stack of the network response of each individual cross-correlation envelope is directly related to its quality (spiky shape).

The grid search is performed on a grid covering the whole studied zone down to 50 km depth, with a 1 km  $\times$  1 km  $\times$  1 km resolution. In order to compare relative levels of tremor between different days, each daily network response function  $b(\mathbf{r}, t)$  is normalized first by its integral on the total grid volume V:

$$\ell(\mathbf{r},t) = \frac{b(\mathbf{r},t)}{\int_{V} b(\mathbf{r},t) \mathrm{d}\mathbf{r}}$$
(10)

Then, those spatially normalized response functions are temporally re-normalized by the maximum of the whole time period:

$$\tilde{\ell}(\mathbf{r},t) = \frac{\ell(\mathbf{r},t)}{\max_{\mathbf{r}}\ell(\mathbf{r},t)}$$
(11)

The integral-normalized network response (10) summing to unity on the total grid volume, it can be referred to as a presence likelihood of the tremor source. Similarly, the re-normalized response (11) represents this presence likelihood relatively to the maximum likelihood of the considered period. The most likely daily source position corresponds to the maximum of this function in space.

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#### 3.2 Results: 3D location of volcanic tremors

Fig. 2 shows examples of renormalized network response function (11) for selected 351 days. Most probable daily tremor source location is marked by a black star. This fig-352 ure clearly shows that during the studied time period, volcanic tremors are originating 353 from different volcanoes or from different activity phases of a particular volcano. Thus, 354 Fig. 2a shows a day of quite deep tremor activity beneath Shiveluch volcano. Fig. 2b and 355 Fig. 2c respectively correspond to a deep and a shallow tremor activity beneath Klyuchevskoy. 356 Fig. 2d and Fig. 2f respectively show days of shallow tremor activity beneath Kizimen 357 and Tolbachik. Finally, a response function corresponding to a day during the calm pe-358 riod in 2012 without tremor activity, but maybe dominated by a regional earthquake, 359 appears in Fig. 2e for comparison. For more information, supporting information S3 shows 360 cross-correlations envelopes corresponding to Fig. 2c and Fig. 2e. 361

If some days correspond to shallow tremor sources (Figs. 2c, 2d and 2f), other days 362 such as 2009-088 of Fig. 2a and 2009-276 of Fig. 2b are showing a deep tremor source. 363 This result is quite interesting, because it is the first time that a deep volcanic tremor 364 is located by a cross-correlation based method. It means that during those days, the lo-365 cated volcanic tremor has a source associated with body waves. In the context of the KVG, 366 those body waves are emitted by volcanic tremor sources located in the consolidated bedrock 367 under unconsolidated sedimentary or volcanic deposits. Note that such an emergence of 368 body waves in cross-correlations has already been observed in other different contexts 369 and scales, such as body waves propagating through the active magmatic system of the 370 Piton de la Fournaise volcano (Nakata et al., 2016), body waves reflected from the Moho 371 (Poli, Pedersen, & Campillo, 2012) or body waves from the mantle transition zone (Poli, 372 Campillo, et al., 2012). 373

374

#### 3.3 Results: Time evolution of volcanic tremor sources 3D location

The issue of the vertical migration of volcanic tremor sources with time is addressed 375 here. The tremor presence likelihood (11) is daily analyzed beneath each studied vol-376 cano (Shiveluch, Klyuchevskoy, Tolbachik and Kizimen from North to South) to check 377 for the source depth. To do so, a circular zone of a given radius is defined around each 378 volcano (the zone is centered on volcanoes summit as appearing on Fig. 1) and the max-379 imum value of the likelihood (11) over those circular zones is selected at every depth and 380 every day. The expected value  $\bar{z}$  of the likelihood for each zone and each time is calcu-381 lated with  $\bar{z}(t) = \int \max_{x,y} \tilde{\ell}(\sqrt{x^2 + y^2} < r, z, t) z dz$  and represented as a black dot in 382 order to better visualize the expected position of the tremor sources. Fig. 3 shows the 383 result of such an analysis for 25 km radius zones, as shown in Fig. 1. Note that the like-384 lihood has been clipped below a value of 0.7 in order to narrow the focus on the highly-385 coherent time segments. 386

This time-depth analysis clearly enhances the same time periods of tremor activity as the detection of tremors by Droznin et al. (2015) and Soubestre et al. (2018), but with the benefit that here each tremor is clearly associated to a specific volcano. Disregarding the depth information and looking only at the time axis of Fig. 3, this result showing when each volcano is under tremor is interesting from a monitoring point of view and this approach could be usefully run in real-time in a volcanological observatory. Thus, volcanic tremor activity is automatically located at Shiveluch volcano in March 2009,

