

1 **Tsunami Source Inversion Using Time-derivative Waveform of Offshore Pressure**

2 **Records to Reduce Effects of Non-Tsunami Components**

3

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15

## 16 **Summary**

17       Offshore ocean bottom pressure gauges (OBPs) are often used to estimate the  
18 spatial distribution of the initial sea-surface height associated with offshore earthquakes  
19 (the tsunami source model). However, the sensors sometimes record pressure changes  
20 that are neither related to tsunamis nor seafloor coseismic displacements (the non-tsunami  
21 components) due to sensor rotation or tilt associated with ground shaking or due to long-  
22 term mechanical drift. These non-tsunami components can be a source of error when  
23 accurately estimating the tsunami source model and thus need to be removed to provide  
24 reliable coastal tsunami forecasts. This paper proposes a new method that uses time-  
25 derivative waveforms of the pressure time series from OBP records to robustly estimate  
26 the tsunami source model, even when OBP data are perturbed by non-tsunami  
27 components. Using OBP data associated with the 2011 Off-Miyagi earthquake (Mw 7.2)  
28 and the 2016 Off-Mie earthquake (Mw 5.9), the performance of the method was evaluated  
29 when reducing artefacts due to non-tsunami components. The tsunami source model was  
30 found to be largely distorted when a conventional inversion method was used (because  
31 of the non-tsunami components). However, the artefact was dramatically reduced when

32 using time-derivative waveforms, and the predicted coastal tsunami waveforms fitted  
33 reasonably with those of observations, thereby suggesting that the new method effectively  
34 suppresses artefacts caused by non-tsunami components. As the tsunami source models  
35 estimated from pressure and time-derivative waveforms should be similar when OBP data  
36 are not perturbed by non-tsunami components, we would be able to assess whether OBP  
37 data are perturbed by non-tsunami components by evaluating that the tsunami source  
38 models estimated from pressure waveforms and from time-derivative waveforms are  
39 similar to each other.

40

41 **Keywords**

42 Tsunamis, Waveform inversion, Tsunami warning, Numerical modelling

## 43 1 INTRODUCTION

44 Offshore real-time tsunami observation networks have been established over the  
45 past few decades (e.g. Kanazawa & Hasegawa 1997; Hino *et al.* 2001; González *et al.*  
46 2005; Kaneda *et al.* 2015; Kawaguchi *et al.* 2015; Kanazawa *et al.* 2016; Uehira *et al.*  
47 2016). A cabled tsunami observation network using ocean bottom pressure gauges  
48 (OBPs), which is known as the Dense Oceanfloor Network System for Earthquakes and  
49 Tsunamis (DONET), has been constructed off southwestern Japan by the Japan Agency  
50 for Marine-Earth Science and Technology (JAMSTEC; Fig. 1a) (Kaneda *et al.* 2015;  
51 Kawaguchi *et al.* 2015). In addition, the National Research Institute for Earth Science and  
52 Disaster Resilience (NIED) has constructed an observation network, which is known as  
53 the Seafloor Observation Network for Earthquakes and Tsunamis along the Japan Trench  
54 (S-net) (Kanazawa *et al.* 2016; Uehira *et al.* 2016) off northeastern Japan. Real-time  
55 tsunami records are often used to provide rapid and reliable tsunami forecasts (e.g. Titov  
56 *et al.* 2005; Tsushima *et al.* 2009, 2012; Baba *et al.* 2014; Gusman *et al.* 2014; Maeda *et*  
57 *al.* 2015; Yamamoto *et al.* 2016a; 2016b; Tanioka 2018). For example, Tsushima *et al.*  
58 (2009; 2012) developed a tsunami forecasting algorithm (the tsunami Forecasting based

59 on Inversion for initial Sea-surface Height; tFISH) that inverts offshore tsunami data to  
60 estimate the spatial distribution of initial sea-surface height (hereafter referred to as the  
61 tsunami source model) and then provide forecasts of coastal tsunamis based on the  
62 forward calculation.

63 Absolute pressure sensors manufactured by Paroscientific, Inc. (e.g. Watts &  
64 Kontoyiannis 1990; Eble & Gonzalez 1991) are commonly used for offshore tsunami  
65 observations (e.g., Kubota *et al.* 2015; 2017a; 2017b; Kaneda *et al.* 2015; Kawaguchi *et*  
66 *al.* 2015; Kanazawa *et al.* 2016; Uehira *et al.* 2016). However, it has been reported that  
67 the pressure outputs of the Paroscientific sensors strongly depend on their orientation  
68 relative to the Earth's gravitational field, and thus their rotation or tilting can become a  
69 source of observational errors (Chadwick *et al.* 2006). Wallace *et al.* (2016) investigated  
70 the Paroscientific OBP data of an Mw 5.9 earthquake that occurred to the southeast off  
71 Mie-Prefecture, Japan, on 1 April 2016 (hereafter referred to as the Off-Mie earthquake,  
72 Fig. 1), and suggested that an OBP observed a pressure offset increase of ~10 hPa nearest  
73 the epicentre (corresponding to 10 cm of subsidence) (KME18 in Fig. 1a), which was  
74 related neither to the tsunami nor to coseismic seafloor displacement and was actually

75 caused by the rotation or tilting of the sensor associated with ground shaking due to  
76 seismic waves. We here note that a pressure change of 1 hPa is equivalent to a water  
77 height change of 1 cm, if assuming a water density of 1.03 g/cm<sup>3</sup> and a gravity  
78 acceleration of 9.8 m/s<sup>2</sup>.

79 Pressure sensors manufactured by Hewlett Packard, Inc. (Karrer & Leach 1969)  
80 (hereafter, HP) have also used for offshore tsunami observations (e.g., Takahashi 1981;  
81 Kanazawa & Hasegawa 1997; Hino *et al.* 2001), although they have been reported to have  
82 long-term mechanical drifts at a maximal rate of approximately 100 hPa/year (Inazu &  
83 Hino 2011). Kubota *et al.* (2017a) investigated HP pressure data associated with a Mw  
84 7.2 earthquake that occurred off Miyagi-Prefecture, Japan, on 9 March 2011 (hereafter  
85 referred to as the Off-Miyagi earthquake, Fig. 2) and found that the HP sensors drifted at  
86 a rate of ~5 hPa/hr (approximately 0.1 hPa/min) within a few hours after the occurrence  
87 of the earthquake. Long-term trends have also been found in Paroscientific sensors, with  
88 rates of less than tens of hPa/year (e.g. Watts & Kontoyiannis 1990; Polster *et al.* 2009;  
89 Inazu & Hino 2011; Hino *et al.* 2014). Pressure offset changes and long-term trends  
90 (hereafter referred to as the non-tsunami components) neither related to tsunamis nor to

91 seafloor permanent displacement are a large source of error when estimating tsunami  
92 source models and providing coastal tsunami forecasts.

93         Some studies have assessed the effects of random observation errors on tsunami  
94 forecasts (Takagawa & Tomita 2014; Tatsumi *et al.* 2014) and dynamic pressure changes  
95 associated with seismic waves (Saito & Tsushima 2016). However, it has not yet been  
96 adequately assessed how the non-tsunami components perturb the tsunami source model,  
97 and the impact of the non-tsunami components on coastal tsunami forecasts has not yet  
98 been investigated. Therefore, to provide accurate coastal tsunami forecasts, it is necessary  
99 to develop a method that reduces the perturbation (i.e., the artefacts) of the tsunami source  
100 model resulting from non-tsunami components. In the present study, we thus propose  
101 such a method that uses time-derivative waveforms of the pressure time series. We also  
102 use OBP data associated with the 2011 Off-Miyagi earthquake and the 2016 Off-Mie  
103 earthquake to assess how the conventional approach used to estimate the tsunami source  
104 model is affected when OBP data are perturbed by the non-tsunami components.  
105 Furthermore, we assess how the new method proposed in this study effectively reduces  
106 the artefacts due to non-tsunami components.

107

## 108 **2 METHODS**

109 The tsunami waveform inversion used to estimate the tsunami source model  
110 (hereafter referred to as the tsunami source inversion) assumes that observed waveforms  
111 can be expressed as a superposition of Green's function from small unit tsunami source  
112 elements (e.g. Baba *et al.* 2005; Tsushima *et al.* 2009; 2012; Kubota *et al.* 2015). Note  
113 that this approach does not estimate the slip distribution along the fault plane, which has  
114 been employed in many previous tsunami inversion studies (e.g., Satake 1989). An  
115 observational equation for the conventional tsunami source inversion using the pressure  
116 time series can be expressed as follows,

$$117 \quad d_j^{\text{obs}}(t) = \sum_{i=1}^M G_{ij}(t)m_i, \quad (1)$$

118 where  $d_j^{\text{obs}}(t)$  is the observed waveform at the  $j$ th station,  $G_{ij}(t)$  is Green's function,  
119 which is the response to the  $i$ th unit source to the  $j$ th station ( $M$  is the total number of unit  
120 sources) and  $m_i$  is the amount of displacement of the  $i$ th unit tsunami source element.  
121 The inversion approach follows the idea that the least-square objective function,  $s(\mathbf{m})$ , is  
122 minimized, which is expressed as,

123 
$$s(\mathbf{m}) = \sum_{j=1}^N |d_j^{obs}(t) - \sum_{i=1}^M G_{ij}(t)m_i|^2 \rightarrow \min, \quad (2)$$

124 where  $N$  denotes the total number of the stations. Equation (1) is expressed in a vector  
 125 form as

126 
$$\mathbf{d}^{obs} = \mathbf{G}\mathbf{m}, \quad (3)$$

127 where  $\mathbf{d}^{obs}$  is a vector consisting of observed pressure data,  $\mathbf{G}$  is a matrix consisting of the  
 128 Green's function and  $\mathbf{m}$  is a vector representing the displacement of the unit source  
 129 elements. In the tsunami source inversion, a spatial smoothing constraint is often imposed  
 130 as follows,

131 
$$\mathbf{0} = \mathbf{S}\mathbf{m}, \quad (4)$$

132 where a matrix,  $\mathbf{S}$ , denotes the spatial smoothing constraints (e.g. Tsushima *et al.* 2009;  
 133 Gusman *et al.* 2013; Kubota *et al.* 2015). Using equations (1) and (2), a normal equation,  
 134 which is to be solved, is expressed as follows,

135 
$$\begin{pmatrix} \mathbf{d}^{obs} \\ \mathbf{0} \end{pmatrix} = \begin{pmatrix} \mathbf{G} \\ \alpha\mathbf{S} \end{pmatrix} \mathbf{m}, \quad (5)$$

136 where a constant  $\alpha$  indicates the weight of the smoothing constraint. Hereafter, we refer  
 137 to this approach as the conventional inversion.

138 When we assume that the vector  $\mathbf{m}$  (the amount of the displacement of the unit

139 sources) does not depend on time, we can obtain the following observational equation by  
 140 a temporally differentiating equation (1) as

$$141 \quad \frac{\partial a_j^{\text{obs}}}{\partial t}(t) = \sum_{i=1}^M \frac{\partial G_{ij}}{\partial t}(t) m_i, \quad (6)$$

142 where  $\partial/\partial t$  denotes the temporal differentiation. Considering the objective function  
 143 (similar to equation (2)) to be minimum, we can obtain the following equation,

$$144 \quad \mathbf{\dot{d}}^{\text{obs}} = \dot{\mathbf{G}} \mathbf{m}, \quad (7)$$

145 where  $\mathbf{\dot{d}}^{\text{obs}}$  and  $\dot{\mathbf{G}}$  denote a vector consisting of the time-derivatives of the pressure data  
 146 (left hand of equation (6)) and Green's function (right hand of equation (6)), respectively.

147 A normal equation can also be expressed as

$$148 \quad \begin{pmatrix} \mathbf{\dot{d}}^{\text{obs}} \\ \mathbf{0} \end{pmatrix} = \begin{pmatrix} \dot{\mathbf{G}} \\ \beta \mathbf{S} \end{pmatrix} \mathbf{m}, \quad (8)$$

149 where  $\beta$  indicates the weight of the smoothing constrain (hereafter this approach is  
 150 referred to as the time-derivative inversion). By solving the normal equations (equations  
 151 (5) and (8)), the tsunami source model (vector  $\mathbf{m}$ ) is obtained.

152 If the pressure data ( $\mathbf{d}^{\text{obs}}$  in equation (1)) is perturbed by non-tsunami components  
 153 (left panel in Fig. 3), the model vector  $\mathbf{m}$  will not reflect the true tsunami source model.

154 Approximating the pressure offset changes resulting from ground shaking as a step

155 function (similar to OBP data at station KME18 in Fig. 1b) and the long-term trends as a  
156 linear function, the associated time-derivative waveforms would be an impulse function  
157 and a constant, respectively (right panel in Fig. 3), and the time-derivative waveform of  
158 the perturbed pressure data ( $\dot{\mathbf{d}}^{\text{obs}}$  in equation (7)) would be very similar to the original  
159 time-derivative waveform. However, the perturbed pressure waveform is quite different  
160 from the original pressure data, and it is thus expected that the time-derivative inversion  
161 would be less affected by non-tsunami components than conventional inversion.

162

### 163 **3 APPLICATION TO 2011 OFF-MIYAGI EARTHQUAKE**

#### 164 **3.1 Data and methods**

165 We applied the time-derivative inversion to OBP data of the Off-Miyagi earthquake  
166 of 9 March 2011 (Kubota *et al.* 2017a, 2017b) to assess the performance of the time-  
167 derivative inversion. Kubota *et al.* (2017a) estimated the finite fault model of this event  
168 (red rectangle in Fig. 2a) by inverting OBP data obtained near the focal area (green  
169 triangles in Fig. 2a). The spatial distribution of the vertical seafloor displacement  
170 calculated from the finite fault model is shown by the black contour lines in Fig. 2a. Using

171 the result of Kubota *et al.* (2017a) as a benchmark, we assessed the performance of the  
172 time-derivative inversion.

