

1 **Near-Real-Time analysis of the ionospheric response to the 15 January 2022 Hunga**  
2 **Tonga-Hunga Ha'apai volcanic eruption**  
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9 **Key Points:**

- 10 ● We suggest novel methods that detect and determine spatio-temporal characteristics of  
11 ionospheric disturbances in Near-Real-Time (NRT).  
12 ● We analyze large- and small-amplitude ionosphere response to the Tonga eruption in near  
13 (<2000 km) and far-field (~10000 km) in NRT scenario  
14 ● The amplitude of the dTEC/dt response to the Tonga eruption is comparable to the 2011  
15 Tohoku earthquake and the 28 October 2003 solar flare  
16

## 17 **Abstract**

18 We present a near-real-time (NRT) scenario of analysis of ionospheric response to the 15 January  
19 2022 Hunga Tonga-Hunga Ha’apai eruption by using GNSS data in the near field (in the vicinity  
20 of the volcano), and in the far-field (Japan, North America and South America). We introduce a  
21 new method to determine instantaneous velocities using an interferometric approach and using the  
22 time derivative of the total electron content (TEC). Moreover, for the first time, we propose a novel  
23 method that automatically estimates the apparent propagation velocity of ionospheric disturbances  
24 from near-real-time travel-time diagrams. By using our new methods, we analyzed the dynamics  
25 of co-volcanic ionospheric disturbances generated by the Hunga-Tonga eruption, and we estimated  
26 the first propagation velocity in the near-field to be  $\sim 800\text{-}950$  m/s, subsequently decreasing to  
27  $\sim 600$  m/s. Based on these values, we conclude that in the near-field, we detect ionospheric  
28 signatures of acoustic waves. In the far field, the apparent velocity of ionospheric disturbances was  
29 estimated to be between 277 and 365 m/s, which corresponds to the propagation of the Lamb wave.  
30 It is important to note that our new methods can successfully perform at low spatial resolution  
31 networks and with 30-sec cadence data. Also, they enable NRT spatio-temporal analysis of  
32 ionospheric TEC response to smaller-amplitude events.

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34

## 35 **1 Introduction**

36 It is known that natural hazards, such as earthquakes, tsunamis, and volcanic eruptions  
37 generate acoustic and gravity waves that propagate upward in the atmosphere and ionosphere (e.g.,  
38 Calais & Minster, 1995; Miyaki et al., 2002; Heki, 2006; Astafyeva, 2019). Ionospheric  
39 disturbances generated by volcanic eruptions are called co-volcanic ionospheric disturbances (co-  
40 VID). It is known that the co-VID are usually quasi-periodically shaped variations that occur  $\sim 10$   
41 to 45 min after the eruption onset, last for 1-1.5 hours, occur in the near field of a volcano (up to  
42  $\sim 2000$  km), and propagate at velocities in the range of 0.5 km/s - 1.1 km/s (Heki, 2006;  
43 Dautermann et al., 2009; Nakashima et al., 2016; Liu et al., 2017). Since the first-time detection  
44 of co-VID by Heki (2006), nowadays, the disturbances can be systematically detected by ground-  
45 based GNSS receivers. Shults et al. (2016) introduced for the first time a term “Ionospheric  
46 Volcanology” that refers to the use of ionospheric measurements for the interests of volcanology.  
47 For instance, from the co-VID measurements, it is possible to determine the location of an eruptive  
48 volcano, the time of eruption onset (Shults et al., 2016), and estimate volcanic eruption power  
49 (Heki, 2006; Dautermann et al., 2009; Manta et al., 2021). Ionosphere-based methods would  
50 complement conventional ones, which use data from nearby seismometers and infrasound stations.  
51 The accuracy of those conventional methods decreases in absence of instrumentation within  $\sim 100$   
52 km from a volcano. To make a new step toward ionospheric volcano monitoring and warning  
53 systems we must develop real or near-real-time (NRT) methods.

54 In this work, for the first time, we analyze spatio-temporal characteristics of ionospheric  
55 disturbances generated by the 15 January 2022 massive eruption of Hunga Tonga-Hunga Ha’apai

56 (HTHH) volcano in the NRT scenario. We suggest novel methods that automatically estimate the  
 57 apparent propagation speed and direction of propagation of co-VID and other ionospheric  
 58 disturbances, including small-amplitude ones, in near-real-time. This work is an important step  
 59 toward automatised NRT detection of ionospheric disturbances.  
 60

## 61 **2 Data and Methods**

### 62 **2.1 Total Electron Content by Global Navigation Satellite Systems**

63

64 Global Navigation Satellite Systems (GNSS) are a helpful tool for ionospheric sounding.  
 65 Its main advantage is good spatial and temporal resolution. Nowadays, 30-second data from over  
 66 9000 worldwide receivers are available on a daily basis. Each GNSS station can receive signals  
 67 from 40+ GNSS satellites, providing numerous ionospheric observation points. High-rate data (1-  
 68 second or less) are sparser (around 1000 receivers), however, local networks developed vastly in  
 69 this direction over the last year. Phase measurements from dual-frequency GNSS receivers allow  
 70 estimation of the ionospheric total electron content (TEC), which is equal to the number of  
 71 electrons along a line-of-sight (LOS) between a satellite and a receiver:  
 72

72

$$73 \quad TEC_{ij} (phase, slant) = \frac{1}{A} \times \frac{f_i^2 f_j^2}{f_i^2 - f_j^2} \times (L_i \lambda_i - L_j \lambda_j) \quad (1)$$

74

75 where  $A = 40.308 \text{ m}^3/\text{s}^2$ ,  $L_i$  and  $L_j$  are phase measurements,  $\lambda_i$  and  $\lambda_j$  are wavelengths at the two  
 76 the given frequencies (for Global Positioning System (GPS)  $i=1, j=2$  and frequencies are 1575.42  
 77 and 1227.60 MHz, respectively). The TEC is measured in TEC units (TECu),  $1 \text{ TECu} = 10^{16}$   
 78 electrons/ $\text{m}^2$ .

79 We use the ionospheric thin shell approximation to calculate the spatial positions of  
 80 ionospheric disturbances. The intersection points between the LOS and this shell (at a fixed altitude  
 81  $H_{\text{ion}}$ ) are ionospheric pierced points (IPP). We use  $H_{\text{ion}} = 320 \text{ km}$  since it is close to the maximum  
 82 ionization height  $H_m F_2$  (based on the nearest ionosonde station NIUE at 169.9E; 19.1S).

