# Mechanical tomography of a volcano plumbing system from GNSS unsupervised modeling

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#### Abstract

Identification of internal structures in an active volcano is mandatory to quantify the physical processes preceding eruptions. We propose a fully unsupervised Bayesian inversion method that uses the point compound dislocation model as a complex source of deformation, to dynamically identify the substructures activated during magma migration. We applied this method at Piton de la Fournaise. Using 7-day moving trends of GNSS data preceding the June 2014 eruption, we compute a total of 15 inversion models of 2.5 million forward problems each, without a priori information. Obtained source shapes (dikes, prolate ellipsoids or pipes) exhibit a global migration from 7-8 km depth to the surface, drawing a "mechanical tomography"? of the plumbing system. Our results allow retrieving geometries compatible with observed eruptive fissures and seismicity distribution, and the retrieved source volume variations made this method a good proxy to anticipate erupted lava in case of no co-eruptive refilling.

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#### Key Points:

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10	•	imaging volcano plumbing system from geodesy and mechanical modeling
11	•	detecting precursory magma migration and anticipating real erupted volume
12	•	new unsupervised real-time modeling tool for volcano monitoring

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#### 13 Abstract

Identification of internal structures in an active volcano is mandatory to quantify the phys-14 ical processes preceding eruptions. We propose a fully unsupervised Bayesian inversion 15 method that uses the point compound dislocation model as a complex source of defor-16 mation, to dynamically identify the substructures activated during magma migration. 17 We applied this method at Piton de la Fournaise. Using 7-day moving trends of GNSS 18 data preceding the June 2014 eruption, we compute a total of 15 inversion models of 2.5 19 million forward problems each, without a priori information. Obtained source shapes (dikes, 20 prolate ellipsoids or pipes) exhibit a global migration from 7-8 km depth to the surface, 21 drawing a "mechanical tomography" of the plumbing system. Our results allow retriev-22 ing geometries compatible with observed eruptive fissures and seismicity distribution, 23 and the retrieved source volume variations made this method a good proxy to anticipate 24 erupted lava in case of no co-eruptive refilling. 25

#### <sup>26</sup> Plain Language Summary

Imaging the interior of an active volcano and estimating volumes of magma in depth are major challenges of eruption anticipation and forecast. In this work we propose an effective method of data processing that combines a new analytical model of theoretical source, and standard ground deformation measurements, in a fully automated process. The method is sensitive to magma migration and behaves like a scanner that displays a 3D image of the volcano plumbing system.

#### 33 1 Introduction

Active volcano edifices might deform due to fluid migration and storage into their 34 so-called plumbing system, an interconnected network of internal volumetric substruc-35 tures like reservoirs, conduits or sills/dikes (Tibaldi, 2015). Indeed, fluid dynamics into 36 the plumbing system involves mechanical constraints (pressure, volume or stress vari-37 ations) that are applied on the internal boundaries of the medium, inducing deforma-38 tion and displacements that usually reach the free surface. This behavior highly depends 39 on the medium rheology (Sparks et al., 2019), and deformation intensity can sometimes 40 be much below the instrumental detection capability. Yet, monitoring volcano deforma-41 tion has been commonly used for more than half a century to detect the subtle warn-42 ing signals of a volcanic eruption linked to the pressurization of magma body or magma 43 transfers at depth (see Dzurisin (2003) for a complete review and Segall (2010) for lim-44 itations of the deformation methods). In this context, the characterization of the magma 45 feeding system (location, volume, shape, etc.) with short-term and reliable quantitative 46 parameters is an important prerequisite for understanding and anticipating any erup-47 tive activity. Inversion of geodetic data with mechanical models has natural capability 48 to locate the pressure source in depth and quantify its characteristics from surface ob-49 servations (see for instance Toutain et al. (1992); Cayol and Cornet (1998); Beauducel 50 et al. (2004); Anderson et al. (2010); Peltier et al. (2016)). Moreover, any quantitative 51 volcano model needs boundary conditions, in particular those common to the magma 52 fluid dynamics and the volcano mechanical behavior, i.e., the plumbing system geom-53 etry. Imaging these structures using various tomography methods has the main goal of describing, in a more quantitative way than any geological approach, the internal struc-55 tures, which might be used in other geophysical or geochemical dynamic modeling as a 56 priori information. 57