173 We used seven Paroscientific OBPs (GJT3, P02, P03, P06, P07, P08 and P09)  
174 installed by Tohoku University and two HP OBPs (TM1 and TM2) installed by the  
175 Earthquake Research Institute (ERI) of the University of Tokyo (green triangles in Fig.  
176 2a; described in detail in Kubota *et al.* (2017a)) (sampling interval of 1 s). We also used  
177 data from coastal GPS buoys (Kato *et al.* 2005; Kawai *et al.* 2012) of the Port and Airport  
178 Research Institute (PARI) of the National Institute of Maritime, Port and Aviation  
179 Technology (MPAT) (yellow squares in Fig. 2a) to assess the performance of the coastal  
180 tsunami forecasts.

181 We processed tsunami data using the following procedure, which is the same as that  
182 presented in Kubota *et al.* (2017a). We removed ocean-tide components using a  
183 theoretical tide model (Matsumoto *et al.* 2000). To reduce the high-frequency pressure  
184 changes attributed to seismic and hydroacoustic waves (e.g., Matsumoto *et al.* 2012; Saito  
185 2013; 2017; Saito & Tsushima 2016; An *et al.* 2017; Kubota *et al.* 2017b), we then  
186 calculated the moving average with a time window of 60 s and applied a causal lowpass

187 filter to the OBP records (a cut-off period of 400 s) and a bandpass filter to GPS buoy  
188 records (passband of 400–3600 s) (Saito 1978). Furthermore, we removed hydrostatic  
189 pressure due to the water column above the OBPs, using the mean from a 20-min time  
190 window recorded prior to the focal time.

191 In the tsunami source inversion, we distributed  $12 \times 16$  small unit source elements  
192 with a size of  $20 \text{ km} \times 20 \text{ km}$  in an area of  $130 \text{ km E-W} \times 170 \text{ km N-S}$  (rectangular area  
193 in Fig. 4a and 4b) with a horizontal spacing of 10 km (overlapping with the adjacent unit  
194 sources). Details of the unit source elements are described in Kubota *et al.* (2015). For  
195 simplicity, displacement of initial sea-surface height was assumed to be equal to seafloor  
196 displacement. We calculated the tsunami Green's function using a linear long wave  
197 equation with a finite difference method in local Cartesian coordinates (e.g. Satake 1995;  
198 Saito *et al.* 2014), and the equations used in this study were as follows,

$$\begin{aligned} 199 \quad \frac{\partial P}{\partial t} &= -g_0 h \frac{\partial \eta}{\partial x}, \\ 200 \quad \frac{\partial Q}{\partial t} &= -g_0 h \frac{\partial \eta}{\partial y}, \quad (8) \\ 201 \quad \frac{\partial \eta}{\partial t} &= -\frac{\partial P}{\partial x} - \frac{\partial Q}{\partial y}, \end{aligned}$$

202 where the parameters  $P$  and  $Q$  are the vertically-averaged horizontal velocity in  $x$ - and  $y$ -

203 directions, respectively; the parameter  $\eta$  is the water height from the static sea surface;  $h$   
204 is water depth and  $g_0$  is the gravitational constant. This equation was discretized on a  
205 staggered spatial grid of 2 km by interpolating ETOPO1 1-arcmin bathymetric data  
206 (Amante & Eakins 2009). The temporal grid interval was set as 1 s. We assumed that  
207 deformation of all unit sources started simultaneously (i.e. an infinite rupture propagation  
208 velocity) and that the duration of the unit source deformation was 0 s. Static pressure  
209 offsets related to seafloor permanent deformation were considered using the method  
210 proposed by Tsushima *et al.* (2012), which subtracts the pressure change components due  
211 to seafloor deformation from the pressure change due to sea-surface fluctuation at OBP  
212 station points (a schematic illustration of this procedure is shown in Fig. S1). When  
213 calculating Green's function for the time-derivative inversion, we calculated the temporal  
214 differentiation of the calculated waveforms. Finally, we applied the same filter as those  
215 applied to observed waveforms.

216 In the inversion, we used a smoothing constraint weight of  $\alpha = 0.5$  for the  
217 conventional inversion and  $\beta = 0.01$  for the time-derivative inversion. These values were  
218 determined so that the maximal displacement of the tsunami source model would be

219 equivalent to that of seafloor vertical deformation calculated using the finite fault model  
220 of Kubota *et al.* (2017a) (black contours in Fig. 2a), which we considered to be the  
221 benchmark.

222

### 223 **3.2 Validation of time-derivative inversion**

224 First, to determine whether the time-derivative inversion could provide the same  
225 performance in resolving the tsunami source model as the conventional inversion, we  
226 analysed the OBP data for the Off-Miyagi earthquake. It was considered that if the OBP  
227 data had not been perturbed by non-tsunami components, then the tsunami source models  
228 estimated by both inversion methods would be similar. We used a time window from 1  
229 to 20 min after the focal time for the inversion (white background area in Fig. 4c and 4d).  
230 Fig. 4a and 4b show the tsunami source models estimated using the conventional  
231 inversion and the time-derivative inversion, respectively; the results are seen to be quite  
232 similar, and both the calculated pressure and time-derivative waveforms agree well with  
233 observations (Fig. 4c and 4d). We measured the agreement between the observed and  
234 calculated waveforms based on variance reduction (VR) as follows,

235 
$$\text{VR} = \left( 1 - \frac{\sum_i \sum_k [d_i^{\text{obs}}(k\Delta t) - d_i^{\text{calc}}(k\Delta t)]^2}{\sum_i \sum_k [d_i^{\text{obs}}(k\Delta t)]^2} \right) \times 100 (\%), \quad (9)$$

236 where  $d_i^{\text{obs}}(k\Delta t)$  and  $d_i^{\text{calc}}(k\Delta t)$  are the observed and calculated OBP data at  $t = k\Delta t$  for  $i$ th  
 237 OBP station, respectively ( $\Delta t$  is the sampling interval). We used a time window of 1 to 20  
 238 min after the focal time to calculate the VR, and obtained relatively high VRs for both  
 239 pressure and time-derivative waveforms from both the conventional and time-derivative  
 240 inversions (pressure waveform: 99.3% using the conventional inversion and 96.6% using  
 241 the time-derivative inversion; time-derivative waveform: 97.2% using the conventional  
 242 inversion and 97.3% using the time-derivative inversion).

243 Both inversion results effectively reproduce the leading tsunami waves observed by  
 244 coastal GPS buoy waveforms from approximately 0–40 min (Fig. 4e). The discrepancy  
 245 in the latter part of the GPS buoy waveforms (after  $\sim 40$  min) is probably related to the  
 246 nonlinearity and a lack of fine-scale bathymetry near the coast (e.g. Satake 1995; Saito *et*  
 247 *al.* 2014). Fig. 5a and 5b show comparisons of arrival times and maximal tsunami heights  
 248 of the leading wave between observed and calculated tsunami waveforms (the arrival time  
 249 was defined as the time when the amplitude exceeded 1 cm). The arrival times and  
 250 maximal heights of the conventional inversion (blue bars in Fig. 5a and 5b), the time-

251 derivative inversion (red bars) and the observations (black bars) are all very close to one  
252 another. Based on these results, we thus concluded that the time-derivative inversion  
253 provided a performance as good as the conventional inversion in estimating the tsunami  
254 source model, when the OBP data are not perturbed by the non-tsunami components.

255

### 256 **3.3 Synthetic test using datasets with non-tsunami components**

257 We then assessed how the tsunami source model obtained using conventional  
258 inversion is perturbed by non-tsunami components (when OBP data contain non-tsunami  
259 components), and assessed how use of the time-derivative inversion reduces artefacts due  
260 to non-tsunami components. We prepared synthetic datasets by adding artificial pressure  
261 offset changes to observed OBP data from the 2011 Off-Miyagi earthquake (i.e., pressure  
262 data was artificially perturbed) and assuming pressure changes of 20–50 hPa (Fig. 6,  
263 Table 1), which correspond to those due to the rotation of Paroscientific pressure sensors  
264 with rotation angles of  $\sim 30\text{--}90^\circ$  (Chadwick *et al.* 2006). We assumed the pressure offset  
265 change was a ramp function with a finite duration of  $T_{\text{offset}} = 10$  s, in consideration of the  
266 duration of strong ground shaking (a few tens of seconds). The pressure offset change

267  $p^{\text{offset}}(t)$  is expressed as follows,

$$268 \quad p^{\text{offset}}(t) = \begin{cases} 0 & (t \leq 0) \\ p_o \times \frac{t}{T} & (0 < t \leq T_{\text{offset}}), \\ p_o & (T_{\text{offset}} < t) \end{cases} \quad (10)$$

269 where  $p_o$  is the given pressure offset value (which is summarised in Table 1). After

270 perturbing the pressure data, they were then processed using the same method as that

271 applied to pressure data without the perturbation (hereafter referred to as the original data).

272 After data processing, we considered the perturbed pressure data to be the observed data

273 and estimated the tsunami source model. All other settings were the same as those

274 employed in the original analysis described in the previous section. Note that the first 1-

275 min of data were not used for the inversion because of the instability of the pressure data.

276 The inversion results are shown in Fig. 6., where it is evident that the estimation of

277 the tsunami source model with the conventional inversion (Fig. 6a) is quite different from

278 that estimated using original data (Fig. 4). The pressure waveforms calculated from the

279 tsunami source model obtained by the conventional inversion (blue lines in Fig. 6c)

280 explain the artificially-perturbed (i.e., observed) pressure waveforms (grey dashed lines

281 in Fig. 6c) very well (VR = 99%), but the original pressure waveforms (black lines) are

282 not explained at all (VR = -1466%). However, the time-derivative calculated pressure

283 waveforms from the tsunami source model (blue lines in Fig. 6d) do not explain the initial  
284 part (< approximately 5 min) of the time-derivative waveforms relating to original  
285 pressure waveforms, but the latter part is reasonably explained (VR = -142%). Although  
286 the tsunami source model obtained using the time-derivative inversion (Fig. 6b) is similar  
287 to that obtained by original data, the results are not exactly the same. The pressure  
288 waveforms calculated from this tsunami source model explain the original pressure  
289 waveforms reasonably well (red lines in Fig. 6c, VR = 44.9%), and the discrepancy found  
290 between the time-derivative waveforms of observational and synthetic data is much  
291 smaller than that for the conventional inversion (red lines in Fig. 6d, VR = 80.1%). These  
292 results show that the conventional inversion is unable to remove the artefacts due to  
293 pressure offset changes, whereas the time-derivative inversion dramatically reduces them.

294         Although the artefact is dramatically reduced by the time-derivative inversion,  
295 artefacts due to the offset changes are not completely removed; this is considered likely  
296 to be related to the temporal smoothing effect due to the moving average and the low-  
297 pass filter (grey dashed lines in Fig. 6d). However, we find that the forecasted arrival time  
298 and maximal amplitude at the coastal GPS buoys (blue bars in Fig. 5c and 5d) tend to be

299 early and large (by ~5–10 min and ~5 cm, respectively) compared to observations (black  
300 bars) when conventional inversion is used, but they are reasonably explained when the  
301 time-derivative inversion is used (red bars in Fig. 5c and 5d).

302 We also conducted tests assuming a linear pressure trend at a rate of 0.5 hPa/min  
303 (Fig. S2) and smaller pressure offset values (less than 10 hPa) (Figs S3 and S4). All results  
304 show that artefacts due to non-tsunami components are reduced well using the time-  
305 derivative inversion; however, these results are not achieved when using the conventional  
306 inversion. The forecast tsunami arrival time and maximal height of the GPS buoys using  
307 the synthetic dataset containing linear trends (Fig. S2) are shown in Fig. 5e and 5f,  
308 respectively. When using the conventional inversion, the forecast arrival time is  
309 approximately 10 min earlier than the observation, whereas it is nearly similar to the  
310 observation when the time-derivative inversion is used. These synthetic tests thus  
311 demonstrate that the time-derivative inversion effectively reduces the artefacts in the  
312 tsunami source model due to the non-tsunami components and improves the forecast of  
313 the arrival time and maximal height of the coastal tsunami.

314

## 315 4 APPLICATION TO 2016 OFF-MIE EARTHQUAKE

### 316 4.1 Data and analysis

317 In this section, we report results of applying the time-derivative inversion to OBP  
318 data from the Mw 5.9 Off-Mie earthquake (Wallace *et al.* 2016; Asano 2018; Nakano *et*  
319 *al.* 2018; Takemura *et al.* 2018). The pressure changes due to the tsunami with a maximal  
320 amplitude of  $\sim 2$  hPa (equivalent to a tsunami of  $\sim 2$  cm) were clearly observed by the  
321 DONET OBPs, and a few hPa of pressure offset-level changes were also observed at  
322 DONET1 stations near the epicentre (for example, KME17, KM19, KME20, and  
323 KME22) (Fig. 1b). One OBP station nearest the epicentre (KME18) observed a large  
324 pressure offset change of approximately 10 hPa, which could be attributed to the tilting  
325 or rotation of the sensors in relation to strong ground shaking, as noted by Wallace *et al.*  
326 (2016). Kubo *et al.* (2018) investigated the site amplification characteristics of DONET1  
327 stations and found that station groups KMA and KME (blue and red inverted triangles in  
328 Fig. 1a) had large site amplifications due to thick subseafloor sediments. Kubo *et al.*  
329 (2018) and Nakamura *et al.* (2018) reported peak ground accelerations (PGAs) of  $\sim 700$   
330 gal by DONET strong motion seismometers at KME18 during the 2016 Off-Mie

331 earthquake, and also found that the site amplification observed at KME18 during this  
332 event was more than 40 times larger than that expected from the empirical relation. In  
333 addition, Kubo *et al.* (2018) suggested that a nonlinear soil response occurred at DONET1  
334 seismometers near the epicentre. These results support the idea that the pressure  
335 waveform at KME18 station is perturbed by non-tsunami components due to strong  
336 ground shaking. Therefore, when estimating the tsunami source model, data from the  
337 OBP at station KME18 were excluded.

