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Near-Real-Time analysis of the ionospheric response to the 15 January 2022 Hunga Tonga-Hunga Ha'apai volcanic eruption

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- 8

9 Key Points:

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

17 Abstract

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

33 34

35 **1 Introduction**

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

In this work, for the first time, we analyze spatio-temporal characteristics of ionospheric disturbances generated by the 15 January 2022 massive eruption of Hunga Tonga-Hunga Ha'apai (HTHH) volcano in the NRT scenario. We suggest novel methods that automatically estimate the apparent propagation speed and direction of propagation of co-VID and other ionospheric disturbances, including small-amplitude ones, in near-real-time. This work is an important step toward automatised NRT detection of ionospheric disturbances.

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61 **2 Data and Methods**

62 2.1 Total Electron Content by Global Navigation Satellite Systems

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

72

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

74

where A = 40.308 m³/s², L_i and L_j are phase measurements, λ_i and λ_j are wavelengths at the two the given frequencies (for Global Positioning System (GPS) i=1, j=2 and frequencies are 1575.42 and 1227.60 MHz, respectively). The TEC is measured in TEC units (TECu), 1 TECu = 10¹⁶ electrons/m².

We use the ionospheric thin shell approximation to calculate the spatial positions of ionospheric disturbances. The intersection points between the LOS and this shell (at a fixed altitude H_{ion}) are ionospheric pierced points (IPP). We use $H_{ion} = 320$ km since it is close to the maximum ionization height HmF2 (based on the nearest ionosonde station NIUE at 169.9E; 19.1S).

To study the co-VID signatures driven by the HHTH volcano eruption, we analyze data of 83 24 ground-based GNSS-receivers in the near-field, i.e., under ~2000 km away from the volcano. 84 To extract the co-VID signatures from the TEC data series, researchers usually apply 1-4 mHz 85 band-pass filters (Heki, 2006; Shults et al., 2016; Nakashima et al., 2016; Manta et al., 2021). 86 However, in a real-time scenario it is not possible because of the following reasons: a) the 87 impossibility to stack long series of data in NRT; b) such signal properties as arrival time, 88 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 removes the bias and trend caused by the satellite orbit motion. In addition, our dTEC/dt approach 91 will not modify the amplitude of the co-VID. 92

94 2.2. The "D1-GNSS-RT" method

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

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

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

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

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

We use the first maximum of each data series to fit the first velocity slope. They are sorted based on the source distance - from the closest to the farthest. By analyzing the velocity between the current and previous maximum point we decide whether this maximum is "physically" suitable for the fitting process (velocity between two points should be in the range between 0.1 and 5 km/s and should not vary for more than 20% with respect to the velocity between two previous points; after picking the first 8 suitable maxima we add a new condition - the velocity should not change for more than 50% of the average velocity of all previous points). After the list of suitable points is finished, we fit the slope line by linear regression in these points.

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

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

149

150 **3 Results and Discussion**

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

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

168 169

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

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

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

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

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The 30-sec NRT-TTD for all satellites and receivers (e.g. all LOS) is shown in Figure 3a. 196 From these data, our newly developed fitting method estimates the velocity to be 621.1 m/s. This 197 value is in line with previous retrospective observations for the ionospheric response to the Hunga-198 Tonga eruption (Themens et al., 2022), as well as with our "D1-GNSS-RT" results. The error of 199 the velocity estimations is less than 10% for both NRT and Ouasi-NRT method (Figure 3b.c). The 200 difference between NRT and Quasi-NRT estimations is 11,1%. We can observe the existence of 201 202 the co-VID signatures before the fitted slope line on Figure 3a, but the amplitudes of the disturbances were not sufficient for the "picker" part of the automatic NRT TTD fitting technique. 203 204

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

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As mentioned above, only 16 GNSS receivers in the near-field of the HTHH volcano provided 1-sec data, which is too few to use the 1-sec "D1-GNSS-RT" method. Fortunately, these limits do not apply to NRT TTD. Figure 4 shows the dTEC/dt-based TTD plotted for the nearfield co-VID. We note that the high-rate response to the HTHH volcanic eruption is more complex than the 30-sec one. Figure 5b demonstrates the occurrence of four dTEC/dt disturbances that are, most likely, related to four independent eruptive events that occurred between 04:00 and 05:30 UT. The separate events can be distinguished on TTD based on the characteristics of the ionospheric responses, such as signal shape, the apparent velocity of propagation, and the amplitude.