We propose, in this paper, to follow the magma circulation and/or accumulation by locating and quantifying pressurisation sources in space and time using unsupervised deformation source modeling from GNSS observations at Piton de la Fournaise (PdF). We introduce here the innovative term of "mechanical tomography", since the method uses magma ascent as an active source that progressively "illuminates" the complex magma plumbing system, and finally gives an image of the internal substructure geometries, which
have been activated during the PdF unrest.

PdF (La Réunion Island, Indian Ocean, Figure 1a) is an active basaltic volcano, often in eruption with an average of 2 eruptions per year since 1979, date of the creation of the Observatoire Volcanologique du Piton de la Fournaise (OVPF) from the Institut de Physique du Globe de Paris. The recent eruptive activity mainly occurs inside an uninhabited caldera, called Enclos Fouqué, where a terminal cone topped by two craters (Bory and Dolomieu), gradually built up (Figure 1b).

The location and the shape of the shallow magma reservoirs below the volcano are 71 still debated even if recent geodetic, seismic and geochemical studies converge on a global 72 scheme of a plumbing system constituted of several reservoirs, variably connected and 73 distributed from 10 km depth to the near-surface (Battaglia et al., 2005; Peltier et al., 74 2009; Di Muro et al., 2014; Boudoire et al., 2019). Passive S and P-wave tomographies 75 made on PdF, using ambient seismic noise and P-wave first arrival times for earthquakes, 76 respectively, show 1) a high S-wave velocity zone from -1 to 1.9 km below the terminal 77 cone interpreted as a preferential paths for magma injections (Brenguier et al., 2007) and 78 2) a high-velocity plug at sea level, under the summit craters, interpreted as an intru-79 sive, solidified dike-and-sill complex with little fluid magma storage (Prôno et al., 2009). 80 Two low P-wave velocity anomalies, which may highlight magma reservoirs, are found 81 from 0 to 1 km a.s.l. and from 1 to 2 km b.s.l. (Prôno et al., 2009). At greater depth, 82 spatio-temporal distribution of the seismicity located by OVPF may evidence the pres-83 ence of a deeper reservoir at around 7.5 km depth below the Bory crater (Battaglia et 84 al., 2005; Peltier et al., 2009). 85

In June, 2014, after an unusual period of 41 months of dormancy, PdF showed signs 86 of unrest with the start of a slow edifice inflation and an increase of the shallow (j2 km)87 depth) seismic activity on June 9. Two seismic crises (not associated with rapid ground 88 deformation) occurred on June 13 and 17, with 360 and 687 shallow volcano-tectonic earth-89 quakes, respectively. A last seismic crisis that lasted one hour and 16 minutes (888 shal-90 low volcano-tectonic earthquakes located between 0.3 and 1.5 km a.s.l. below the Dolomieu 91 crater; associated with rapid ground deformation) led to an eruption on June 20, 21:35 92 (UTC time). The eruptive fissures opened on the external and south south-eastern slope 93 of the Dolomieu crater (2348-2480 m elevation; Figure 1b). Eruptive activity ended on 94 June 21, 17:09 (UTC time) and emitted about  $0.4\pm0.2$  million m<sup>3</sup> of lava flows (no DRE). 95

#### 96 2 Methods

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#### 2.1 GNSS Data Processing