338 We processed DONET OBP data using the same method as used with the 2011 Off-  
339 Miyagi earthquake (cut-off period of the low-pass filter was 60 s). We estimated the  
340 tsunami source model (Fig. 7) by manually selecting OBP stations and time windows  
341 used for inversion based on a visual inspection of OBP waveforms (drawn by thick black  
342 lines in Fig. S6). As it was suspected that DONET1 OBP waveforms at stations near the  
343 epicentre were also perturbed by non-tsunami components (as with station KME18), due  
344 to the large peak ground acceleration during the earthquake (Kubo *et al.* 2018; Nakamura  
345 *et al.*, 2018), OBP data from stations KMA03, KMD15, KMD16, KME17, KME19,  
346 KME20 and KME22 were also excluded (in addition to KME18) from analysis (grey

347 inverted triangles in Fig. 7). Furthermore, coastal tsunami data were not used to discuss  
348 the accuracy of the coastal tsunami forecast, because the observed tsunami height at the  
349 coast was very small (less than a few cm). We set the analytical area as  $100 \text{ km} \times 100 \text{ km}$ .  
350 To avoid both over-fitting and over-smoothing during analysis, we used smoothing  
351 constraint weights of  $\alpha = 0.5$  for the conventional inversion and  $\beta = 0.01$  for the time-  
352 derivative inversion, which were determined based on the trade-off curve between the  
353 smoothing weight and the VR values (Fig. S5).

354

## 355 **4.2 Results**

356 The estimated tsunami source models obtained from conventional and time-  
357 derivative inversions were found to be similar to each other (Fig. 7a and 7b, respectively),  
358 and a pair of uplift and subsidence areas with maximal amplitudes of approximately +3  
359 cm and -2 cm, respectively, were estimated. To compare the tsunami source model with  
360 the seismic analysis, we calculated the seafloor vertical displacement based on the  
361 centroid moment tensor (CMT) solution of the U.S. Geological Survey (USGS). In this  
362 calculation, we assumed one planar rectangular fault such that its centre coincided with

363 the USGS centroid. The fault length, width, and slip amount were assumed using the  
364 scaling law of Wells & Coppersmith (1994), and vertical displacement was calculated  
365 using the equations of Okada (1992). We obtained a maximal seafloor vertical  
366 deformation displacement of approximately 3 cm (green contours in Fig. 7), which is  
367 consistent with that of the tsunami source model. The vertical displacement of the tsunami  
368 source model at OBP station KME18 was approximately +1 cm, which is much smaller  
369 than that expected from the observed pressure change ( $\sim -10$  cm). This indicates that the  
370 large offset pressure change at station KME18 was neither due to the tsunami nor to  
371 seafloor permanent displacement, but due to non-tsunami components. The strike of the  
372 hinge-line between the uplift and subsidence is consistent with the strike angle of the  
373 finite fault model of Wallace *et al.* (2016) ( $215^\circ$ , yellow rectangles in Fig. 7a and 7b).  
374 The vertical displacements expected from our tsunami source model at stations KMA03,  
375 KME17 and KME22 are less than 1 cm (Fig. S6a) and are therefore inconsistent with the  
376 observed pressure changes (displacements of approximately 1 cm). However, the arrival  
377 times, amplitudes and durations of the tsunami are reasonably consistent with those of  
378 the observation. In addition, the expected coastal tsunami heights of the conventional and

379 time-derivative inversions are similar (e.g. approximately 1.5 cm at station 301) (Fig.  
380 S6e).

381 To investigate the discrepancies of pressure offset changes at stations KMA03,  
382 KME17 and KME22 between calculations and observations, we conducted an inversion  
383 that included OBP data obtained near the epicentre (KME18 was excluded from analysis)  
384 (Figs S7 and S8) and compared the result with the tsunami source model excluding the  
385 stations near the epicentre (Fig. 7, hereafter referred to as the reference tsunami source  
386 model). We found that the tsunami source model obtained using the time-derivative  
387 inversion (Fig. S7b) was similar to the reference tsunami source model, whereas the  
388 tsunami source model using the conventional inversion had a wider subsidence area  
389 extending around stations KMA03, KME17 and KME22 (Fig. S7a). Since the tsunami  
390 source models using the conventional and time-derivative inversions should resemble one  
391 another when OBP data are not perturbed by non-tsunami components (as shown in the  
392 previous section), the discrepancy of the tsunami source models suggests that OBP data  
393 from stations KMA03, KME17 and KME22 are also perturbed by non-tsunami  
394 components, although the amplitudes are very small (approximately 1 cm).

395           Tsuji *et al.* (2017) interpreted the rupture process of this earthquake as being related  
396 to an ancient splay fault system in an accretionary prism (Tsuji *et al.* 2014), based on the  
397 comparison of the strike angles between of the ancient splay fault and of the finite fault  
398 model of Wallace *et al.* (2016) (yellow rectangle in Fig. 7). However, as the tsunami  
399 source model estimated using the conventional inversion strongly depends on non-  
400 tsunami components, it is difficult to discuss the rupture process of the Off-Mie  
401 earthquake, as previously discussed by Tsuji *et al.* (2014), whereas we can discuss it by  
402 using the time-derivative inversion, which can reduce the artefacts of the tsunami source  
403 model. In addition, conducting a comparison between tsunami source models from both  
404 inversions is effective for distinguishing whether or not near-source OBPs contain the  
405 non-tsunami component. The time-derivative inversion is thus useful for discussing the  
406 detailed rupture processes of tsunami-associated earthquakes, which is not easily  
407 achieved using the conventional inversion alone.

408           For a real-time tsunami forecast, we conducted a tsunami source inversion using  
409 only the early part of DONET OBP data (a time window from 1 to 5 min following the  
410 focal time) (Fig. 8). As there was no time to inspect OBP data to exclude waveforms

411 containing non-tsunami components from the tsunami source inversion, we also used  
412 OBP data from station KME18, which contained a large apparent pressure offset change.  
413 When using the conventional inversion, a large amount of subsidence (approximately -  
414 10 cm) was estimated around station KME18 (Fig. 8a) and results from the tsunami  
415 source model were found to be quite different from those of the reference tsunami source  
416 model (Fig. 7). However, results from the tsunami source model using the time-derivative  
417 inversion (Fig. 8b) were very similar to those of the reference tsunami source model,  
418 which suggests that the time-derivative inversion effectively reduces non-tsunami  
419 components, even when providing a real-time analysis. The latter part of OBP waveforms  
420 calculated from the tsunami source model obtained by conventional inversion (blue lines  
421 in Fig. S8) do not match the observations at all, whereas that using the time-derivative  
422 inversion provide a reasonable fit (red lines in Fig. S9). Although the expected coastal  
423 tsunami height is only a few cm, which is less than the noise level (Fig. S9e), the expected  
424 maximal amplitudes at coastal stations using the conventional inversion (blue lines in Fig.  
425 S9e) are nearly twice as large as the forecast using the time-derivative inversion (red  
426 lines).

427

## 428 **5 DISCUSSION**

429 To provide an accurate and reliable tsunami forecast, it is important to quickly  
430 obtain highly accurate information from the tsunami source model. We thus investigated  
431 the relationship between the tsunami source inversion and the end time of the inversion  
432 time window using OBP data associated with the 2011 Off-Miyagi earthquake (Fig. 9).  
433 Changing the end time of the inversion time window from 2 min to 20 min after the focal  
434 time (the start time of the time window was fixed to 1 min), we conducted a tsunami  
435 source inversion. The other settings were the same as those of the original analysis.

436 To evaluate the temporal stability of the inversion, we investigated the temporal  
437 evolution of the VR for the observed and calculated waveforms using a time window of  
438 1–20 min (Fig. 9a). We also calculated the temporal evolution of the total volume of  
439 displaced seawater ( $V$ ) (Fig. 9b), which is defined as follows,

$$440 \quad V = \sum_i \sum_j |u_{ij}| \times \Delta x \times \Delta y, \quad (11)$$

441 where  $u_{ij}$  is the displacement of the tsunami source at the  $(i, j)$ th grid in the  $x$  and  $y$   
442 directions ( $\Delta x$  and  $\Delta y$  are the horizontal grid intervals of 2 km). The temporal evolution

443 of the VR for pressure waveforms and total volumes was found to be quite similar for the  
444 two inversion methods; however, the temporal evolution of the total volume was stable  
445 after 10 min when the time-derivative inversion was used (Fig. 9b). These results suggest  
446 that the time-derivative inversion is not necessarily better than the conventional inversion  
447 with respect to convergence time, but it is slightly better in terms of stability.

448         There would be another approach to simultaneously estimate the tsunami source  
449 model and the non-tsunami component at each OBP station from pressure waveforms.  
450 The advantage of such approach would be that the first few minutes of data following the  
451 occurrence of the earthquake could be used to estimate the tsunami source model,  
452 although our approach did not use. However, such an approach should have a trade-off  
453 between the estimated seafloor displacement and the estimated non-tsunami components  
454 (an example of this possible trade-off is shown in Fig. S10). If a shorter time window  
455 were used that did not include the peak amplitude of the tsunami, it would be possible  
456 that the gradual pressure change associated with the tsunami is wrongly estimated as a  
457 linear trend. This would be a disadvantage to provide fast ( $< \sim 10$  min from the  
458 earthquake) and reliable tsunami forecasts. Our approach using time-derivative

459 waveforms is advantageous in avoiding such trade-off, because only the displacement of  
460 the unit tsunami source elements are the unknown parameters.

461         In a practical tsunami forecast, we also need to consider the artefacts due to high-  
462 frequency pressure changes associated with seismic and hydroacoustic waves (e.g.,  
463 Matsumoto *et al.* 2012; Saito 2013; 2017; An *et al.* 2017; Kubota *et al.* 2017b); although  
464 these effects are not included in the analysis. As the dominant period of hydroacoustic  
465 waves is less than  $\sim 10$  s (e.g., Matsumoto *et al.* 2012; Saito 2013; 2017), whereas tsunami  
466 waves have much longer dominant periods ( $> \sim 100$  s), hydroacoustic components can be  
467 removed from OBP waveforms by applying a lowpass filter with an appropriate cut-off  
468 period. In addition, previous studies have also reported dynamic pressure changes caused  
469 by the reaction force from the seawater to the seafloor (in response to the seafloor  
470 accelerating the seawater during seafloor displacement) (a dominant period of  $< \sim 50$  s),  
471 (e.g., Saito 2017; An *et al.* 2017; Kubota *et al.* 2017b). However, although it appears that  
472 this component may affect the inversion, Saito & Tsushima (2016) found that the effects  
473 are only minimal, because such short-period pressure components cannot be expressed  
474 by the superposition of Green's function of the tsunami (which has much longer dominant

475 periods ( $> \sim 100$  s)). We also note that in the practical tsunami forecast the consideration  
476 of the additional time to process OBP data is required. But it would not be a major concern  
477 with respect to the proficiency of contemporary hardware, and will be even less of a  
478 concern when using high-performance computers developed in the future.

479         In the investigation of the temporal evolution of inversion stability, we used OBP  
480 data that were not perturbed by non-tsunami components. Our results showed similar  
481 temporal evolutions for both the conventional and time-derivative inversions. In the  
482 synthetic test we assumed that the pressure data were perturbed by non-tsunami  
483 components and found that the tsunami source models for conventional and time-  
484 derivative inversions were very different. Based on these results, it would be very useful  
485 to compare tsunami source models using conventional and the time-derivative inversions  
486 to enable a real-time validation of pressure data quality and to distinguish whether or not  
487 pressure data contain non-tsunami components in real time.

488

## 489 **6 CONCLUSIONS**

490         We propose a new method using the time-derivative waveforms of the pressure

491 time series (rather than the raw pressure time series) to estimate the spatial distribution of  
492 initial sea-surface height (the tsunami source model) using OBP data, with the aim of  
493 reducing artefacts due to non-tsunami pressure components. Using OBP data associated  
494 with the Off-Miyagi earthquake that occurred on 9 March 2011 (Mw 7.2), the proposed  
495 method was found to work as well as the conventional method. We also conducted a  
496 performance test using a synthetic dataset and artificially perturbing OBP data. The  
497 tsunami source model obtained using the conventional inversion approach provided large  
498 seafloor displacements around OBPs due to artificial non-tsunami components, and the  
499 forecast coastal tsunami arrived earlier and had a larger amplitude than the observation.  
500 However, when time-derivative waveforms were used for the inversion, artefacts due to  
501 non-tsunami components were dramatically suppressed, and the forecast coastal tsunami  
502 waveforms reasonably matched those of the observation.

503 We also applied the new method to OBP data associated with the 2016 Off-Mie  
504 earthquake (Mw 5.9), and the estimated tsunami source model was found to be consistent  
505 with the USGS CMT solution. The tsunami source model also suggested that OBPs near  
506 the epicentre contained non-tsunami components (with an amplitude of approximately 1-

507 cm) because of sensor tilting or rotation. We then analysed OBP data based on quasi-real-  
508 time analysis, and the estimated tsunami source model obtained using the conventional  
509 method provided very different results from those obtained using post analysis. However,  
510 the newly developed tsunami source model provided results that were quite similar to  
511 those obtained by careful post analysis, even when including OBP data perturbed by non-  
512 tsunami components.

513 We assessed the time window used for inversion to discuss the temporal stability  
514 of the inversion and found that the tsunami source model obtained using the time-  
515 derivative inversion was stable after inversion convergence (~ 10 min from the focal time),  
516 whereas the total volume of displaced seawater was unstable when the conventional  
517 method was used. For practical tsunami forecasting, it would be useful to compare  
518 tsunami source models using both inversion methods to validate real-time OBP data  
519 quality, as it is considered that the methods would provide identical results if OBP data  
520 are not perturbed by non-tsunami components and the results will be different if OBP  
521 data are perturbed.