83 To study the co-VID signatures driven by the HHTH volcano eruption, we analyze data of  
 84 24 ground-based GNSS-receivers in the near-field, i.e., under  $\sim 2000 \text{ km}$  away from the volcano.  
 85 To extract the co-VID signatures from the TEC data series, researchers usually apply 1-4 mHz  
 86 band-pass filters (Heki, 2006; Shults et al., 2016; Nakashima et al., 2016; Manta et al., 2021).  
 87 However, in a real-time scenario it is not possible because of the following reasons: a) the  
 88 impossibility to stack long series of data in NRT; b) such signal properties as arrival time,  
 89 amplitude, and spectral components can be affected by the filter parameters (Maletckii et al.,  
 90 2020). For NRT, we propose to use the TEC time derivative, which works as a high-pass filter and  
 91 removes the bias and trend caused by the satellite orbit motion. In addition, our  $d\text{TEC}/dt$  approach  
 92 will not modify the amplitude of the co-VID.  
 93

## 94 **2.2. The “D1-GNSS-RT” method**

95

96 By using the TEC time derivative approach, Maletckii and Astafyeva (2021a) introduced  
97 a method “D1-GNSS-RT” allowing to calculate spatio-temporal properties of traveling  
98 ionospheric disturbances (TID) in NRT (Figure 1). To detect TID, the “D1-GNSS-RT” method  
99 first analyses TEC data series to find the local maximum value (LMV). Then, it computes the  
100 cross-correlation function for each pair of time series around the LMV to calculate the difference  
101 in TID arrivals. Finally, based on these time shifts and by using an interferometric approach it  
102 estimates the horizontal velocities of TID propagation. The “D1-GNSS-RT” method was tested on  
103 several earthquakes but only showed good results with 1-sec data and on dense GNSS networks,  
104 such as Japan GEONET. The latter restrictions make it challenging to apply this method to the  
105 analysis of the co-VID generated by the HHTH volcanic eruption. The spatial coverage around the  
106 Tonga Islands is rather sparse, and only 16 out of 24 GNSS stations provide both 1-sec and 30-sec  
107 cadence data, while the others are limited to only 30-sec cadence data (Figure 2a). Besides, 30-sec  
108 dTEC/dt signals have smaller amplitudes and narrower spectral composition, which results in less  
109 pronounced signals as compared to 1-sec dTEC/dt data (Figure S1).

110 Here, for the first time, we introduce a new “D1-GNSS-RT” applicable to 30-sec data. The  
111 main developments are presented in Figure 1. They include: 1) increase of the LMV window to 7  
112 minutes, 2) increase of the cross-correlation window to 24 minutes; 3) decrease of the threshold of  
113 the coefficient of the cross-correlation function down to 0.7. However, unfortunately, these new  
114 parameters modify the definition of NRT from 15 minutes for 1-sec data to 30 minutes for 30-sec  
115 data.

116 When the “D1-GNSS-RT” is not applicable (e.g., sparse GNSS coverage), the horizontal  
117 TID velocity can be estimated by using travel-time diagrams, or hodocrones, that present the TEC  
118 variations with respect to the source location and time. Similar to the D1-GNSS-RT, for NRT-  
119 TTD we also use the dTEC/dt parameter. As the source, we take the volcano position. From TTD,  
120 the velocity can be estimated as the slope, however, up to now, there was no NRT-compatible  
121 automatique method to do that. Here, for the first time, we developed a novel technique to fit the  
122 slope line in NRT.

123

## 124 **2.3 The NRT TTD method and fitting technique**

125

126 The automatic NRT TTD fitting technique consists of two stages: 1) the first maximum  
127 “picker” and 2) the “fitter” based on these maxima. To select the maximum along with all dTEC/dt  
128 values, we pick the values exceeding a standard deviation of the series and a threshold of 0.15  
129 TECu. In the case of the multiple values in the 120-second windows, we chose the centered one in  
130 this window. We also remove outliers from the final list of maxima in the given series (values that  
131 can appear only with velocities exceeding 5 km/s).

132 We use the first maximum of each data series to fit the first velocity slope. They are sorted  
133 based on the source distance - from the closest to the farthest. By analyzing the velocity between

134 the current and previous maximum point we decide whether this maximum is “physically” suitable  
135 for the fitting process (velocity between two points should be in the range between 0.1 and 5 km/s  
136 and should not vary for more than 20% with respect to the velocity between two previous points;  
137 after picking the first 8 suitable maxima we add a new condition - the velocity should not change  
138 for more than 50% of the average velocity of all previous points). After the list of suitable points  
139 is finished, we fit the slope line by linear regression in these points.

140 In the case of the Quasi-NRT method, we added a second round for the picking process.  
141 After we obtain the first NRT velocity we compare all first maximum velocities with this value. If  
142 it lies in a 20% difference border interval, we pick this maximum. The new list of points is used  
143 for the Quasi-NRT fitting. Since the second round would require more time, we call this method  
144 “Quasi-NRT”. However, the Quasi-NRT method seems to be more accurate, therefore it can be  
145 used to determine NRT-method accuracy in a particular case.

146 We implement these techniques to the HTHH eruption (Sections 3.1 and 3.2), but also to  
147 the M6.6 16 July 2007 Chuetsu earthquake, which is the smallest earthquake ever recorded in the  
148 ionosphere (Cahyadi and Heki, 2015) and the 4 August 2020 Beirut explosion (Section 3.3).  
149

### 150 **3 Results and Discussion**

151 As shown recently, the explosive eruption of HTHH volcano produced quite a significant  
152 response in the ionosphere, and eruption-driven traveling ionospheric disturbances (TID) were  
153 observed as far as 20,000 km away from the volcano (Themens et al., 2022; Zhang et al., 2022).  
154 The amplitude of the near-field response reached as high value as 5-8 TECu (Astafyeva et al.,  
155 2022). In the case of the  $d\text{TEC}/dt$  parameter, we observe a peak-to-peak disturbance with the  
156 amplitude of  $\sim 8$  TECu, which is extraordinary, as this value exceeds by a factor of 2.5-3 all  
157 previously recorded co-VID (Figure S2). Previously, disturbances with large  $d\text{TEC}/dt$  were only  
158 observed during the 2011 Tohoku-Oki earthquake and during the 28<sup>th</sup> October 2003 solar flare  
159 (Figure S2). The exceptionally high amplitude of the HTHH-driven co-VID can be explained by  
160 the fact that the eruption was accompanied by explosions of extreme power force (e.g., Matoza et  
161 al., 2022; Astafyeva et al., 2022). As known, the amplitude of NH-driven ionospheric disturbances  
162 depend on the magnitude of the initial forcing: larger earthquakes and volcanic eruptions generate  
163 larger disturbances in the ionosphere (Astafyeva et al., 2013; Cahyadi and Heki, 2015; Shults et  
164 al., 2016; Manta et al., 2022).