Among other multidisciplinary networks, OVPF maintains 24 permanent GNSS 98 stations, one of the densest networks on an active volcano. For this study, we only used qq the daily solutions of the 10 stations within a 4 km radius from the summit (Figure 1b), 100 where ground displacements have been significant during the June 2014 unrest (Figure 101 1c). Data shown in the paper were processed using PPP method by the Gipsy-Oasis soft-102 ware (Desai et al., 2014) providing daily solutions in the ITRF2008 referential, with typ-103 ical standard deviations of 5.2 mm, 4.8 mm, and 11.2 mm for eastern, northern and ver-104 tical components, respectively, over 5 years period at a stable station in La Réunion. Hor-105 izontal tectonic motion has been removed from the time series using linear trend values 106 of +17.9 mm/yr and +12 mm/yr for eastern and northern components, respectively. 107

In order to increase the signal to noise ratio of GNSS observations, we compute displacement trends over a 7 days moving-window, which represents a good compromise between the constraint of the source shape and a detailed time tracking of the source. Indeed, as the GNSS daily solutions have relatively high errors, the computation of a linear trend over a few days' sliding sample window increases the sensitivity to detect subtle signals below the error level of individual daily solutions. Typically, the error on a 7-day velocity trend is as low as about 0.14 mm/day, i.e., only 0.8 mm on the displacement.

#### 116 2.2 pCDM Method

The point compound dislocation model (pCDM) has been proposed by Nikkhoo 117 et al. (2016). It provides analytical expressions for surface displacements due to a source 118 composed of three mutually orthogonal tensile point dislocations, one horizontal and two 119 vertical, freely oriented in space (three rotational degrees of freedom around each 3D axis) 120 in an elastic homogeneous half-space. Original equations depend on nine source param-121 eters: three for the hypocenter location (horizontal coordinates and depth), three vol-122 ume variations  $dV_X$ ,  $dV_Y$  and  $dV_Z$  (of the same sign, for each plane perpendicular to its 123 axis), and three for the angles of rotations  $\Omega X$ ,  $\Omega Y$ ,  $\Omega Z$  (see Figure 2). A tenth param-124 eter is the Poisson's ratio that we fixed to 0.25 to consider an isotropic medium. Since 125 equations use the volume dislocation for the deformation source and not the pressure, 126 the model is independent from other elastic parameters. 127

In order to express the total volume dislocation  $\Delta V$ , an easier quantitative parameter for interpretation, we substituted the three volume variations variables with their total value plus two dimensionless shape ratios between 0 and 1, defined as follows:

$$\Delta V = dV_X + dV_Y + dV_Z, \tag{1}$$

$$A = \frac{dV_Z}{\Delta V},\tag{2}$$

$$B = \frac{dV_Y}{dV_X + dV_Y},\tag{3}$$

where  $\Delta V$  is the total volume variation of the source, A is the horizontal over total volume variation ratio, and B is the vertical volume variation ratio.

The pCDM is able to approximate any shape of magma bodies, as dikes, sills, oblate, prolate and other triaxial ellipsoidal shapes (see some examples with corresponding A and B values in Figure 2), and is only relevant at far-field observation points because of the point source approximation. Even if simple, this model is particularly well adapted for real-time monitoring as it gives a first order estimation of the magnitude and shape characteristics of the source(s) at the origin of the surface displacements, and is still easy to implement in an inverse problem.

We also rewrote the original pCDM code in a fully vectorized way (Matlab/GNU 140 Octave and C languages) in order to make it compatible with fast inversion and millions 141 of forward problems. Furthermore, vectorization allows using equations with the varying-142 depth formulation to approximate the topographic effects (Williams & Wadge, 1998; Beaudu-143 cel & Carbone, 2015), i.e., adjusting the source depth at each observation point using 144 station elevation above sea level. This method is also a good way to solve the eternal prob-145 lem of source elevation referencing in half-space models: here the source depths are given 146 in meters b.s.l.. 147

#### 2.3 Inverse Problem

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In order to obtain a description as objective as possible of the volcano internal structures at the origin of the surface displacements, we minimize the a priori information and explore the entire space of the nine model parameters using the GNSS trends as observation data. As a first result of this unsupervised inversion method, we represent the model space probability as a function of source location, in order to display all solutions that are consistent with observations (Tarantola, 2006). Identification of a single volume zone
with higher probabilities confirms the existence of a cluster of good models, a manda tory condition to possibly select one "best model".