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at Klycuchevskoy volcano at the beginning of 2009 and also from September 2009 to the 394 end of 2010, at Kizimen during two episodes of one month in June and September 2011, 395 and at Tolbachik volcano from the end of November 2012 onwards. We also see that dur-396 ing periods of volcanic rest (e.g. March 2011) the likelihood presence is the highest at 397 great depth, and is most likely related to strong earthquakes that resisted our prepro-398 cessing; this is mostly due to the presence of dominating body waves in the seismic wave-399 field at global scales usually observed after big events (Boué et al., 2014; Sens-Schönfelder 400 et al., 2015). For example in this period, the M9.1 Tohoku earthquake (Ozawa et al., 2011) 401 is clearly visible on our results for several days (Fig. 3, beneath the Tolbachik inset). 402

Moreover, the depth of the daily dominating tremor source is made better visible 403 by Fig. 3. Looking at the spatial distribution of volcanoes and seismic stations in the 404 map of Fig. 1, one understands that the source depth is not well constrained at volca-405 noes Shiveluch and Kizimen that are located at the border of the studied zone and count 406 with a few stations. Likewise, Tolbachik volcano is more central in the map but it counts 407 only one station close to it, so that the depth estimation of tremor sources beneath Tol-408 bachik cannot be very precise. Fig. 3 shows that Tolbachik's tremor is an eruptive tremor, 409 beginning the 27th of November 2012 one day after the official beginning of the fissure 410 eruption that started the 26th of November. Looking at Fig. 2f, the tremor source seems 411 to be superficial. This result is consistent with the analysis from Caudron et al. (2015) 412 who relate this eruptive tremor to lava flows. This is not surprising for this tremor to 413 be superficial given that the seismic preparatory phase of Tolbachik's 2012-2013 erup-414 tion was also shallow. Indeed, the eruption was preceded by 4-5 months of shallow (depth 415  $< 5 \,\mathrm{km}$ ) and low-energy volcano-tectonic seismicity (Kugaenko et al., 2015), as well as 416 shallow long-period events clustered in time due to multiple cracks and channels reac-417 tivation (W. B. Frank et al., 2018). The eruption is supposed to have been fed by magma 418 ascending from a shallow crustal magma storage (Fedotov et al., 2010, 2011). 419

On the other hand, the depth of tremor sources is better constrained for Klyuchevskoy volcano that is central in the map and count multiple stations around it. Moreover, the plumbing system of Klyuchevskoy is known to be more complex and to contain deeper levels of magma storage, as evidenced by different types of studies. Thus, Koulakov et al. (2011) imaged a crustal storage beneath Klyucevskoy at intermediate depth of 8-13 km and a deeper one in the mantle below 25 km depth. This latter deep storage was also identified by analysis of deep long-period earthquakes (N. Shapiro et al., 2017). In Fig. 3,

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the 2009-2010 pre-eruptive tremor of Klyuchevskoy volcano seems to affect shallow (depth 427  $< 5 \,\mathrm{km}$ , medium (5 km  $< \mathrm{depth} < 15 \,\mathrm{km}$ ) as well as deeper depths of about 20-25 km 428 at the beginning of the time period. The tremor shows different phases from Septem-429 ber 2009 to December 2010. First, from September to December 2009, the expected po-430 sition of the tremor source (black dots in Fig. 3) progressively migrates from deep depths 431 of about 25 km to medium depths of 10 km. Then, the tremor activity seems to stabi-432 lize around 10 km from January to June 2010, when it starts to focus to shallower depths 433 of a few kilometres. At the beginning of July 2010 the expected position of the tremor 434 source starts to move deeper again down to 15 km, where it stabilizes up to the end of 435 October 2010. This change could be interpreted as a possible reinjection of magma from 436 a deep reservoir. And eventually, from November 2010 onwards the dominant tremor source 437 progressively migrates to the surface, a few months before the beginning of the summit 438 eruption that started in December 2010. This evidence of the sequence of dominant tremor 439 source depth decrease with time getting closer to the eruption is an interesting result con-440 sistent with the dynamics of magmatic eruption processes. 441