165 Below we use our newly developed methods and we estimate spatio-temporal evolution of  
166 HTHH-driven co-VID in the NRT scenario: the amplitude of the velocity, the azimuths of  
167 propagation, and the ionospheric source location.  
168  
169

#### 170 **3.1 Near-Field ionospheric disturbance due to the Tonga Eruption**

171

### 172 **3.1.1 Spatio-temporal characteristics of the co-VID from D1-GNSS-RT. Instantaneous** 173 **velocities' field and source location.**

174  
175 Figure 2 (b-f) summarizes the results of the application of the D1-GNSS-RT method to  
176 the analysis of ionospheric TEC disturbances generated by the 15 January 2022 eruption. The co-  
177 VID velocity field maps for the first arrivals following the Hunga-Tonga eruption are shown in  
178 Fig. 2b–d, and the localization results are presented in Fig. 2e–f. Figure 2b shows the first velocity  
179 vectors at 04:23:30UT, i.e., 525s after the eruption onset time, both on the north-east and south-  
180 west out from the volcano. From the time of the first co-VID detection, in the NRT scenario, we  
181 need 22 minutes more to compute the first velocity field, which is an increase of the time delay for  
182 the NRT method as compared to 1-sec data. The two main reasons are a long 30-sec cross-  
183 correlation window (24 minutes vs. 5 minutes with 1-sec data) and sparse spatial resolution. The  
184 latter signifies fewer IPP that can be selected for correlation triangles after the first co-VID  
185 detection. Therefore, more time is necessary to “form” an interferometric triangle. The first vectors  
186 propagate in directions outward from the source. The first horizontal velocities of the co-VID are  
187 about ~830-900 m/s, i.e., they correspond to acoustic and shock-acoustic waves, and are in line  
188 with retrospective studies (e.g., Themens et al., 2022). The first velocity vectors are used to  
189 compute the first source location at the point with coordinates (17.90S; 176.26E) (Fig. 2e). The  
190 subsequent co-VID evolution during the next 2 minutes maintains the tendency for both the  
191 outward direction of propagation and velocities' values. Further, the velocities decrease to ~500-  
192 600 m/s, while the source locations concentrate northwest of the volcano (Fig. 2f).

### 193 194 **3.1.2 Spatio-temporal characteristics of the co-VID from NRT TTD using 30-sec data.**

195  
196 The 30-sec NRT-TTD for all satellites and receivers (e.g, all LOS) is shown in Figure 3a.  
197 From these data, our newly developed fitting method estimates the velocity to be 621.1 m/s. This  
198 value is in line with previous retrospective observations for the ionospheric response to the Hunga-  
199 Tonga eruption (Themens et al., 2022), as well as with our “D1-GNSS-RT” results. The error of  
200 the velocity estimations is less than 10% for both NRT and Quasi-NRT method (Figure 3b,c). The  
201 difference between NRT and Quasi-NRT estimations is 11,1%. We can observe the existence of  
202 the co-VID signatures before the fitted slope line on Figure 3a, but the amplitudes of the  
203 disturbances were not sufficient for the “picker” part of the automatic NRT TTD fitting technique.

### 204 205 **3.1.3 Spatio-temporal characteristics of the co-VID from NRT TTD using 1-sec data.**

206  
207 As mentioned above, only 16 GNSS receivers in the near-field of the HTHH volcano  
208 provided 1-sec data, which is too few to use the 1-sec “D1-GNSS-RT” method. Fortunately, these  
209 limits do not apply to NRT TTD. Figure 4 shows the dTEC/dt-based TTD plotted for the near-  
210 field co-VID. We note that the high-rate response to the HTHH volcanic eruption is more complex  
211 than the 30-sec one. Figure 5b demonstrates the occurrence of four dTEC/dt disturbances that are,

212 most likely, related to four independent eruptive events that occurred between 04:00 and 05:30  
213 UT. The separate events can be distinguished on TTD based on the characteristics of the  
214 ionospheric responses, such as signal shape, the apparent velocity of propagation, and the  
215 amplitude.

216 The NRT TTD shows one quasi-periodic and three N-shaped signatures (dotted ovals in  
217 Figure 4b). The first quasi-periodic response (in the green circle) has the lowest velocity with  
218 respect to the other disturbances ( $\sim 0.5$  km/s). For the second response, the slope gives the apparent  
219 velocity of  $\sim 1.33$  km/s. It appears to consist of three N-shaped signals which have identical  
220 velocity slopes. Further, we distinguish the third event based on a new increase in the  $d\text{TEC}/dt$   
221 from  $\sim 05:15$  UT. For this component, the velocity slope is  $\sim 2$  km/s. Finally, the fourth event has  
222 an apparent velocity of  $\sim 1.33$  km/s, which distinguishes it from the third event, although it is close  
223 in time.

224 Figure 4a shows an example of  $d\text{TEC}/dt$  signatures for receiver “SAMO” – satellite  
225 GLONASS R21 (in blue-white-red colormap). We also implement a centered moving average  
226 filter (5-sec window) to this series (black curve), which allow to remove noise in data and to  
227 concentrate on useful signals. These results prove an assumption of two types of the signatures:  
228 first, quasi-periodic and then, N-shaped ones. Evenmore, we observed the first co-VID driven  
229 signatures a couple of minutes before the USGS-determined eruption onset time (04:15 UT).  
230 Generally, it takes  $\sim 7$ -10 minutes for disturbances to reach the ionospheric altitudes, therefore the  
231 eruption onset occurred between 04:00 and 04:10 UT.

232 From our NRT-TTD, it is possible to estimate the onset times for all observed co-VID  
233 (Figure 4c). To do so, we first compute the intersection of the velocity slope line with the 0-km  
234 distance from the source. Second, we estimate the time in the intersection point from the TTD.  
235 This time corresponds to the onset time in the ionosphere, which is the time when the eruption-  
236 driven acoustic wave reaches the ionosphere (i.e., the altitude of detection,  $H_{\text{ion}} = 320$  km). Third,  
237 we compute the vertical propagation time for the acoustic wave from the volcano to the ionosphere  
238 by using the sound speed profile derived from the NRLMSISE-2 model (Emmert et al., 2020).  
239 With a weighted average velocity of the sound speed of 470 m/s (Figure S3b), the acoustic wave  
240 will take  $\sim 11.34$  minutes (11 minutes 20 seconds) to reach 320 km of altitude. Finally, we extract  
241 this propagation time from the ionospheric onset times in order to obtain the ground onset times  
242 for all four events (Table S1). From our method it follows that the HTHH volcano began to erupt  
243 at 04:08:26 UT, which is in agreement with satellite observations that suggest the eruption onset  
244 between 04:00 and 04:10 UT (Gusman and Rodger, 2022). Our onset time is also very close to  
245 that estimated by Astafyeva et al. (2022) from raw unfiltered TEC data by retrospective analysis.  
246 However, it is several minutes earlier than seismically-determined onset time (USGS; Poli &  
247 Shapiro, 2022), and  $\sim 20$  minutes earlier than the onset estimated by using a pressure station at  
248 Tonga (Wright et al, 2022). Our work demonstrates that our ionosphere-based NRT approach can  
249 be successfully used along with conventional methods.