Since exhaustive grid exploration of the full model space is not reasonable with 9 157 parameters, we use a Monte Carlo Bayesian parallel algorithm with 5 iterations, each 158 using 500,000 forward problems randomly chosen in the 9-parameters model space, and 159 possible reduction of any parameter search interval between iterations. The aim of this 160 algorithm is to drastically reduce the relevant range of variation for each parameter, with 161 162 a low probability to exclude the best model. Each iteration performs uniform sampling of the 9 parameters except for horizontal position for which we use a normally distributed 163 sampling centered at the summit with a 5-km standard deviation radius. This a priori 164 information improves the inversion performance as we are expecting source in the neigh-165 borhood of the summit area, but does not exclude any possible distant nor deep source 166 location. Then it computes the misfits associated with the created models, using the L1 167 norm between observed and computed data, and draws the curve representing a proxy 168 of the best possible misfit as a function of the parameter value. If the misfit distribu-169 tion has a single significant maximum probability mode, the process selects a smaller range 170 for this parameter with higher probability to constrain the best models. This new in-171 terval will be used as a starting point for the next iteration. A posteriori uncertainties 172 of the best model solution are given by the interval of variation over each parameter that 173 keeps 68% (one standard deviation) of the highest model probabilities for all the iter-174 ations, a total of 2.5 millions forward models. 175

#### 176 **3 Results**

The long-term pre-eruptive edifice inflation is often of very low intensity at PdF 177 (often less than 5 cm; e.g. in (Peltier et al., 2016, 2018), that is why we look at baseline 178 changes (i.e. linear distance between pairs of stations) to better highlight changes in the 179 deformation trend. After 41 months of rest and slow deflation, first signs of edifice in-180 flation at PdF appeared on June 9, 2014 (Figure 1c,d), and accelerated after June 13, 181 with a summit extension well visible on the DSRG-SNEG baseline (see dark red lines 182 in Figure 1d). Intensity of the ground deformation preceding the June 20, 2014 erup-183 tion remained particularly low, i.e. less than 1 cm of horizontal cumulative displacements 184 and about 2 cm maximum of vertical cumulative displacements recorded on the sum-185 mit stations in 11 days. These low intensity of surface observations makes this eruption 186 a good case study for developing sensitive modeling methods. 187

Results of the inversion modeling on 15 periods (12 before and 3 after the eruption) of 7 day sliding sample window allow retrieving the position and the shape of the pressure source at different times. Figure 3 shows results of each inversion as a full description of the model space probabilities in horizontal and vertical projections (see also Table S1 in the supporting information). Three distinct pre-intrusion phases (phases 1-3) before the final dike propagation to the surface (phase 4) can be distinguished.

1) For the periods spanning June 2-8, 3-9, 4-10 and 5-11, no well-constrained source 195 can be found but probable deep deflation diffuse sources seem to be present below the 196 terminal cone.

2) Inflation sources appeared and became more consistent from the June 6-12 period with a narrower range of models and a best model in inflation located at 4.5 km below sea level, i.e. about 7 km below the summit, with a tilted dike shape. The inflation pressure source, the shape of which evolves from a dike to a pipe, remained deep (0 to 3 km below sea level) until June 16.

3) From the June 11-17 to the June 13-19 periods, when the deformation rate accelerated, the inflation source was shallower, located between 0.7 and 1 km a.s.l. (i.e. between 1.8 and 1.5 km below the summit). The last pressure source modelled before the eruption, for the period spanning June 13 to June 19, displayed a volume variation of  $+210,000 \text{ m}^3$ .