522

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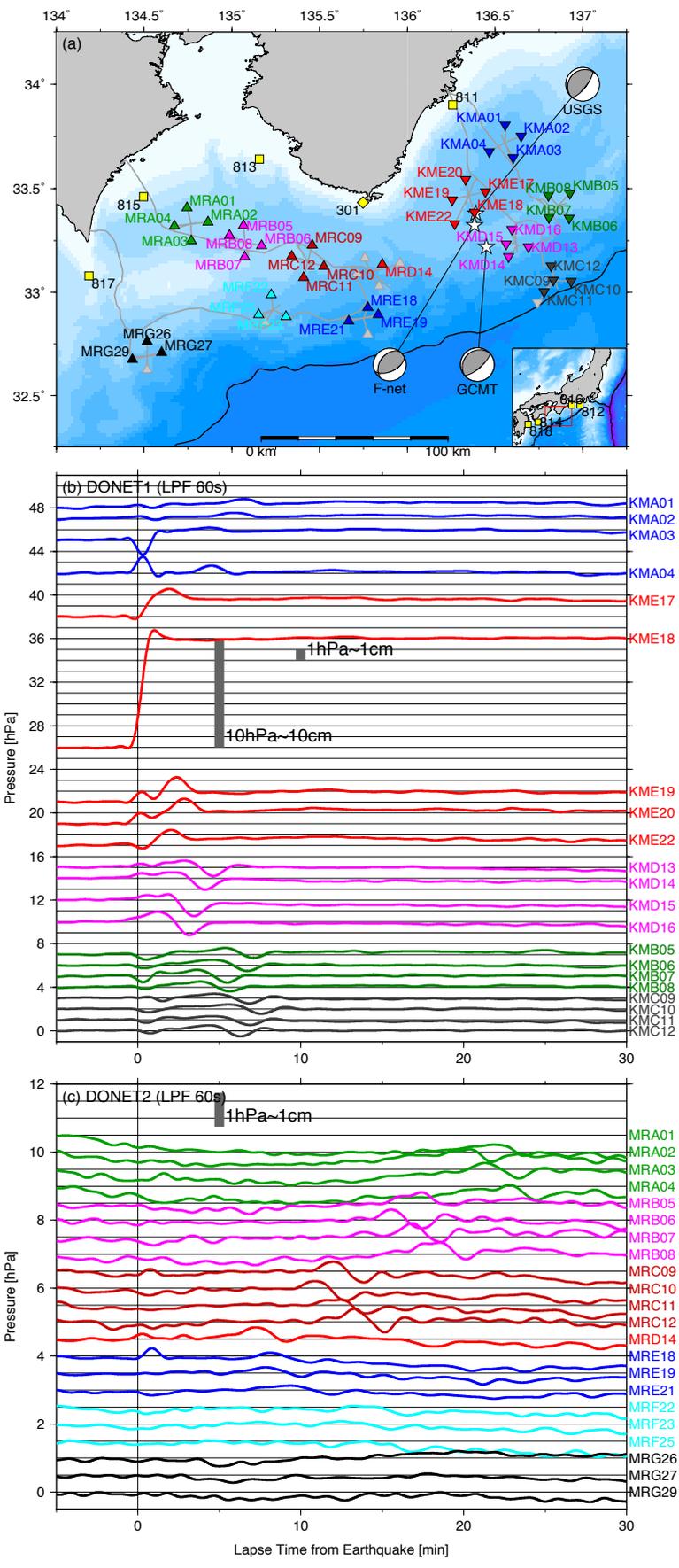
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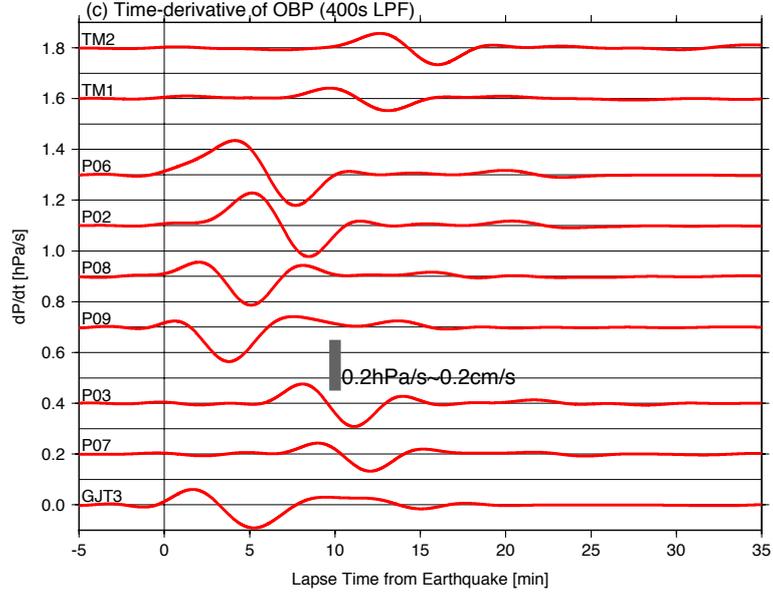
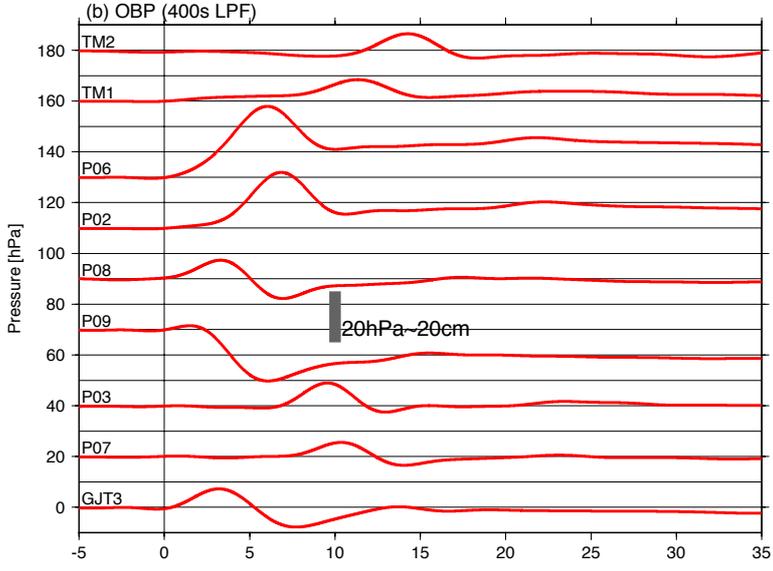
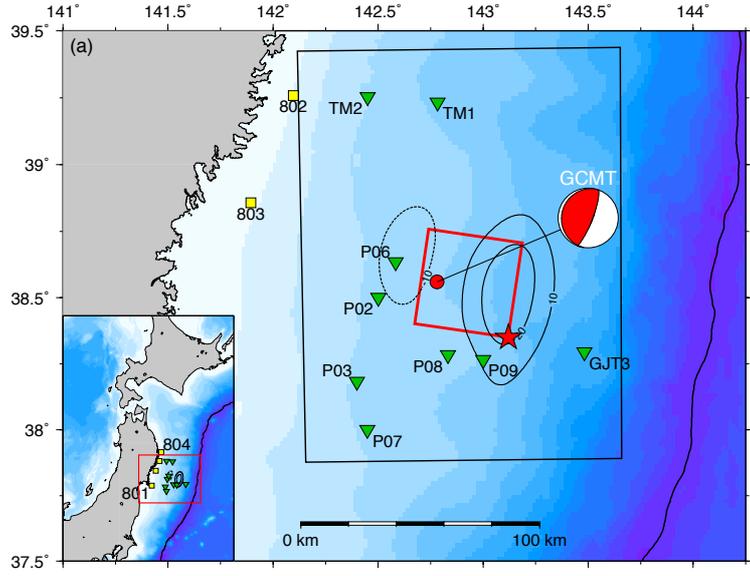
711

712 **Table 1.** Perturbations to OBP data in synthetic test

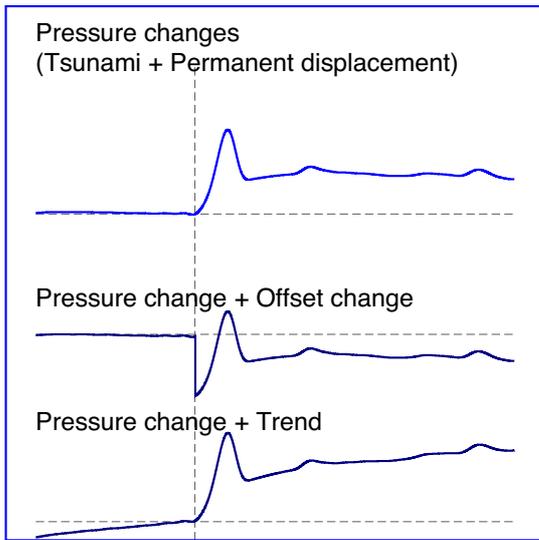
Case	Original	Large offset	Trend	Small offset	Offset at P06
Unit	hPa	hPa	hPa/min	hPa	hPa
	Fig. 4	Fig. 6	Fig. S2	Fig. S3	Fig. S4
GJT3	0	+30	+0.5	+10	0
P02	0	-30	+0.5	+5	0
P03	0	+20	+0.5	+5	0
P06	0	-40	-0.5	+5	-30
P07	0	-20	-0.5	+5	0
P08	0	+50	-0.5	+10	0
P09	0	-50	-0.5	+15	0
TM1	0	0	+0.5	0	0
TM2	0	0	-0.5	0	0



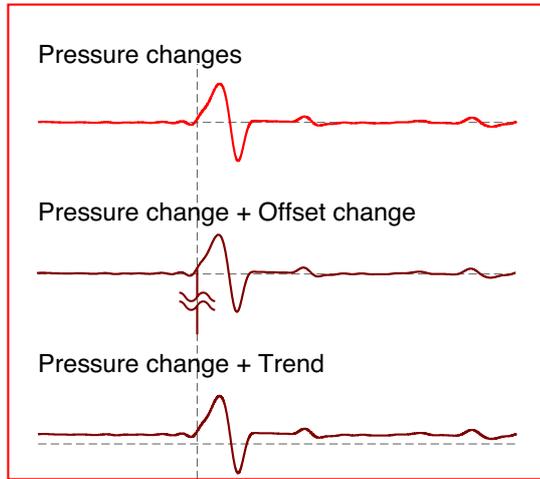
715 **Figure 1.** (a) Location map of the 2016 Off-Mie Earthquake (Mw 5.9). Stars denote  
716 centroids of CMT solutions from Global CMT (Ekström *et al.* 2012), F-net (Kubo *et al.*  
717 2002) and USGS. DONET1 and DONET2 OBP stations are denoted by inverted and  
718 regular triangles, respectively. OBP symbol colours denote station groups. Each OBP  
719 sensor within a group is connected to the same science node, which is a device with the  
720 function of a hub that connects the sensors to the main cable system (Kaneda *et al.* 2015;  
721 Kawaguchi *et al.* 2015). GPS buoys and coastal wave gauges are shown by yellow squares  
722 and diamonds, respectively. Pressure time series observed by (b) DONET1 and (c)  
723 DONET2 OBPs.



725 **Figure 2.** (a) Location map of the 2011 Off-Miyagi earthquake (Mw 7.2). Epicentre  
726 (Suzuki *et al.* 2012) and main rupture area (Kubota *et al.* 2017a) are denoted by red star  
727 and rectangle, respectively. Global CMT solution is also shown. Black contour lines are  
728 seafloor vertical displacement calculated using fault model of Kubota *et al.* (2017a); black  
729 rectangle is analysis area for tsunami source inversion; green triangles and yellow squares  
730 denote OBP stations and coastal GPS buoys, respectively. (b) Observed pressure time  
731 series. (c) Time-derivative waveforms of OBP records.



Time derivative waveforms

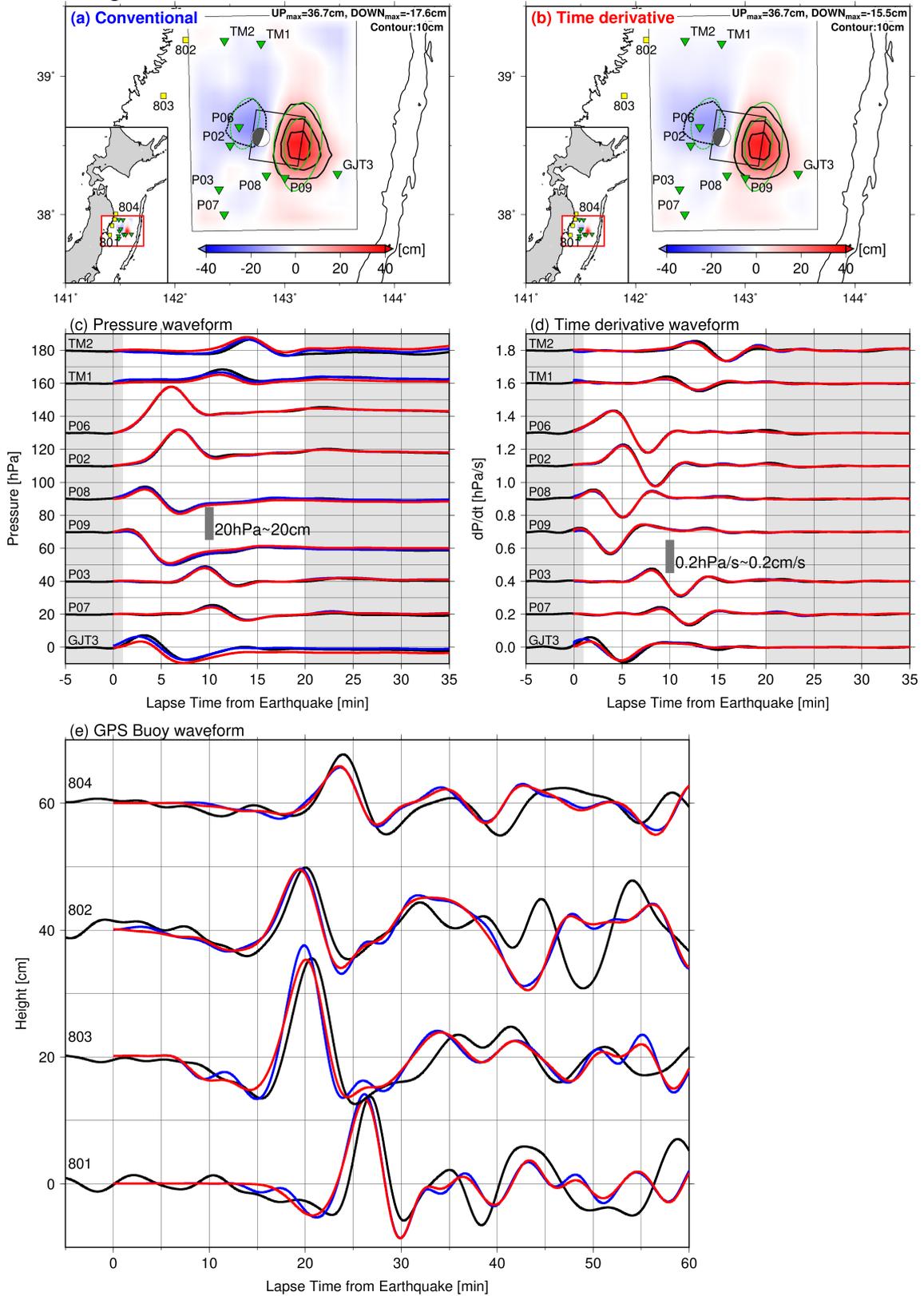


732

733 **Figure 3.** Schematic image of seafloor pressure change associated with earthquakes and

734 non-tsunami components.