250 The occurrence of multiple eruptive events, that is clearly seen in  $d\text{TEC}/dt$  data, is in line  
251 with previous reports. For instance, Wright et al. (2022) identified four independent events that

252 occurred between 04:00 and 05:30 UT: 04:26 UT, 04:36 UT, 05:10 UT, 05:51 UT. Astafyeva et  
253 al, 2022 suggested the occurrence of five eruptive events between 04:00 and 05:30 UT, however  
254 their onset times differ from our estimations, which can be due to difference in the approximations  
255 used.

256

### 257 **3.2 Far-Field ionospheric disturbance due to the Tonga Eruption.**

258

259 Previously, we applied our approach to the analysis of ionospheric response in the near-  
260 field of the HTHH volcano and earthquakes (Maletckii & Astafyeva, 2021a). The near-field  
261 ionospheric disturbances are usually characterized by relatively high velocities (e.g., 800-1200  
262 m/s) and high frequencies (e.g., 4-10 mHz). In this section, we demonstrate how this approach and  
263 our methods can perform in the Far-Field (i.e., several thousands of km away from the source) and  
264 process traveling disturbances with lower velocities and frequencies. Perturbations with such  
265 characteristics include tsunami-induced gravity waves. Therefore, the NRT-method can be used  
266 for early warning systems.

267 To perform in the Far-Field, the following adjustments in our NRT fitting technique were  
268 made: 1) the first maximum threshold is increased from 0.15 TECu/sec to 0.28 TECu/sec (for 30-  
269 second data); 2) the velocity between two points should be in the range between 0.18 and 5 km/s.  
270 The main reason to do so is to prevent “false detections”. Since the ionosphere is an extremely  
271 disturbed medium, different disturbances are always present at any place at any time moment. By  
272 increasing the thresholds, we exclude disturbance not related to the eruption.

273 We apply the adjusted method to the detection of HTHH-driven disturbances in Japan,  
274 North America and Chile, i.e. between 8,000 and 11,000 km away from the HTHH volcano (Figure  
275 5).

276 We use the GEONET GNSS network to detect and characterize traveling ionospheric  
277 disturbances on the Japanese coast (Animation S1). The 30-sec NRT-TTD for satellite GPS G07  
278 and all available receivers is shown in Figure 5 (a-c). The first vivid signatures appeared at ~10:00  
279 UT, ~6 hours after the first eruption. From these data, our newly developed fitting method  
280 estimates the velocity to be 336.5 m/s, which is close to the Lamb wave speed, and is in line with  
281 previous retrospective observations for the ionospheric response to the Hunga-Tonga eruption in  
282 Japan (Themens et al., 2022; Zhang et. al., 2022).

283 The Lamb-wave driven ionospheric disturbances arrived on the West coast of North  
284 America at ~12:00 UT, ~8 hours after the eruption onset (Animation S2). Figure 5 (d-f) shows 30-  
285 sec NRT-TTD for satellite GPS G10 and all available receivers. Based on the TTD, our newly  
286 developed fitting method estimates the velocity to be 365.9 m/s, which is slightly higher than the  
287 Lamb wave, and is in agreement with retrospective analysis of the ionospheric response in North  
288 America (Zhang et. al., 2022).

289 To study the response in the South-West Coast of South America (Animation S3), we used  
290 1-sec data from the Centro Sismológico Nacional Universidad de Chile GNSS archives. The 1-sec  
291 NRT-TTD for satellite GPS G18 is shown in Figure 5 (g-i). We also increased the threshold for

the picker up to 0.75 TECu/sec, since 1-sec data series are noisier and have larger peak-to-peak amplitudes than 30-sec data (Figure S1). The first disturbances arrive at ~12:00 UT, ~8 hours after the first eruption. From these data, our newly developed fitting method estimates the velocity to be 277.6 m/s. We attribute this disturbance to ionospheric response to the Lamb wave propagation.

### 3.3. Ionospheric disturbances driven by other events: small earthquake and explosion

To analyze the applicability and accuracy of the NRT-TTD method and the fitting technique, we analyze two events: 1) the M6.6 16 July 2007 Chuetsu earthquake in Japan, which is the smallest earthquake ever recorded in the ionosphere; 2) the 4 August 2020 Beirut explosion. Both events caused very weak TEC response as compared to the Tonga event (Figure S2).

The response to the Chūetsu earthquake as captured by satellite GPS G26 is presented in Figure 6 (a)-(c). Co-seismic ionospheric disturbances are seen ~10 minutes after the earthquake onset. We estimated their propagation speed to be 949.4 m/s, which is in agreement with the retrospective results (1 km/s by Cahyadi and Heki, 2015). The response to the Beirut explosion was captured by satellite GPS G22 (Figure 6 (d)-(f)). Clear N-shaped disturbances emerged ~12 minutes after the explosion onset and their velocity is estimated to be 883 m/s. Our estimation is in agreement with the retrospective estimations (0.8 km/s by Kundu et. al., 2021). We note that the spatial resolution of the GNSS network was very poor, which made it challenging to automatically process it, but our method succeeded.

## 4 Conclusions

In this work, we performed for the first time a near-real-time analysis of the ionospheric response to the massive 15 January 2022 Hunga Tonga-Hunga Ha'apai explosive eruption. Our main developments and findings are summarized below:

1. For the first time, we introduce a new method to determine spatio-temporal characteristics in the NRT. This method estimates the instantaneous velocities and the ionospheric source location using not only high-rate data but also the “conventional” 30-sec data. In addition, our new method can perform in sparse spatial coverage conditions. We note, however, that 30-sec data increase the NRT time delay between the event onset and the first results to ~30 minutes. By using this method, in a near-real-time scenario applied for the HTHH eruption case, we estimate the first instantaneous velocities to be ~800-900 m/s, which is in line with retrospective studies (e.g., Themens et al., 2022; Zhang et al., 2022), and correspond to acoustic and shock-acoustic waves. The location of the ionospheric source determined by our method is in the northwest of the volcano.
2. For the first time, we present a new method that can estimate the co-VID velocity from a real-time travel-time diagram. For the HTHH volcanic eruption, we observe the apparent co-VID propagation speed to be 621.1 m/s. This value is in line with our “D1-GNSS-RT” results. To further demonstrate the wide applicability of our method, we tested them on lower-amplitude

331 TEC responses in Japan, North America and Chile on the day of the HTHH eruption, and to  
 332 the Beirut explosion of 4 August 2020, which was registered by a very sparse GNSS network,  
 333 and the M6.6 Chuetsu earthquake of July 2007, which is the smallest earthquake ever  
 334 registered in the ionosphere. In all cases, our method managed to capture the response and to  
 335 correctly estimate the velocities.