4) On June 20, a rapid (1 hour and 16 min of seismic crisis with rapid ground de-207 formation; Figure 1c) and final magma dike injection propagated to the surface and fed 208 the eruption. The shape of the dike appeared in our models only from the June 16-22 209 sliding-window. For the two previous periods including the beginning of the eruption (June 210 14-20 and June 15-21), the best models were ellipsoid sources, probably because of the 211 212 influence of two sources (the pre-eruptive source and the final dike reaching the surface) associated with the integrating effect of the 7-day trend calculation. Volume variations 213 for the two last co-eruptive periods (15-21) and (16-22) were relatively constant, with 214 values of  $+230,000 \text{ m}^3$  and  $+300,000 \text{ m}^3$ , respectively. 215

#### **4** Discussion

Even with very small ground displacements (less than 1 cm), we are able to image the refilling of the shallow magma plumbing system preceding the June 2014 eruption at PdF. Our previous attempts to detect pre-eruptive magma migration using deformation data may have failed because of the use of too simple isotropic sources as primary models (Beauducel et al., 2014). Success of the inversion using more complex sources may evidence the more frequent deformation sources with flat or elongated shape, like dikes or pipes at PdF.

Our results highlight from June 12 overpressures inside this system at decreasing 224 depth with time (Figure 4), from 7 - 8 km b.s.l. (at the lower limit of our model space) 225 where a reservoir has already been suggested by seismicity in 1998 (Battaglia et al., 2005), 226 to 1.5 km a.s.l., where the shallower reservoir is supposed to lie (1.3 - 1.9 km depth be-)227 low the summit (Peltier et al., 2016)). Most of the pressure sources below sea level are 228 vertically elongated and seems to highlight the volcano deep conduit connecting the two 229 reservoirs, forming a continuum more or less filled by fluids (Figures 3,4). Rather than 230 following a clear magma migration, which may have required a higher time sampling fre-231 quency, we were able to deliver a mechanical tomography of the PdF plumbing system 232 (from about 10 km depth to the surface; Figure 4). The synthetic 3D view of the dif-233 ferent sources identified during the June 2014 pre-eruptive unrest evidences a gap at sea 234 level, at the same level where Prôno et al. (2009) describe high-velocity plug interpreted 235 as a solidified complex with little fluid storage, and where Battaglia et al. (2005) describe 236 a discontinuity in the upward migration of the seismicity preceding the 1998 eruption, 237 which occurred after 6 years of rest. In June 2014, no deep seismicity was recorded dur-238 ing the upward magma migration. Following the 1998 eruption, most of the deep recharges 239 were not accompanied by deep seismicity, and the majority of the earthquakes are lo-240 cated above sea level (Lengliné et al., 2016; Duputel et al., 2019). This is the sign of a 241 more or less deep open conduit, which fed the 34 eruptions that occurred between 1998 242 and 2014 (Roult et al., 2012). Our results show thus the importance of the method we 243 used, which makes it possible to see what seismology does not when the system is already 244 open. Thus, the first models (phase 1) show deep deflation sources before the deep magma 245 migration starts (Figures 3,4). This volume loss might correspond to the emptying of 246 a deeper reservoir(s) before the magma starts to "drill" and follows a path to shallower 247 levels (phase 2; Figure 4). Source locations during phase 3 suggests that the fracturing 248 that allows magma to reach the surface (phase 4) started from about 1.5 km depth be-249 low the southern border of the Dolomieu crater (Figures 3,4). The two seismic crises, 250 on June 13 and 17, with earthquakes above sea level, show already shallow pressure source(s) 251 at that time. 252

The estimated source volume variations are relatively stable during phases 3 and 4, i.e. around 0.25 Mm<sup>3</sup> (see Figure 3) despite the displacement increase over one order

of magnitude between pre- and co-eruptive periods. This volume stability makes sense 255 as it might translate a finite volume of magma involved in the last magma migration pro-256 cess. However, when using an isotropic source this process cannot be properly modeled, 257 as for a given depth, the volume variation must be proportional to surface displacements. 258 Using pCDM, the source shape has the capability to be adjusted while keeping a con-259 stant volume variation and maintaining a shallow depth. In addition, the final volume 260 variation is of the same order of magnitude as the one obtained using an isotropic source 261 for the whole co-eruptive period  $(130.000 \text{ to } 190.000 \text{ m}^3 \text{ (Peltier et al., 2016)})$ , and close 262 to the real erupted volume of  $0.4\pm0.2$  Mm<sup>3</sup>, i.e. a DRE volume of 0.17-0.28 assuming 263 a porosity range of 30–58% (Di Muro et al., 2014). 264