#### 442 **4** Conclusion

In this paper, we developed a cross-correlations based method to automatically lo-443 cate volcanic tremor sources in 3D. The method consists in a first preprocessing step to 444 normalize the data and calculate the daily network covariance matrix, a second step of 445 filtering this matrix by its first eigenvector to retrieve the daily dominant tremor source, 446 and a last step to locate this source from the moveout information of the first eigenvec-447 tor. An important aspect of the developed method is that once the S-wave velocity model 448 and all parameters selected, the data analysis and the following detection and location 449 of tremor sources are fully automatic without need of additional *a priori* information. 450 A main difference between the developed cross-correlations based location method of vol-451 canic tremor and other existing ones, is that instead of assuming a surface wave type prop-452 agation and a shallow origin of the tremor, we estimate the tremor source depth and we 453 are able to locate shallow as well as deep tremors. Another main difference is that the 454 considered cross-correlations are not directly those of the inter-stations cross-correlations 455 matrix, but those derived from the first eigenvector of the network covariance matrix that 456 characterizes the dominant component of the recorded wavefield. 457

The daily application of the developed method onto four and a half years of data 458 (from January 2009 to June 2013) from the Klyuchevskoy group of volcanoes allowed 459 to daily locate the dominant volcanic tremor source, what is interesting from the mon-460 itoring point of view of a volcanological observatory. It also enabled to track the tem-461 poral evolution of the 3D location of the pre-eruptive volcanic tremor beneath Klyuchevskoy 462 from deep to shallow parts of the plumbing system, from September 2009 to December 463 2010 when the summit eruption started. This latter result is interesting, first because 464 to our knowledge it is the first time a deep volcanic tremor source is located with a cross-465 correlation based method. Second, because the evolution of the tremor source depth dur-466 ing one year and half including pre and co-eruptive activity is quite nicely recovered. Ac-467 cording to the analysis, initially multiple tremor sources are acting simultaneously from 468 the surface to depths down to 20-25 km, and then the dominating tremor source progres-469 sively migrates to the surface. This surface migration pattern is observed twice during 470 the 2009-2010 tremor of Klyuchevskoy, first from September 2009 to the end of June 2010 471 and second from mid July to December 2010 after a possible re-injection in July-August. 472 This behavior of volcanic tremor sources affecting a wide depth range and then migrat-473 ing to the surface can be explained by the model proposed by Stasiuk et al. (1993) of 474 progressive magma draining from the bottom of vertical pathways during eruptions. 475

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Figure 1. Seismic network (stations: black inverted triangles) monitoring the Klyuchevskoy volcanic group and surrounding volcanoes (white triangles). The Russian Kamchatcka peninsula is located by a red square on the Earth globe. The 25 km radius circular zones used for the analysis of the vertical migration of volcanic tremor sources with time addressed in Sec. 3.3 appear as red zones on the map. Note that those circular red zones appear as ellipses here because of the Miller projection (a modified Mercator projection) used for drawing the map.



Figure 2. Volcanic tremor source presence likelihood (corresponding to (11)) corresponding to six different days (red dot in top-left black curves). On each response function, the most probable tremor source location is marked by a black star. Each response function appears as a top-view of its values in the horizontal plane corresponding to the depth of the tremor source (center panel), and two side-views of its values in vertical planes corresponding respectively to the depth-longitude plane (bottom panel) and the depth-latitude plane (right panel) of the tremor source. Seismic stations and volcanoes respectively appear as black inverted triangles and with triangles in the horizontal top-view (center panel), and first three initial letters of each volcano help the reader to orient. a) Quite deep tremor beneath Shiveluch on day 2009-088. b) Deep tremor be--27- neath Klyuchevskoy on day 2009-276. c) Shallow tremor beneath Klyuchevskoy on day 2010-273.



Figure 3. Time evolution of tremor sources depth beneath each studied volcano (Shiveluch, Klyuchevskoy, Tolbachik and Kizimen from North to South). A circular zone of radius r = 25 km is defined around each volcano (the zone is centered on volcanoes summit as appearing on Fig. 1) and the maximum value of the likelihood (11) over those circular zones is selected at every depth and every day. The expected value  $\bar{z}$  of the likelihood for each zone and each time is calculated with  $\bar{z}(t) = \int \max_{x,y} \tilde{\ell}(\sqrt{x^2 + y^2} < r, z, t)zdz$  and represented as a black dot in order to better visualize the expected position of the tremor sources. The time of the 2011 M9.1 Tohoku earthquake (Ozawa et al., 2011) is indicated with a black star beneath the results from the Tolbachik volcano.

Article Figure - Fig 1.