- 336 3. Our dTEC/dt near-field NRT-TTD suggest the occurrence of four distinct eruptions between  
 337 04:00 and 05:30 UT. From the velocity slopes in NRT-TTD, we estimate the onset time for  
 338 the four events at 04:08:43 UT, 04:31:00 UT, 05:02:30 UT, and 05:05:21 UT. The multi-  
 339 eruption scenario is an agreement with the analysis of surface pressure data (Wright et al.,  
 340 2022) and that of the unfiltered ionospheric TEC data (Astafyeva et al., 2022).
- 341 4. We emphasize that the amplitude of the dTEC/dt ionospheric response to the HTHH volcanic  
 342 eruption is unprecedentedly strong: the peak-to-peak dTEC/dt disturbance amplitude  
 343 exceeded by a factor of 2.5-3 all previously recorded co-VID. Such extreme values emphasize  
 344 the unprecedented power of the HTHH volcano explosion, and are comparable to the  
 345 ionospheric response to the 2011 Great Tohoku-Oki earthquake and the 28 October 2003 solar  
 346 flare.

347  
 348 Our results once again demonstrate the advantages of the use of the dTEC/dt parameter as  
 349 the effective NRT tool to rapidly determine dynamic characteristics of ionospheric disturbances.  
 350 We also demonstrate that an ionosphere-based method can be a reliable alternative for detection  
 351 of natural hazard events. This is especially important and useful for the analysis of submarine  
 352 events, such as the HTHH volcanic eruption, where ground-based instrumentation is very limited.

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 356 “tec-suite” codes developed by I. Zhivetiev (<https://tec-suite.readthedocs.io/en/latest/>).

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 358 fruitful discussions within an ad-hoc Geoazur-IPGP-NGI working group on the 2022 Hunga Tonga  
 359 volcano eruption.

### 360 **Open Research**

362 The Near-field and the Beirut explosion GNSS data are available from the CDDIS data  
 363 archives (<https://cddis.nasa.gov/archive/gnss/data/daily/>). The Japan and the Chūetsu earthquake  
 364 GNSS data are available from the GeoSpatial Authority of Japan (GSI, [terras.go.jp](https://terras.go.jp)).  
 365 [http://datahouse1.gsi.go.jp/terras/terras\\_english.html](http://datahouse1.gsi.go.jp/terras/terras_english.html). The North America West Coast data are  
 366 available from the UNAVCO data archives (<https://data.unavco.org/archive/gnss/rinex/>). The  
 367 South America West Coast data are available from the Centro Sismológico Nacional Universidad  
 368 de Chile data archives (<http://gps.csn.uchile.cl/data/>) and Instituto Geografico Nacional Argentino  
 369 (<https://www.ign.gov.ar/NuestrasActividades/Geodesia/Ramsac/DescargaRinex>; Piñón et al.,  
 370 2018)

371 Ionosonde station NIUE data are available from the DIDBase Web Portal  
 372 ([https://lgdc.uml.edu/common/DIDBMonthListForYearAndStation?ursiCode=ND61R&year=20](https://lgdc.uml.edu/common/DIDBMonthListForYearAndStation?ursiCode=ND61R&year=2022)  
 373 [22](https://lgdc.uml.edu/common/DIDBMonthListForYearAndStation?ursiCode=ND61R&year=2022)).