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

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

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

The occurrence of multiple eruptive events, that is clearly seen in dTEC/dt data, is in line with previous reports. For instance, Wright et al. (2022) identified four independent events that occurred between 04:00 and 05:30 UT: 04:26 UT, 04:36 UT, 05:10 UT, 05:51 UT. Astafyeva et
al, 2022 suggested the occurrence of five eruptive events between 04:00 and 05:30 UT, however
their onset times differ from our estimations, which can be due to difference in the approximations
used.

256

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

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

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

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

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

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

To study the response in the South-West Coast of South America (Animation S3), we used 1-sec data from the Centro Sismológico Nacional Universidad de Chile GNSS archives. The 1-sec 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.

296

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 303 Figure 6 (a)-(c). Co-seismic ionospheric disturbances are seen ~ 10 minutes after the earthquake 304 onset. We estimated their propagation speed to be 949.4 m/s, which is in agreement with the 305 retrospective results (1 km/s by Cahyadi and Heki, 2015). The response to the Beirut explosion 306 was captured by satellite GPS G22 (Figure 6 (d)-(f)). Clear N-shaped disturbances emerged ~12 307 minutes after the explosion onset and their velocity is estimated to be 883 m/s. Our estimation is 308 309 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 310 automatically process it, but our method succeeded. 311

312

313 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 317 the NRT. This method estimates the instantaneous velocities and the ionospheric source 318 location using not only high-rate data but also the "conventional" 30-sec data. In addition, 319 our new method can perform in sparse spatial coverage conditions. We note, however, that 320 30-sec data increase the NRT time delay between the event onset and the first results to \sim 30 321 minutes. By using this method, in a near-real-time scenario applied for the HTHH eruption 322 case, we estimate the first instantaneous velocities to be ~800-900 m/s, which is in line with 323 retrospective studies (e.g., Themens et al., 2022; Zhang et al., 2022), and correspond to 324 acoustic and shock-acoustic waves. The location of the ionospheric source determined by our 325 method is in the northwest of the volcano. 326

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

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

- 3. Our dTEC/dt near-field NRT-TTD suggest the occurrence of four distinct eruptions between
 04:00 and 05:30 UT. From the velocity slopes in NRT-TTD, we estimate the onset time for
 the four events at 04:08:43 UT, 04:31:00 UT, 05:02:30 UT, and 05:05:21 UT. The multieruption scenario is an agreement with the analysis of surface pressure data (Wright et al.,
 2022) and that of the unfiltered ionospheric TEC data (Astafyeva et al., 2022).
- 4. We emphasize that the amplitude of the dTEC/dt ionospheric response to the HTHH volcanic
 eruption is unprecedentedly strong: the peak-to-peak dTEC/dt disturbance amplitude
 exceeded by a factor of 2.5-3 all previously recorded co-VID. Such extreme values emphasize
 the unprecedented power of the HTHH volcano explosion, and are comparable to the
 ionospheric response to the 2011 Great Tohoku-Oki earthquake and the 28 October 2003 solar
 flare.
- 347