In spite of the elastic and homogeneous assumptions of the models, adequacy of the volume variation obtained from deformation and erupted volume seems consistent with the case of short eruptions during which no long-term refilling occurred as for the June 2014 eruption. We demonstrate with our modelling the need of looking at shorter time periods to evidence migration processes and complex internal shallow structures. GNSS daily solutions are certainly a limitation in this context, and higher frequency may help in identification of finer structures.

#### <sup>272</sup> 5 Conclusion

The low intensity of surface observations preceding the June 20, 2014 eruption at 273 PdF makes this eruption a good case study for developing and validating sensitive mod-274 eling methods. Our work provides good insights into the refilling of the shallow magma 275 reservoir the days preceding the eruption. The pCDM method we used allows tracking 276 the gradual migration of the magma to the upper reservoir, and the final dike propaga-277 tion to the surface, by discriminating both the shape, location and volume of the source. 278 The rewriting of the original pCDM code in a fully vectorized way allows for fast inver-279 sion and is easy to implement to give first-order modelling results, helpful notably for 280 crisis management. With similar results as obtained at Mt. Etna by Cannavò et al. (2015), 281 our method is fast and fully unsupervised, without a priori information on the source 282 parameters except the choice of the pCDM itself. In view of these promising results, we 283 implemented the method as an extension of the GNSS module in the WebObs system, 284 an integrated web-based system for data monitoring and network management, imple-285 mented in 15 observatories worldwide (Beauducel et al., 2020). The module was initially 286 developed with a simple isotropic point source (Beauducel et al., 2014, 2019) and we added 287 the possibility of setting a pCDM source and associated parameters for real-time mod-288 eling. This has been especially useful during recent crisis managements (Moretti et al., 289 2020; Peltier et al., 2020). 290

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Figure 1. a) Location map of La Réunion Island; b) Zoom on the most active part of Piton de la Fournaise and summit craters. GNSS permanent stations (black triangles), June 2014 lava flows (solid red patch), and selected baselines (color solid lines) are shown; c) baseline variations (i.e. distance changes between pairs of stations) on the pre-, co-, and post-eruptive periods (same colors as in b); d) zoom on the pre-eruptive precursory baseline variations. Grey area indicates eruption time. Topographic data from ETOPO1, SRTM, SHOM, and RGEALTI © IGN 2016.



**Figure 2.** pCDM dislocation plans and rotation angles definition. Example of source shapes and the associated A and B values. Surface of each point dislocation is enlarged to be proportional to its associated volume variation.

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**Figure 3.** Temporal evolution of the daily pCDM source solution from Bayesian inversion of 7-day displacement trends from June 8 to 22, 2014, in map and vertical cross-section views. Time interval of each model is given in the figure. Color map indicates the maximum probability level combined with the volume variation sign (yellow-orange-red for inflation, green-cyan-blue for deflation). Black, red and green arrows are observed displacements, modeled displacements and residual, respectively. Ellipses are errors. Best model source location and shape are indicated as grey plans, and their source approximate shape (E: Ellipsoid, S: Sill, P: Pipe, D: Dike) and volume variation (Mm<sup>3</sup>) are indicated.



**Figure 4.** Synthetic and virtual 3D view of the different best models identified during the June 2014 pre-eruptive unrest at Piton de la Fournaise. Colors stand for the most recent date of each time window. Size of each source is proportional to its associated volume variation.