374 Figures were plotted by using Python (ver. 3.7, libraries “matplotlib.pyplot”:  
 375 [https://matplotlib.org/3.5.0/api/ as\\_gen/matplotlib.pyplot.html](https://matplotlib.org/3.5.0/api/as_gen/matplotlib.pyplot.html) and “cartopy”:  
 376 <https://scitools.org.uk/cartopy/docs/latest/>)  
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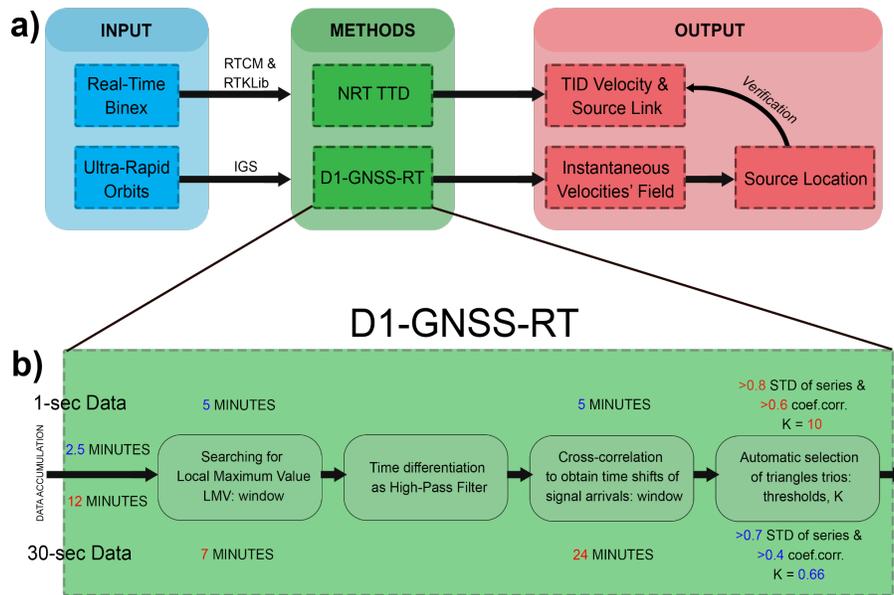
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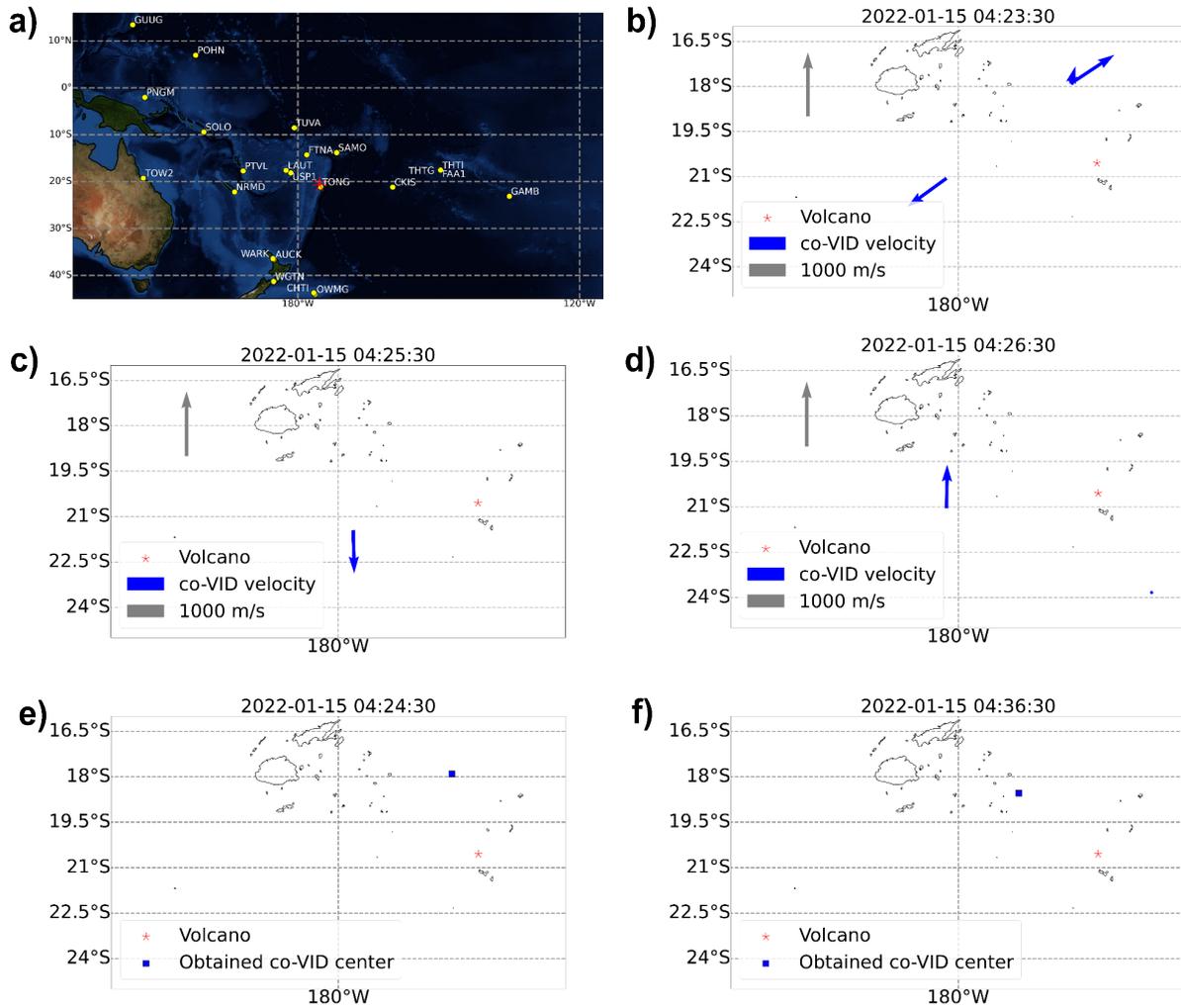
461 **Figures Captions**

462 **Figure 1.** Scheme of methods developed and implemented in this work. “D1-GNSS-RT” and NRT  
 463 TTD methods require Real-Time TEC (can be transferred by RTKlib software (Takasu, 2013) and  
 464 RTCM protocol (RTCM, 2020)) and orbits (can be obtained by Ultra-Rapid Orbits provided by  
 465 IGS (Noll, 2010)) data. “D1-GNSS-RT” method calculates the instantaneous velocities’ field and  
 466 the direction of propagation for the detected disturbances. Based on these results, we compute the  
 467 source location. NRT TTD estimates TID velocity and verifies the link with the source location.  
 468 Panel (b) shows the difference in parameters between the 1-sec “D1-GNSS-RT” method that was  
 469 developed previously Maletckii and Astafyeva (2021a) and 30-sec “D1-GNSS-RT” that was  
 470 developed and implemented here for the first time.