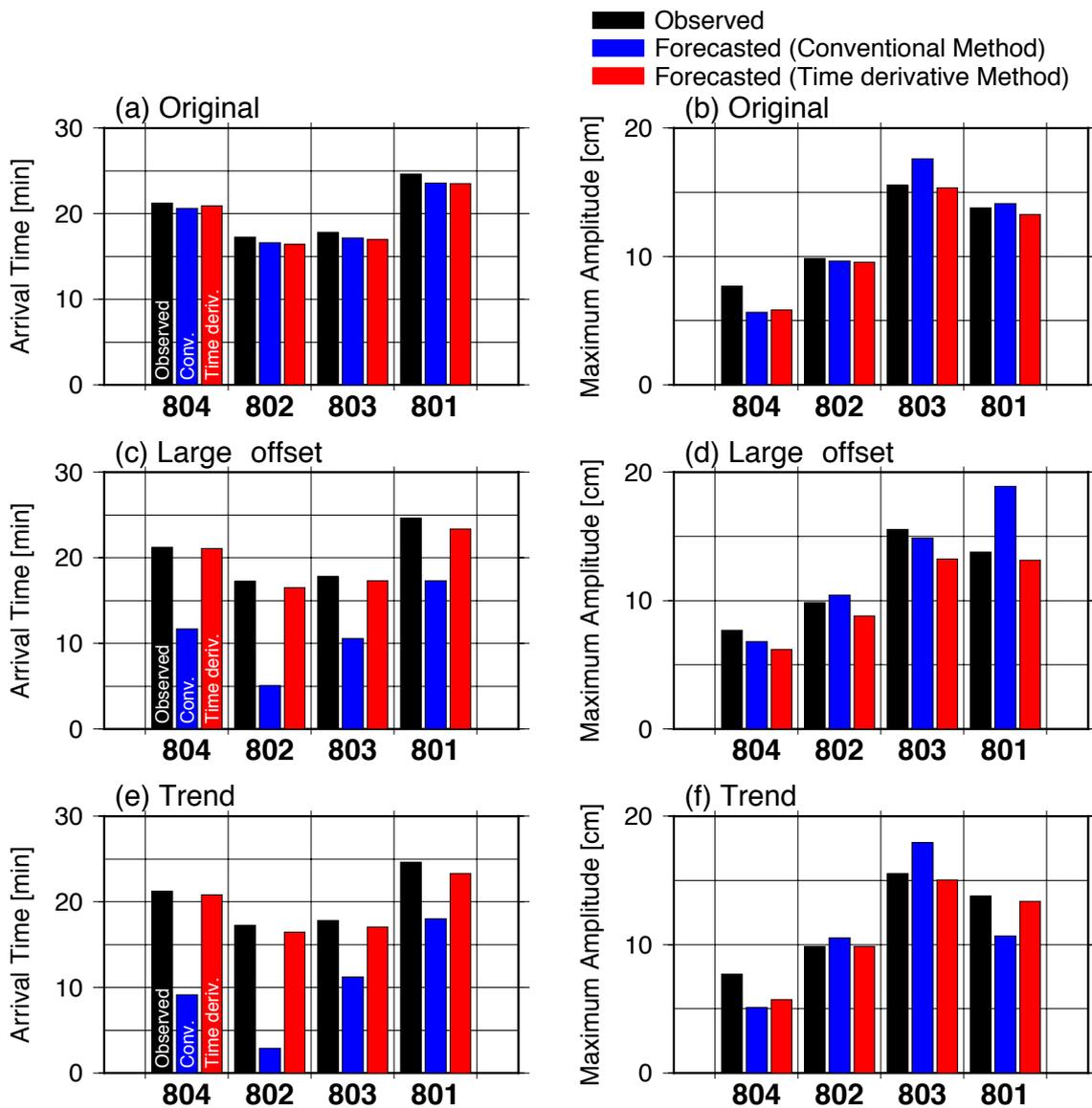
### Original data



735

736

737 **Figure 4.** Inversion results for the 2011 Off-Miyagi earthquake using OBP data without  
738 adding artificial perturbation (original data). Tsunami source models obtained using (a)  
739 conventional inversion and (b) time-derivative inversion. Uplifted and subsided areas are  
740 shown in red and blue, respectively; green contour lines are the seafloor vertical  
741 displacement expected using the fault model of Kubota *et al.* (2017a) with 10-cm intervals.  
742 Comparisons of (c) pressure and (d) time-derivative waveforms between observed  
743 waveforms (black) and calculated waveforms. Waveforms calculated from source models  
744 obtained using conventional and time-derivative inversions are shown in blue and red,  
745 respectively. A time window from 1 to 20 min (white background area) was used in the  
746 inversion. (e) Comparison of coastal GPS buoy waveforms between observed (black) and  
747 forecast (blue and red) waveforms.



748

749 **Figure 5.** Comparison of arrival times of first waves and maximal amplitudes between

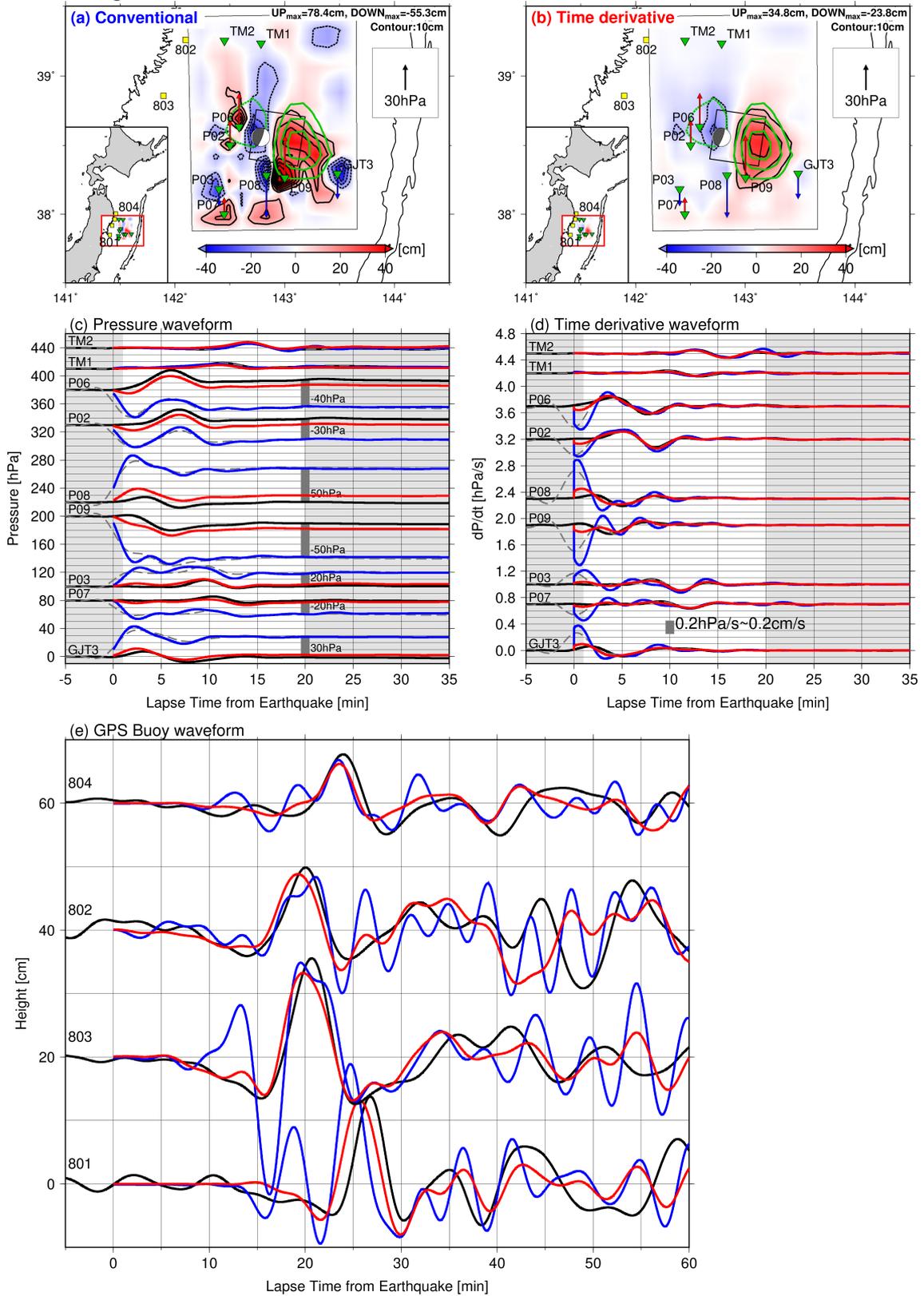
750 observation (black) and forecasts from conventional inversion (blue) and time-derivative

751 inversion (red), using (a, b) original pressure data (Fig. 4), (c, d) synthetic pressure data

752 with pressure offset changes (Fig. 6) and (e, f) synthetic pressure data with a long-term

753 trend (Fig. S2).

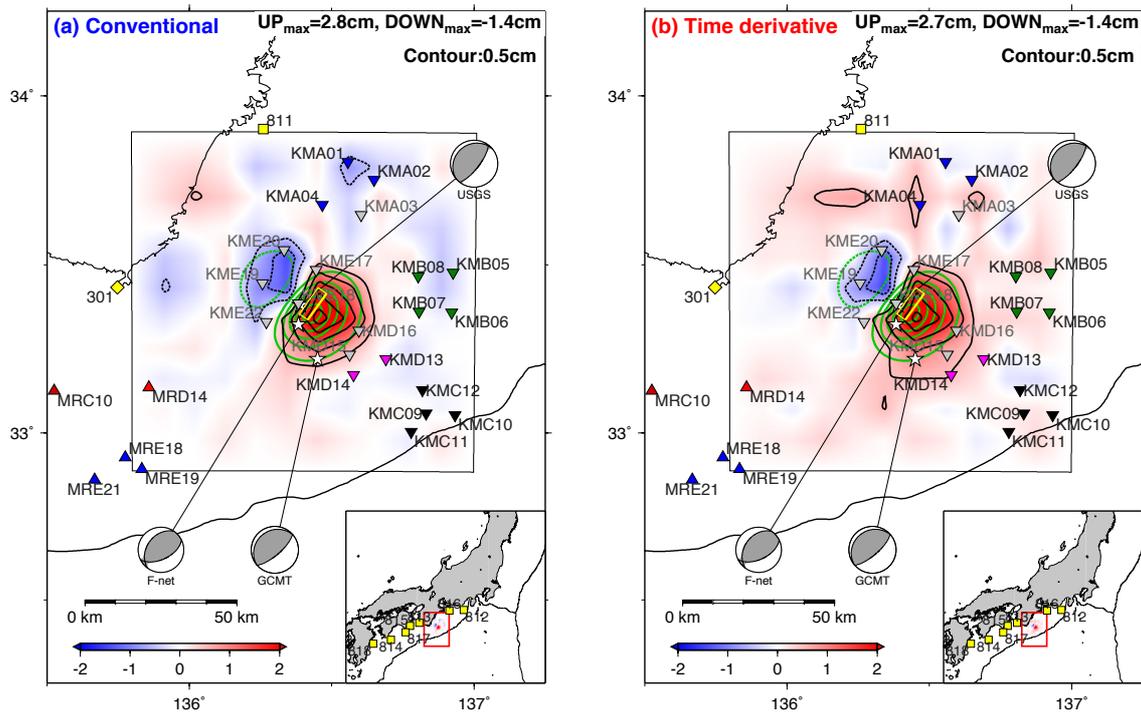
### Large offset



754

755

756 **Figure 6.** Inversion results of the 2011 Off-Miyagi earthquake using synthetic OBP data  
757 containing artificial pressure offset changes. Tsunami source model from OBP data  
758 without artificial perturbation using conventional inversion (Fig. 4a) shown by green  
759 contours; grey dashed lines in (c) and (d) are synthetic data used in analysis; other  
760 explanations are same as those in Fig. 4.