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### Supporting Information for "Mechanical tomography of a volcano plumbing system from GNSS unsupervised modeling"

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#### Contents of this file

1. Table S1

#### Additional Supporting Information (Files uploaded separately)

1. Caption for Movie S1

#### Introduction

Table S1 shows all parameters of the best models obtained for the 15 periods of time as used to produce the Figure 4 of the main paper. These models correspond to the best

June 19, 2020, 2:50am

Movie S1 shows best models in 3D perspective view.

#### Movie S1.

The short 1 minute length movie shows the best models sources in a 3D perspective view with illuminated topography. The camera position makes two loops around the volume with an additional sinusoidal vertical movement. In the main text, the Figure 4 is a single view selection from the same representation.

June 19, 2020, 2:50am

**Table S1.** Summary of the best model deformation sources: time period (days of June 2014), source type (deflation '-' or inflation '+'), source approximate shape as ellipsoid (E), sill (S), pipe (P) or dike (D) and main orientation, coordinates (latitude South, longitude East, in degree), depth b.s.l. (in km), volume variation  $\Delta V$  (in Mm<sup>3</sup>), source geometry ( $A, B, \Omega X, \Omega Y$ ,  $\Omega Z$ ) and global misfit of the best-fit models at 68.3% confidence for each period we defined.

Time	Source	Source	Lat. S	Lon. E	Depth	$\Delta V$	A	В	$\Omega X$	$\Omega Y$	$\Omega Z$	Misfit
Period	Type	Shape	(deg)	(deg)	$(\mathrm{km})$	$(\mathrm{Mm}^3)$			(deg)	(deg)	(deg)	(mm)
02-08	_	Vert. E	21.244	55.756	$+4.1\pm0.7$	$-9.5\pm0.8$	0.07	0.87	-13	-15	+32	5.3
03-09	_	Vert. P	21.246	55.721	$+7.5\pm0.4$	$-7.1\pm0.7$	0.04	0.56	+36	-2	+38	1.8
04-10	_	Vert. P	21.248	55.723	$+3.7\pm0.6$	$-8.2\pm0.8$	0.10	0.47	+4	+10	-15	1.9
05-11	_	Vert. E	21.239	55.726	$+3.0\pm0.3$	$-6.7\pm0.9$	0.02	0.70	-21	+4	+34	2.2
06-12	+	Tilt. D	21.239	55.709	$+4.5\pm0.6$	$+3.2\pm0.7$	0.93	0.83	-37	+43	-7	4.5
07-13	+	Hori. S	21.240	55.713	$-0.6\pm0.7$	$+0.15\pm0.3$	0.99	0.34	-4	+19	-10	3.6
08-14	+	Hori. S	21.238	55.712	$+2.9\pm0.8$	$+1.4\pm0.7$	0.97	0.40	+26	+24	+19	3.3
09-15	+	Vert. P	21.245	55.715	$+3.1\pm0.6$	$+5.3\pm1.0$	0.07	0.50	-29	-10	+19	2.3
10-16	+	Vert. P	21.258	55.718	$+2.5\pm0.4$	$+6.4\pm1.0$	0.07	0.44	+4	-1	-9	2.4
11-17	+	Vert. P	21.250	55.710	$-0.7\pm0.6$	$+0.23\pm1.0$	0.05	0.55	-20	-9	-12	2.1
12-18	+	Vert. P	21.247	55.711	$-1.0\pm0.2$	$+0.19\pm0.01$	0.06	0.46	-12	-2	+27	1.9
13-19	+	Vert. P	21.251	55.711	$-1.0\pm0.2$	$+0.21\pm0.06$	0.00	0.43	-1	+8	-14	2.5
14-20	+	Vert. E	21.251	55.714	$-1.8\pm0.05$	$+0.08\pm0.01$	0.16	0.84	+15	+17	+11	3.6
15-21	+	Tilt E	21.247	55.715	$-2.3\pm0.05$	$+0.24\pm0.01$	0.66	0.65	+33	-24	-16	49
16-22	+	Tilt D	21.248	55.715	$-2.2\pm0.01$	$+0.3\pm0.01$	0.87	0.00	+43	-22	+2	76