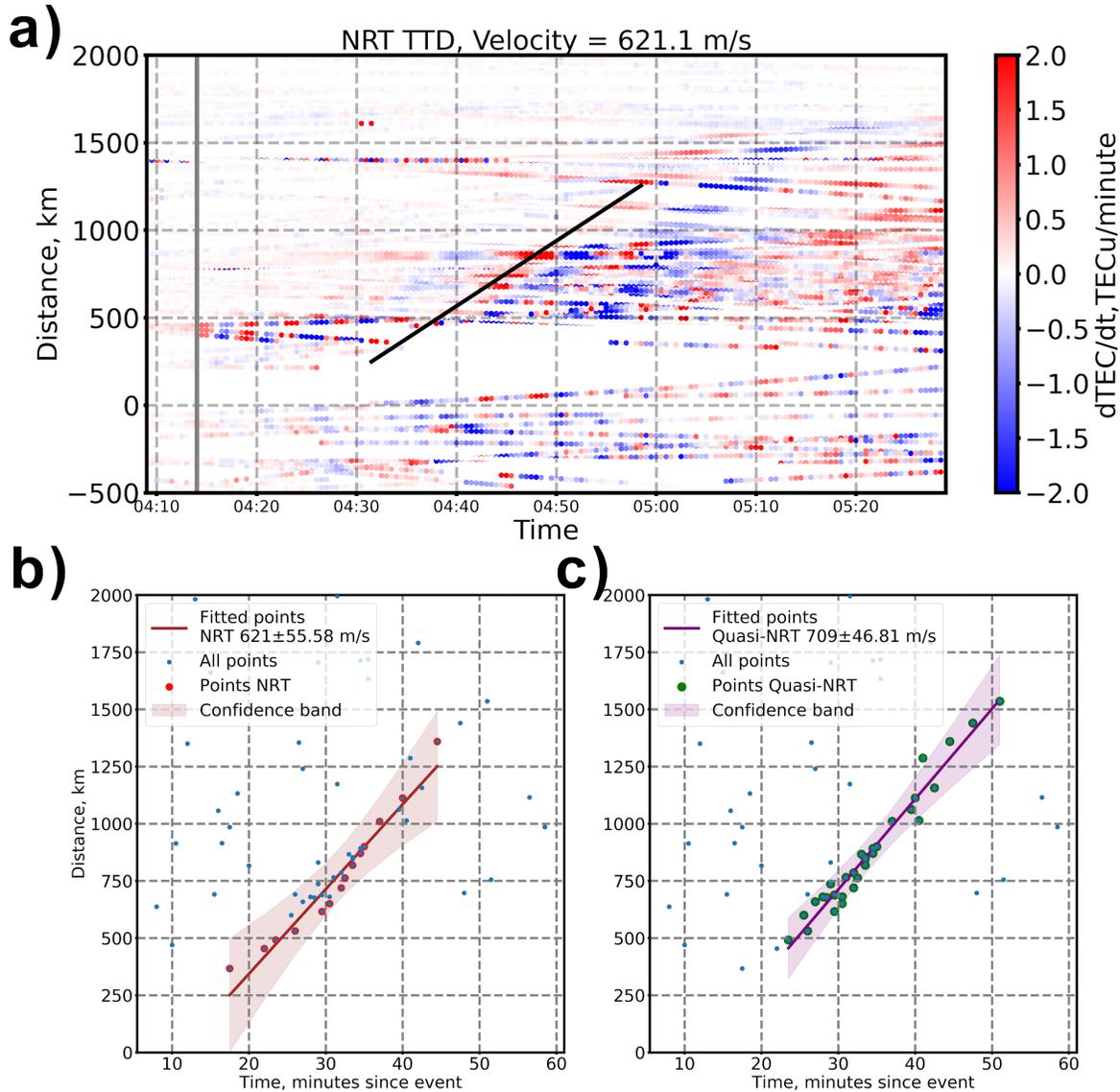
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492 **Figure 2.** Geometry of near-field GNSS observations **(a)** and the results of the D1-GNSS-RT  
 493 method **(b-f)**. **(a)** The Hunga Tonga-Hunga Ha’apai volcano (red star, 175.382W; 20.53S) and  
 494 GNSS receivers (yellow dots) network used in this work. The receivers that provide both 30 sec  
 495 and 1 sec data are: “CKIS”, “FAA1”, “FTNA”, “LAUT”, “PTVL”, “SAMO”, “SOLO”, “THTG”,  
 496 “TONG”, “TOW2”, “TUVA”, “USP1”. The others provide only 30 second data; **(b-d)** The first  
 497 instantaneous velocities’ field obtained by the “D1-GNSS-RT”. Gray arrow denotes the velocity  
 498 vector of 1000 m/s. The blue arrows correspond to the instantaneous velocities’ field of co-VID;  
 499 **(e-f)** the source locations (blue crosses) obtained from the instantaneous velocity vectors.



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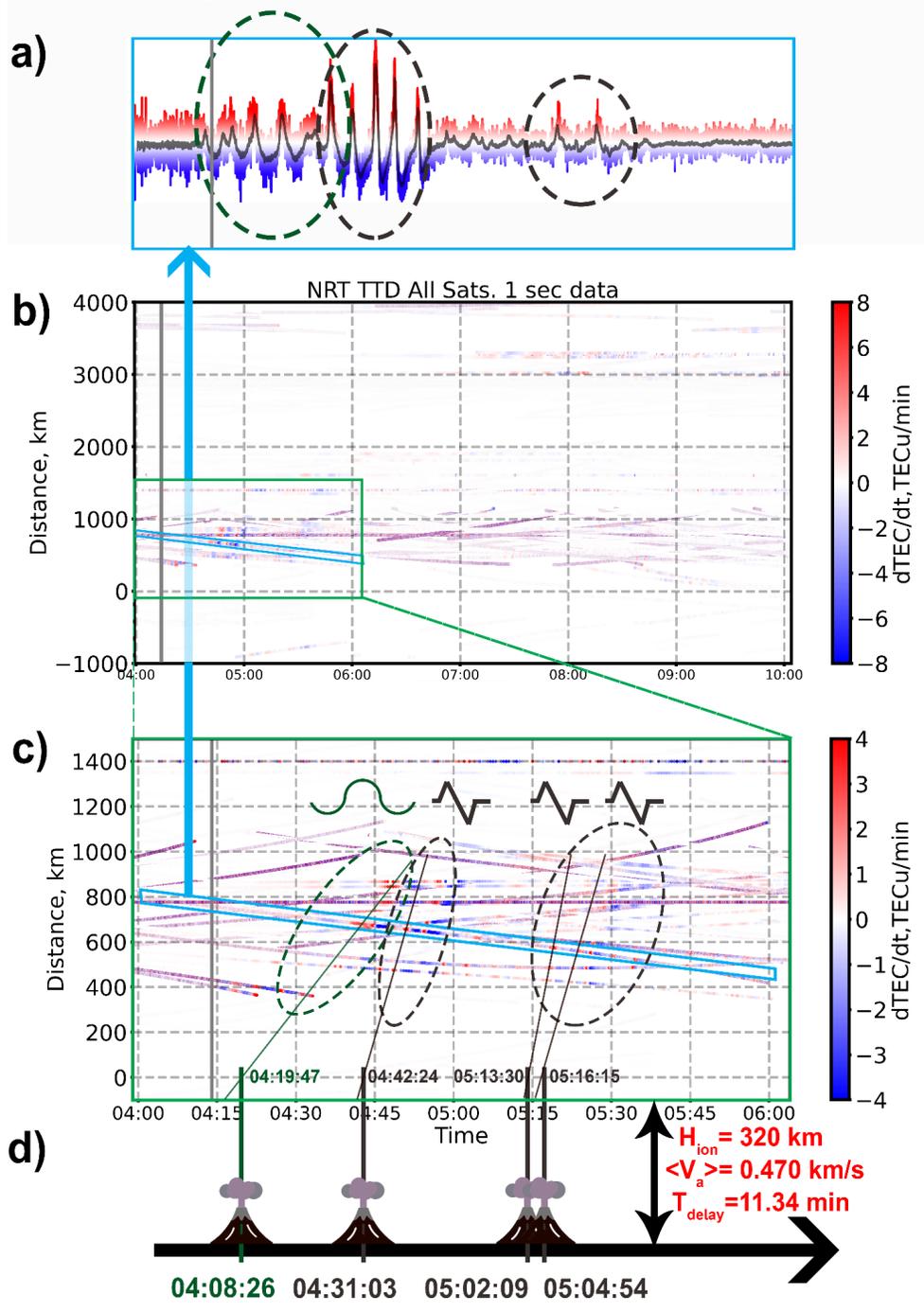
510 **Figure 3.** Application of NRT-TTD method to the near-field observations. **(a)** NRT TTD using  
 511 30-sec data and the estimated co-VID velocity (black line). Gray vertical line shows the USGS  
 512 onset time at 04:15UT. The source is located in the Hunga Tonga-Hunga Ha’apai volcano. The  
 513 black line was fitted by the newly developed automatic NRT-algorithm. **(b-c)** The two fitting  
 514 algorithms that estimate the velocity from the TTD slope: **(b)** the NRT - the brown line, **(c)** the  
 515 Quasi-NRT - purple. The blue dots correspond to the first maximums picked in each data series.  
 516 The red and the green dots are used for the linear regression by the NRT and the Quasi-NRT  
 517 algorithms, respectively.  
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**Figure 4.** **(a)** dTEC/dt variations from a receiver “SAMO” - a satellite R21 LOS, blue-white-red curve - 1-sec data, black curve - 5 second centered smoothed data; **(b, c)** NRT TTD plotted using 1-sec data **(b)** and **(c)** zoom on the near-field dTEC/dt response from 04:00 to 05:30 UT. Gray