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

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360

361 **Open Research**

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

Ionosonde station NIUE data are available from the DIDBase Web Portal

371 (https://lgdc.uml.edu/common/DIDBMonthListForYearAndStation?ursiCode=ND61R&year=20 372 373 22). 374 Figures were plotted by using Python (ver. 3.7, libraries "matplotlib.pyplot": https://matplotlib.org/3.5.0/api/ as gen/matplotlib.pyplot.html "cartopy": 375 and https://scitools.org.uk/cartopy/docs/latest/) 376 377 References 378 1. Astafyeva, E. (2019) Ionospheric detection of natural hazards. Reviews of Geophysics 379 57(4), 1265-1288. doi.org: 10.1029/2019RG000668 380 2. Astafyeva, E., Maletckii, B., Mikesell, T. D., Munaibari, E., Ravanelli, M., Coisson, P., et 381 al. (2022). The 15 January 2022 Hunga Tonga eruption history as inferred from ionospheric 382 Geophysical Research Letters, e2022GL098827. doi.org: observations. 49. 383 10.1029/2022GL098827 384 3. Astafyeva, E., Shalimov, S., Olshanskaya, E., & Lognonné, P. (2013). Ionospheric 385 response to earthquakes of different magnitudes: Larger quakes perturb the ionosphere 386 longer. Geophysical Research stronger and Letters. 40(9). 1675-1681. 387 388 https://doi.org/10.1002/grl.50398 4. Calais, E. & Minster, J. B. (1995) GPS detection of ionospheric perturbations following 389 the January 17, 1994, Northridge earthquake. Geophysical Research Letters, 22, 1045-390 1048. doi.org: 10.1029/95GL00168 391 5. Cahyadi, M. N. & Heki, K. (2015) Coseismic ionospheric disturbance of the large strike-392 slip earthquakes in North Sumatra in 2012: $M_{\rm w}$ dependence of the disturbance 393 amplitudes, Geophysical Journal International, V. 200, I. 1, doi.org: 10.1093/gji/ggu343 394 6. Dautermann, T., Calais, E., & Mattioli, G. S. (2009). Global Positioning System detection 395 396 and energy estimation of the ionospheric wave caused by the 13 July 2003 explosion of the Soufrière Hills Volcano, Montserrat. Journal of Geophysical Research, 114(B2), B02202. 397 doi.org: 10.1029/2008JB005722 398 7. Emmert, J. T., Drob, D. P., Picone, J. M., Siskind, D. E., Jones, M., Mlynczak, M. G., et 399 al. (2020). NRLMSISE 2.0: A whole-atmosphere empirical model of temperature and 400 401 neutral species densities. Earth and Space Science, 8, e2020EA001321. doi.org: 10.1029/2020EA001321 402 8. Gusman, A.R. & Roger, J. (2022). Hunga Tonga - Hunga Ha'apai volcano-induced sea 403 level oscillations and tsunami simulations. GNS Science webpage, 404 doi.org: 405 10.21420/DYKJ-RK41 9. Heki, K. (2006) Explosion energy of the 2004 eruption of the Asama Volcano, central 406 407 Japan, inferred from ionospheric disturbances, Geophys, Res. Lett., 33, L14303, doi: 10.1029/2006GL026249. 408 10. Kundu, B., Senapati, B., Matsushita, A. et al. (2021) Atmospheric wave energy of the 2020 409 August 4 explosion in Beirut, Lebanon, from ionospheric disturbances. Sci Rep 11, 2793. 410 doi.org: 10.1038/s41598-021-82355-5 411 11. Liu, X., Zhang, Q., Shah, M., & Hong, Z. (2017). Atmospheric- ionospheric disturbances 412 413 following the April 2015 Calbuco volcano from GPS and OMI observations. Advances in Space Research, 60 (12), 2836–2846. doi.org: 10.1016/j.asr.2017.07.007 414

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461 Figures Captions

Figure 1. Scheme of methods developed and implemented in this work. "D1-GNSS-RT" and NRT
 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

- the direction of propagation for the detected disturbances. Based on these results, we compute the source location. NRT TTD estimates TID velocity and verifies the link with the source location.
- Panel (b) shows the difference in parameters between the 1-sec "D1-GNSS-RT" method that was
- developed previously Maletckii and Astafyeva (2021a) and 30-sec "D1-GNSS-RT" that was
- 470 developed and implemented here for the first time.





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



Figure 3. Application of NRT-TTD method to the near-field observations. (a) NRT TTD using 510 30-sec data and the estimated co-VID velocity (black line). Gray vertical line shows the USGS 511 onset time at 04:15UT. The source is located in the Hunga Tonga-Hunga Ha'apai volcano. The 512 black line was fitted by the newly developed automatic NRT-algorithm. (b-c) The two fitting 513 algorithms that estimate the velocity from the TTD slope: (b) the NRT - the brown line, (c) the 514 Quasi-NRT - purple. The blue dots correspond to the first maximums picked in each data series. 515 The red and the green dots are used for the linear regression by the NRT and the Quasi-NRT 516 algorithms, respectively. 517



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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

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



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



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

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