761

762 **Figure 7.** Tsunami source model of the 2016 Off-Mie earthquake obtained using (a)

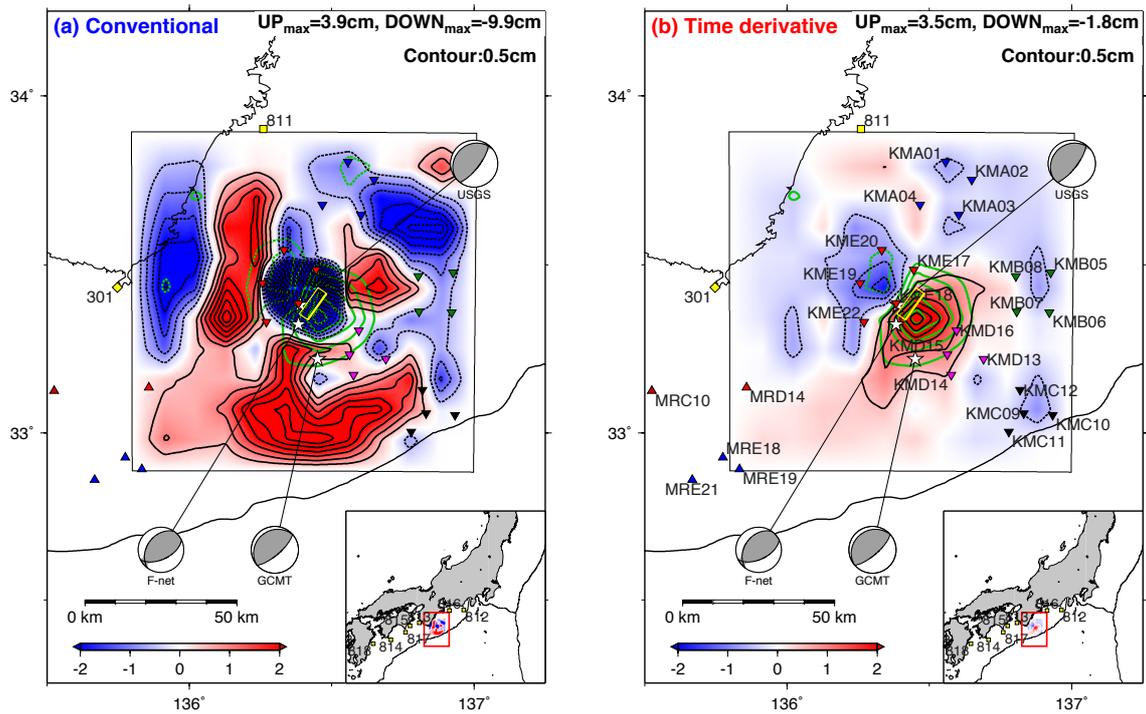
763 conventional and (b) time-derivative inversions, without OBP data near the epicentre

764 (OBPs not used in analysis are shown in grey). Colours of OBPs used in the inversion are

765 the same as Fig. 1. The interval of the contour lines is 0.5 cm. Green contours denote

766 seafloor vertical displacement expected from the USGS CMT solution; yellow rectangles

767 denote finite fault model of Wallace *et al.* (2016).



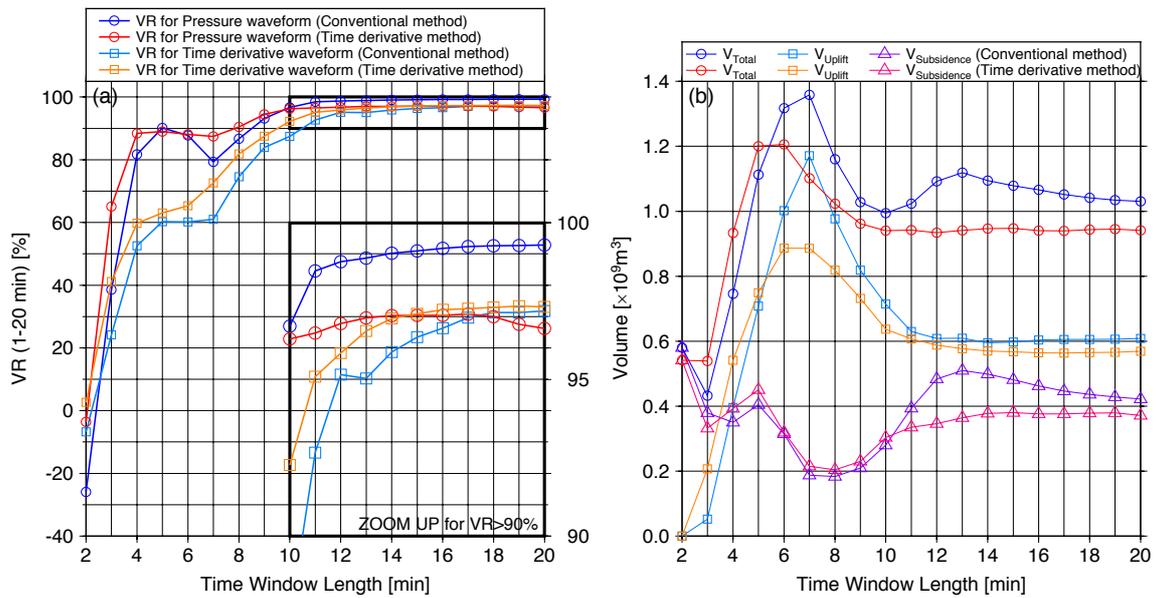
768

769 **Figure 8.** Tsunami source model of the 2016 Off-Mie earthquake obtained using (a)

770 conventional and (b) time-derivative inversions with all the OBPs and a time window of

771 1 to 5 min. Green contours denote seafloor vertical displacement obtained in post analysis

772 (Fig. 7a); other explanations are same as those in Fig. 7.



773

774 **Figure 9.** (a) Temporal evolution of VR between observed and calculated waveforms

775 using time window from 1 to 20 min. VR of pressure waveform obtained using

776 conventional (blue) and time-derivative (red) inversions, and time-derivative waveform

777 obtained using conventional (light blue) and time-derivative (red) inversions. (b)

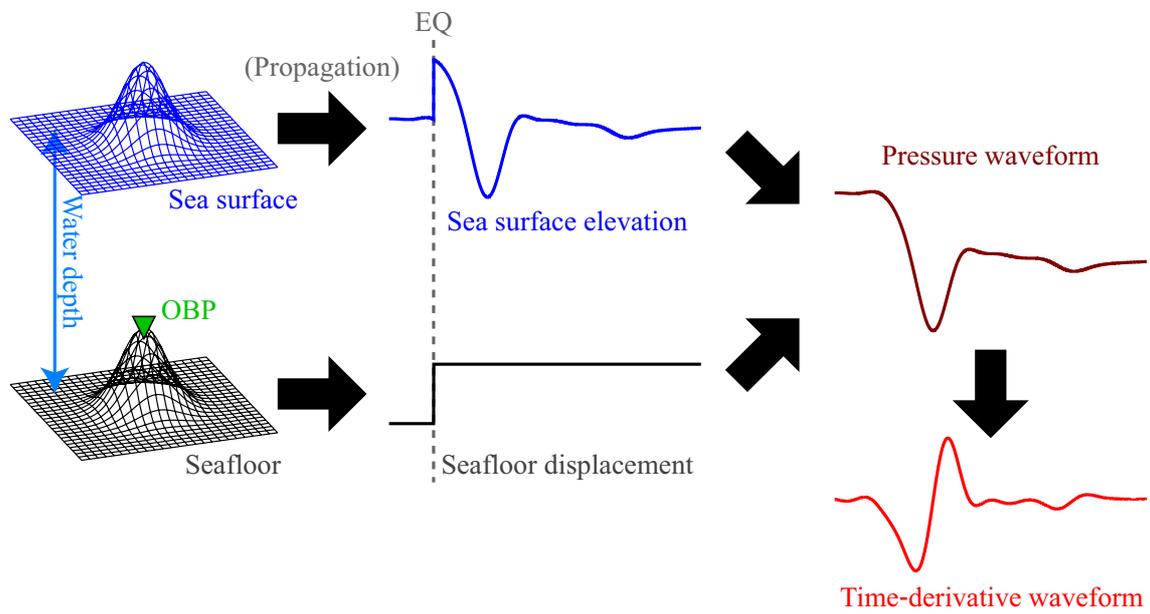
778 Temporal evolution of volumes of displaced seawater. Temporal evolution of total

779 displacement (blue and red), uplifted region (light blue and orange) and subsided region

780 (purple and pink) are shown.

781

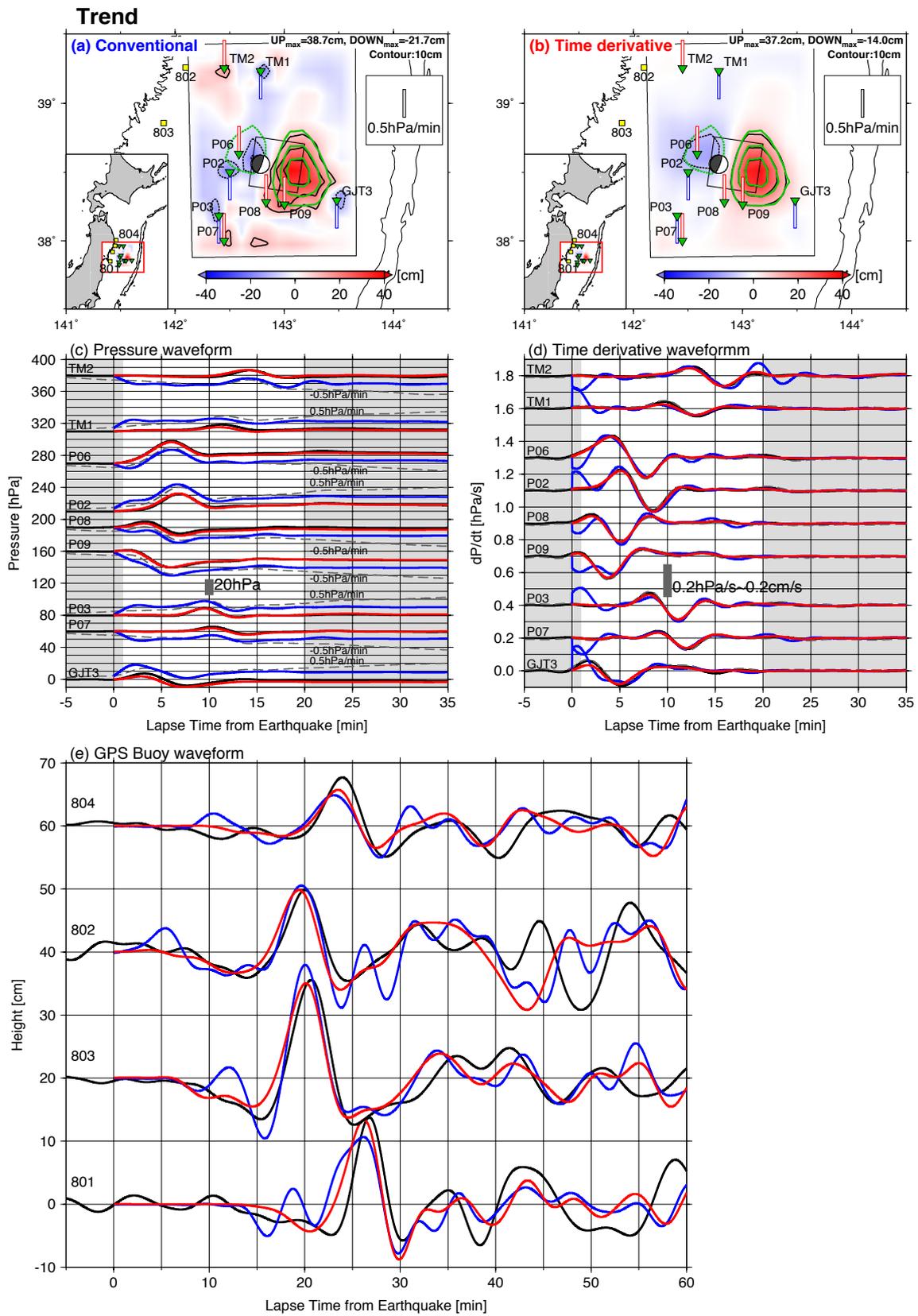
782 **Supplementary materials**



783

784 **Figure S1.** Schematic illustration used to calculate Green's function for conventional and

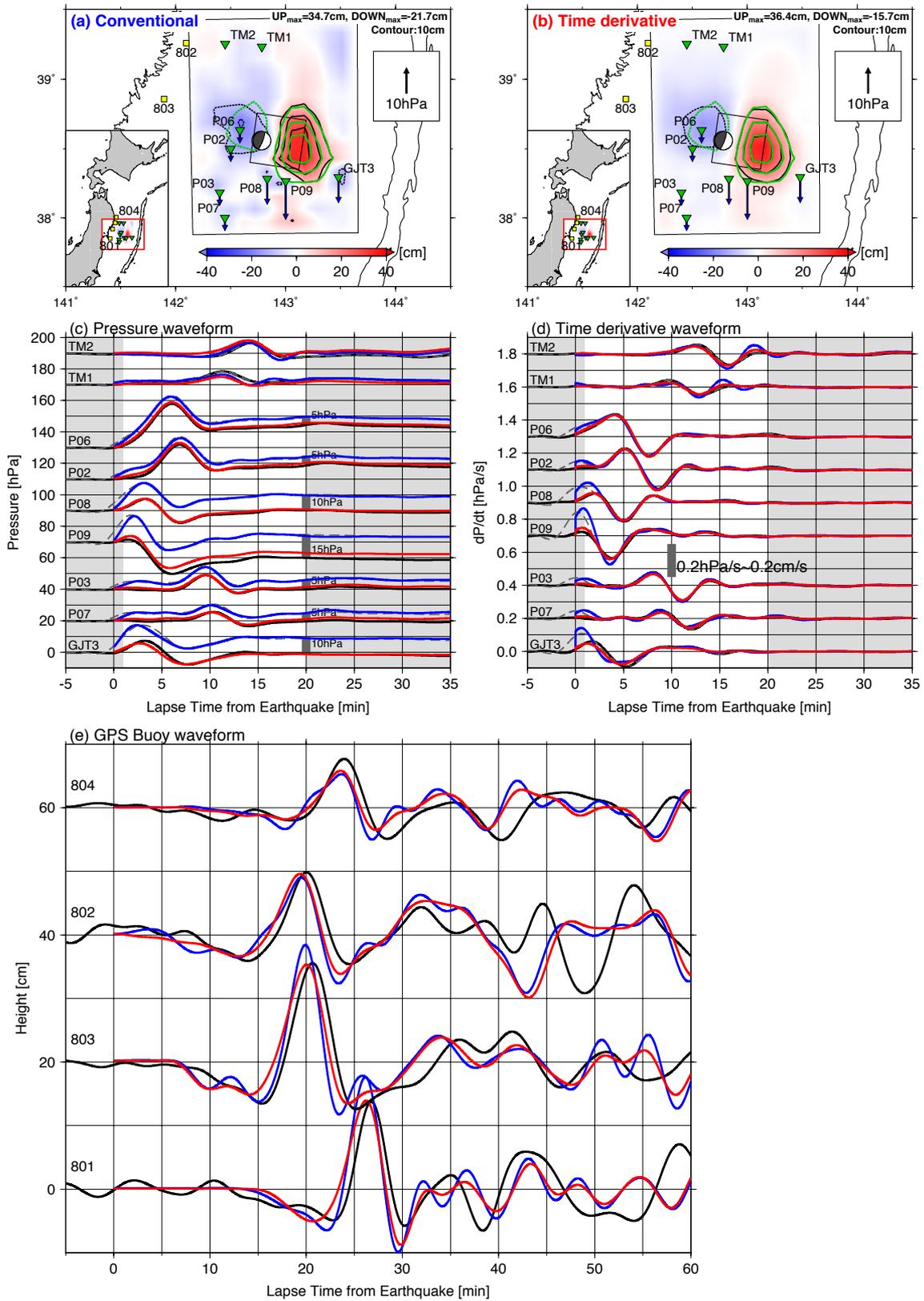
785 the time-derivative inversions.