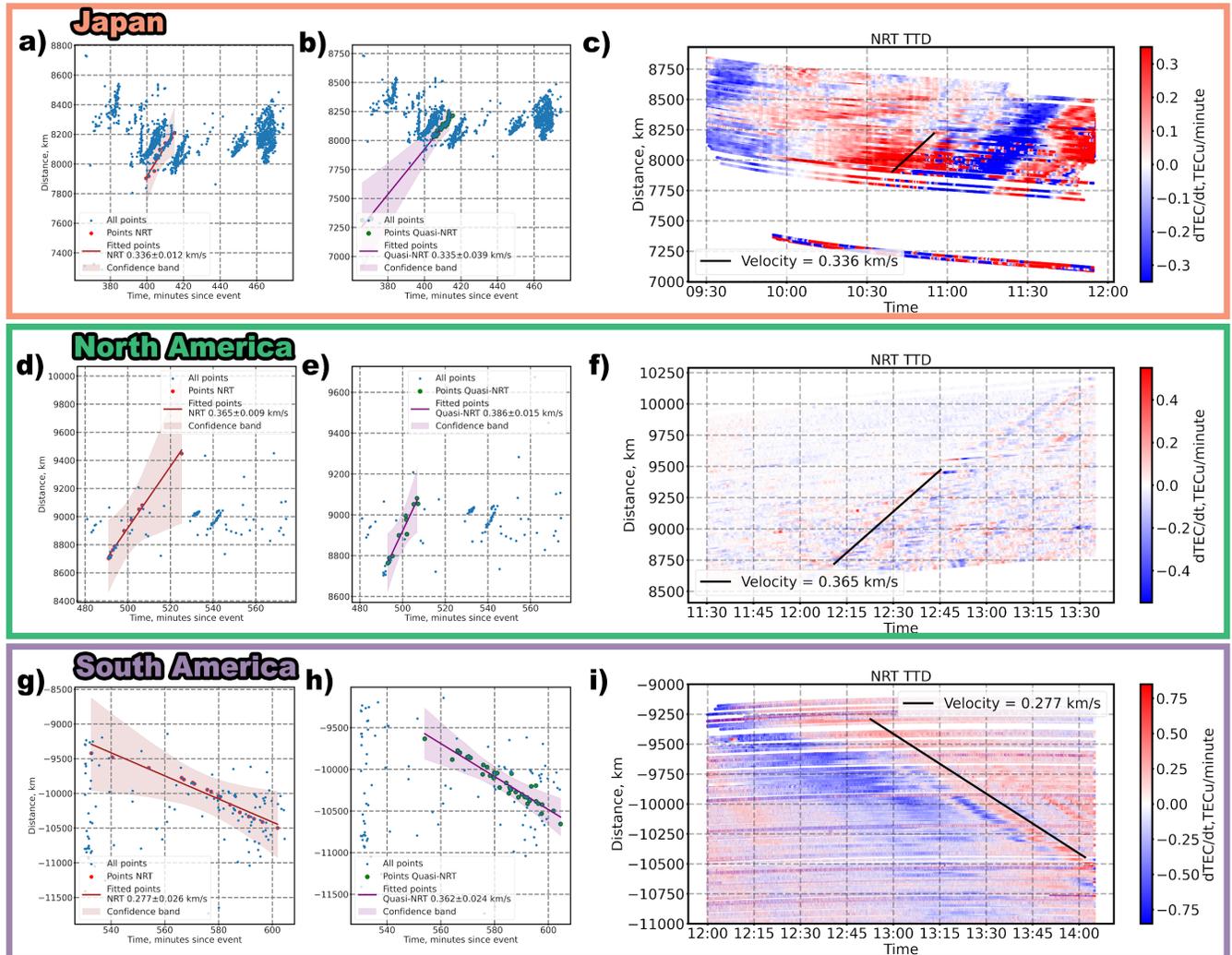
526 vertical line denotes the USGS onset time, the circles highlight four different disturbances detected  
 527 in the near-field of the HTHH volcano (green - quasi-periodic signature, dark brown - N-shape  
 528 ones). The slopes denote the apparent velocities of these four disturbances; **(d)** schematic  
 529 representation of multi-eruption scenario and the onset time for each event



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533 **Figure 5.** Application of NRT-TTD method to the far-field observations of ionospheric response.  
 534 **(a,b,c)** the Japanese GNSS network and satellite GPS G07; **(d,e,f)** North American GNSS  
 535 receivers and satellite GPS G10; **(g,h,i)** South American GNSS receivers and satellite GPS G10.  
 536 The source is located in the Hunga Tonga-Hunga Ha’apai volcano. The black lines **(c,f,i)** were  
 537 fitted by the automatic NRT-algorithm. The blue dots on panels **(a,b,d,e,g,h)** correspond to the  
 538 first maxima of each series. The red and the green dots **(a,b,d,e,g,h)** are used for the linear  
 539 regression by the NRT and the Quasi-NRT algorithms, respectively.



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542 **Figure 6.** Application of NRT-TTD fitting technique to the M6.6 Chuetsu earthquake of 16 July  
 543 2007 **(a,b,c)** and the Beirut explosion of 4 August 2020 **(d,e,f)**. The blue dots on panels **(b,c,e,f)**  
 544 correspond to the first maxima of each series. The red **(b,e)** and the green **(c,f)** dots are used for  
 545 the linear regression by the NRT and the Quasi-NRT algorithms, respectively. NRT velocity's

546 slope - the brown line on (b,e), the black line on (a,d); the Quasi-NRT (c,f) - purple, the event  
 547 onset time is indicated by a vertical gray line (a,b).