786

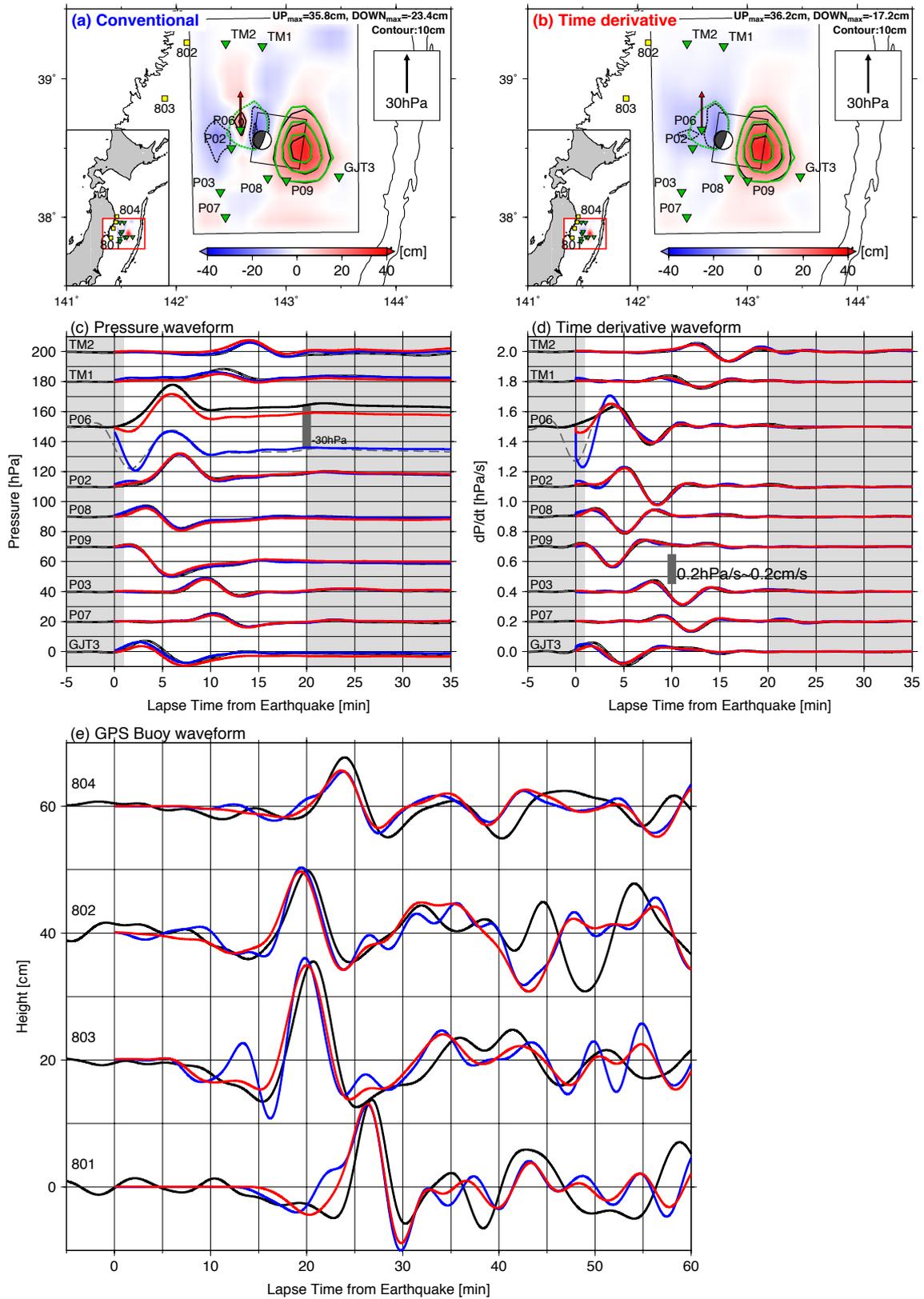
787 **Figure S2.** Inversion results of the 2011 Off-Miyagi earthquake using synthetic OBP data

788 containing artificial long-term trend; other explanations are same as those in Fig. 6.



789

790 **Figure S3.** Inversion results for the 2011 Off-Miyagi earthquake using synthetic OBP  
791 data containing small (5–10 hPa) artificial pressure offset changes; other explanations are  
792 same as those in Fig. 6.



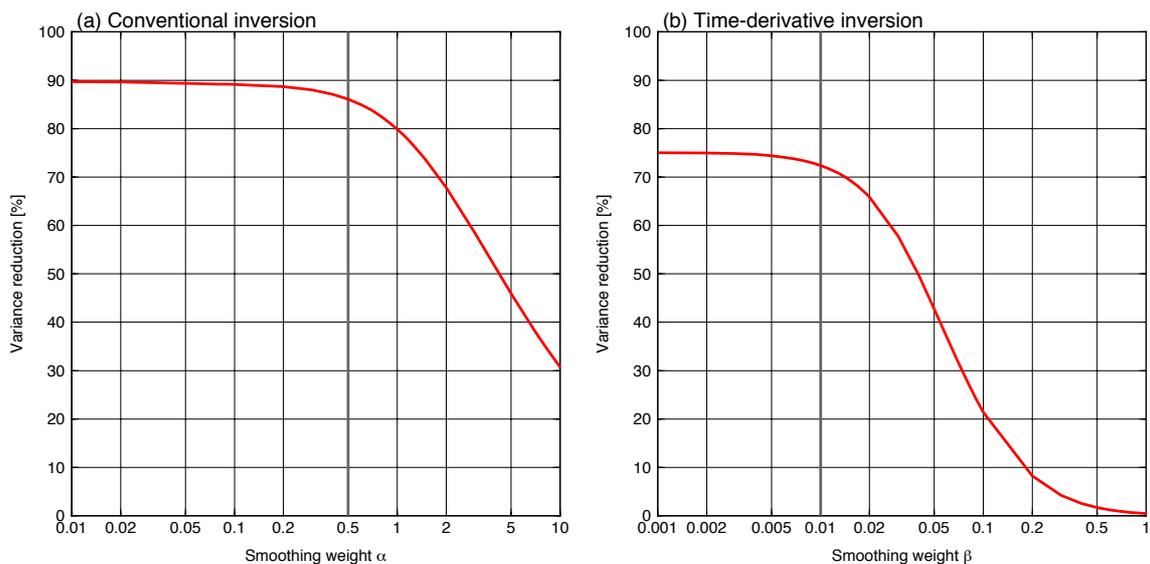
793

794 **Figure S4.** Inversion results for the 2011 Off-Miyagi earthquake using synthetic OBP

795 data containing large (30 hPa) artificial pressure offset change at station P06; other

796 explanations are same as those in Fig. 6.

797

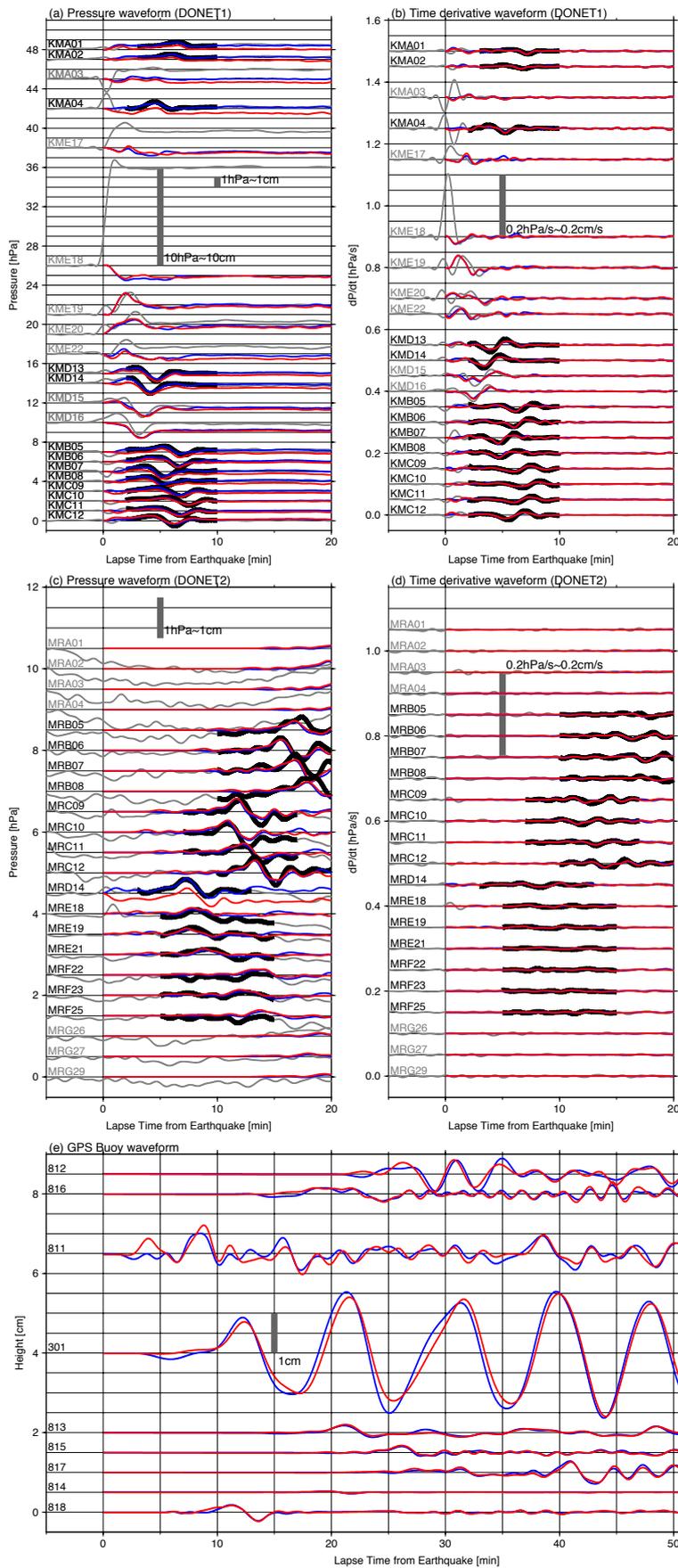


798

799 **Figure S5.** Trade-off curve between smoothing weight and VR, for (a) conventional

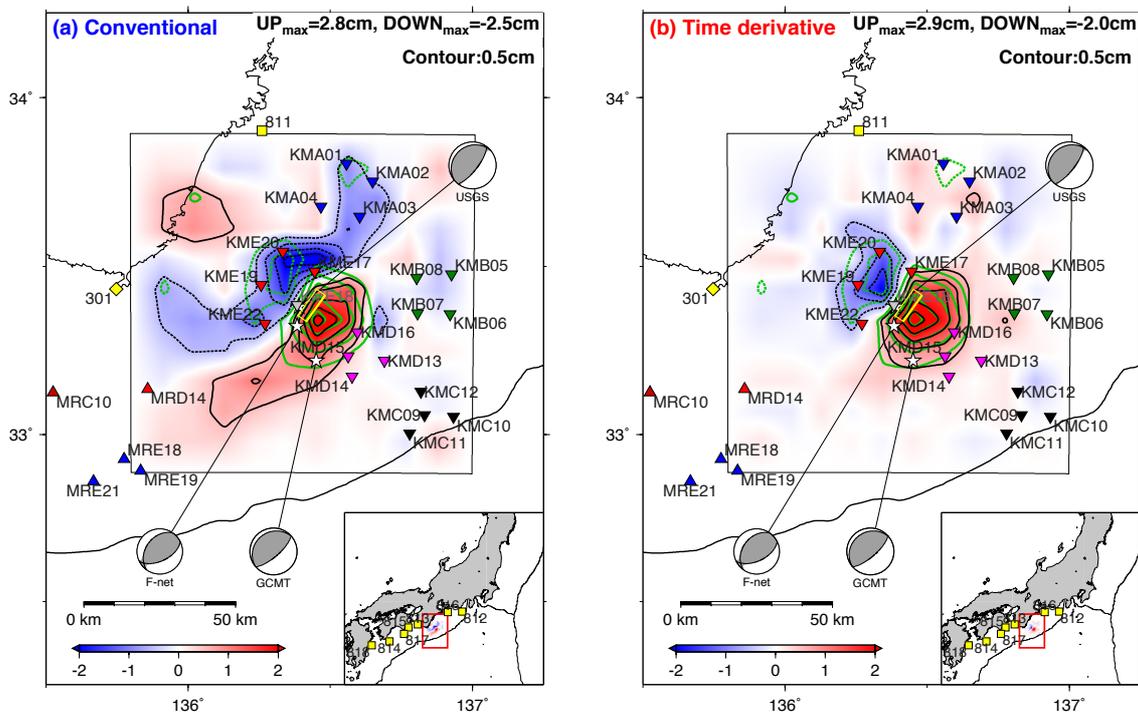
800 inversion and (b) time-derivative inversion when analysing the Off-Mie earthquake. Grey

801 lines denote weight of smoothing constraint adopted in this study.



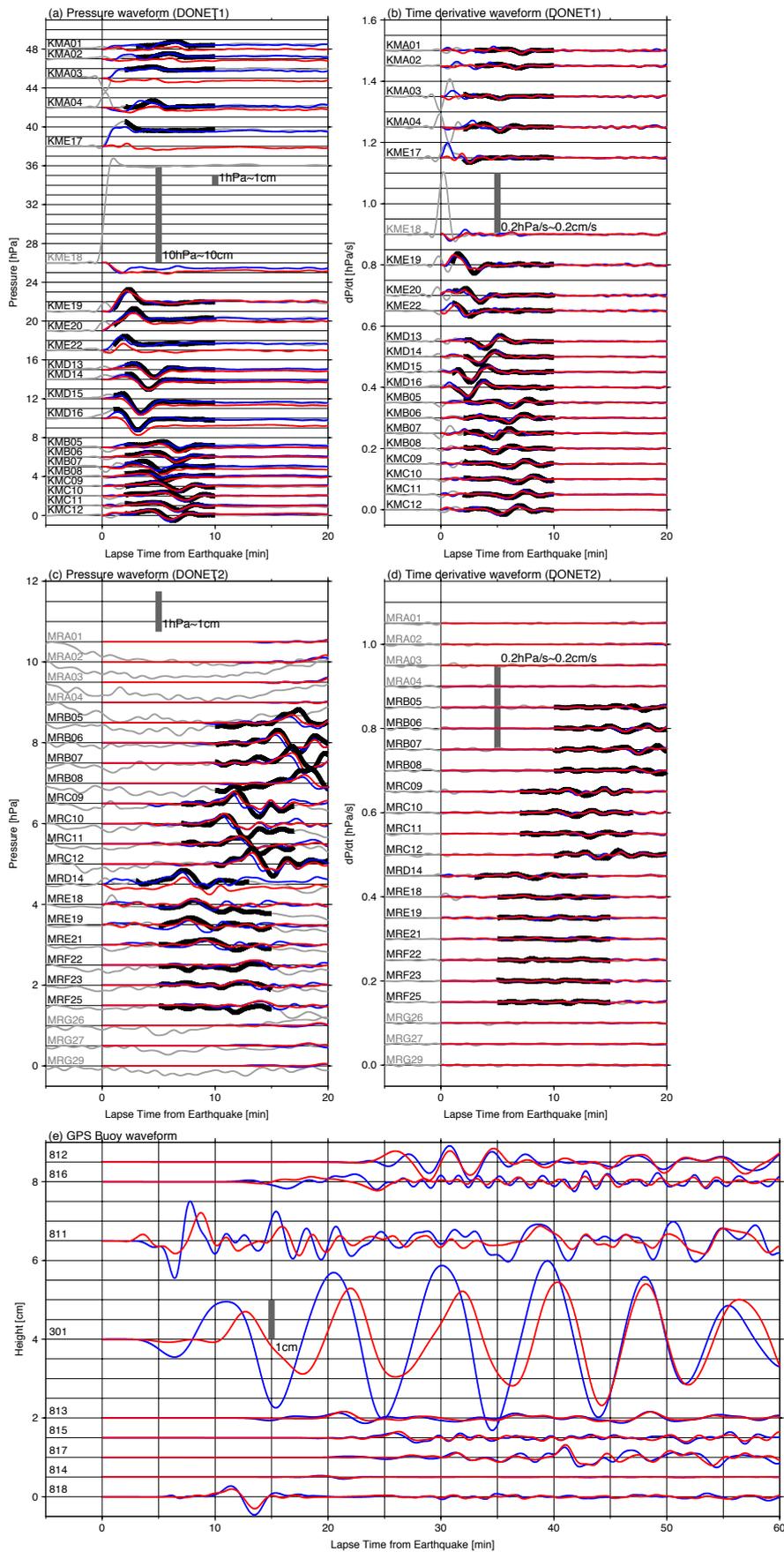
803 **Figure S6.** Comparisons of (a) pressure waveforms and (b) time-derivative waveforms  
 804 at DONET1 OBPs, and (c) pressure waveforms and (d) time-derivative waveforms at  
 805 DONET2 OBPs for the 2016 Off-Mie earthquake between observed tsunami waveforms  
 806 (black) and calculated waveforms calculated from tsunami source model with  
 807 conventional inversion (blue) and time-derivative inversion (red) with OBP data apart  
 808 from epicentre (Fig. 7). Observed waveforms drawn by thick black lines denote the time  
 809 windows used in the inversion analysis. (e) Comparison of waveforms at coastal GPS  
 810 buoys.

811

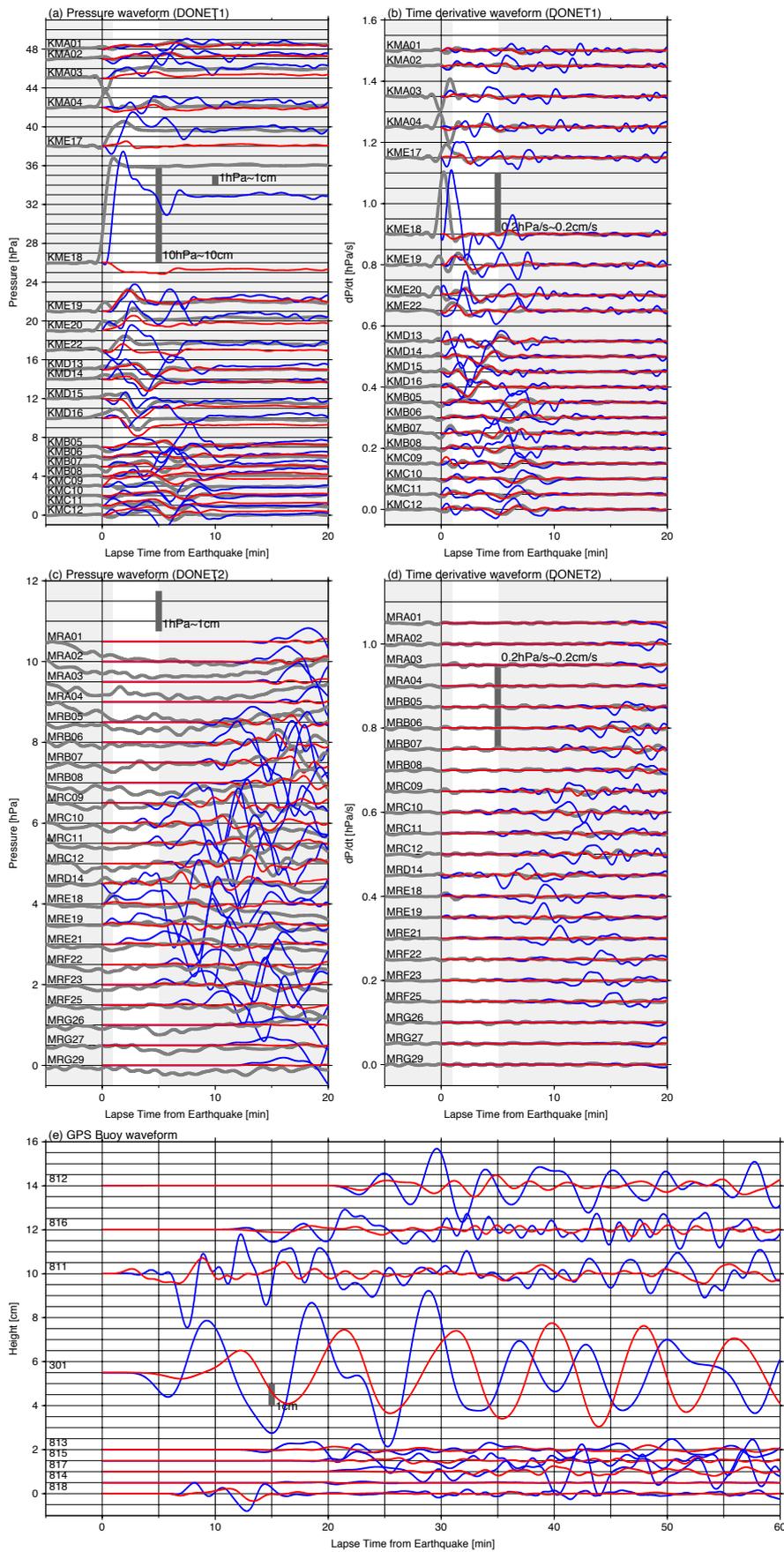


812

813 **Figure S7.** Tsunami source model of the 2016 Off-Mie earthquake obtained using (a)  
814 conventional and (b) time-derivative inversions without OBP data at KME18. Green  
815 contours denote tsunami source distribution obtained from analysis using OBP data apart  
816 from epicentre shown in Fig. 7a. Other explanations are the same as those in Fig. 7.  
817

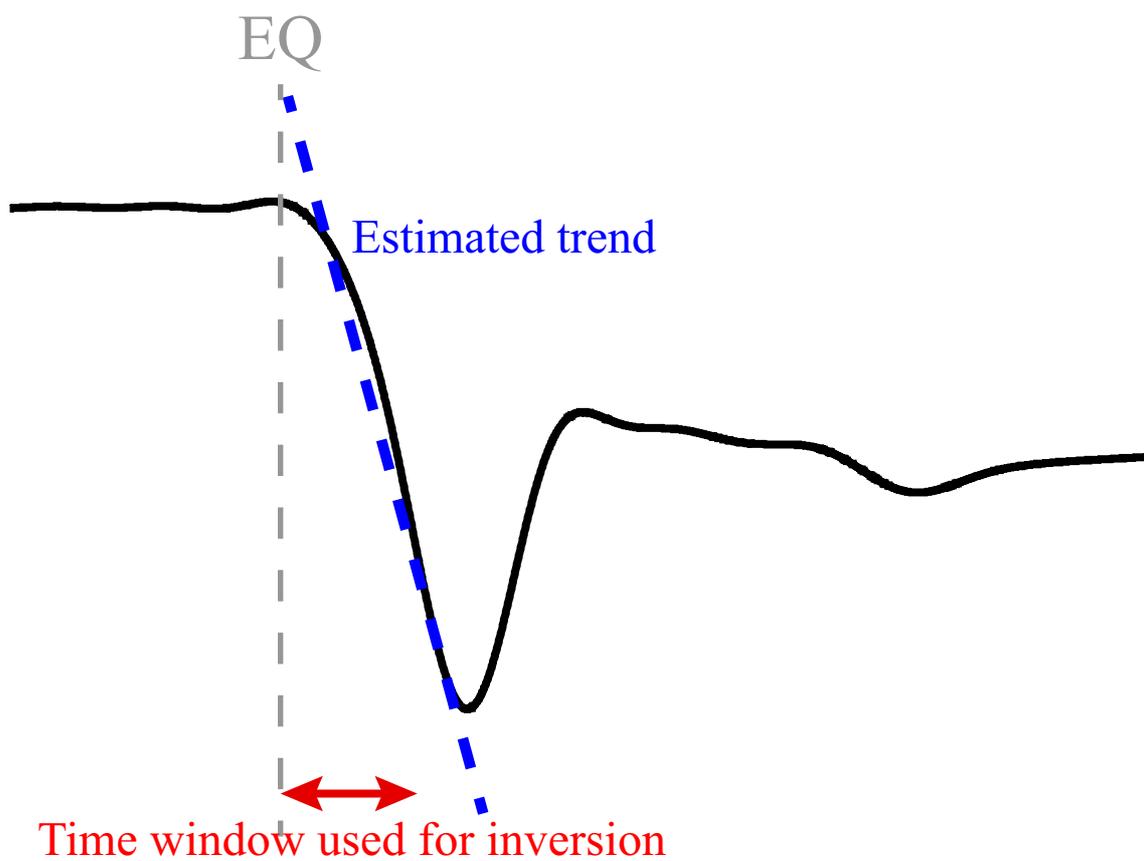


819 **Figure S8.** Comparisons of waveforms for the 2016 Off-Mie earthquake between  
820 observed tsunami waveforms (black) and calculated waveforms calculated from tsunami  
821 source model with conventional inversion (blue) and time-derivative inversion (red),  
822 obtained from OBP data except for KME18 (Fig. S7); other explanations are the same as  
823 those in Fig. S6.



825 **Figure S9.** Comparisons waveforms for the 2016 Off-Mie earthquake between observed  
826 tsunami waveforms (black) and calculated waveforms calculated from tsunami source  
827 model obtained using all OBP data with time window from 1 to 5 min (Fig. 8). White  
828 background area denotes time window used for inversion. Note that scale of vertical axis  
829 is different from that in Figs. S6e and S8e. Other explanations are the same as those in  
830 Fig. S6.

831



832

833 **Figure S10.** Schematic illustration of possible trade-off situation in simultaneous

834 estimation of tsunami and linear trend; black line denotes observed tsunami waveform. It  
835 is possible to estimate the tsunami as a linear trend (blue dashed line), if tsunami data  
836 with a short time window are employed (denoted by red arrow).