

1 **Towards tsunami early-warning with Distributed**
2 **Acoustic Sensing: Expected seafloor strains induced by**
3 **tsunamis**

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12 **Key Points:**

- 13 • A review of coupling mechanisms between long-period surface gravity signals (tsunami
14 waves) and seafloor cables.
- 15 • Seafloor compliance and the cable's Poisson effect are anticipated to primarily in-
16 fluence DAS responses to tsunami waves.
- 17 • Numerical simulations indicate the feasibility of tsunami early-warning via DAS
18 after significant seismic signal filtration.

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Abstract

Tsunami wave observations far from the coast remain challenging due to the logistics and cost of deploying and operating offshore instrumentation on a long-term basis with sufficient spatial coverage and density. Distributed Acoustic Sensing (DAS) on submarine fiber optic cables now enables real-time seafloor strain observations over distances exceeding 100 km at a relatively low cost. Here, we evaluate the potential contribution of DAS to tsunami warning by assessing theoretically the sensitivity required by a DAS instrument to record tsunami waves. Our analysis includes signals due to two effects induced by the hydrostatic pressure perturbations arising from tsunami waves: the Poisson's effect of the submarine cable and the compliance effect of the seafloor. It also includes the effect of seafloor shear stresses and temperature transients induced by the horizontal fluid flow associated with tsunami waves. The analysis is supported by fully coupled 3-D physics-based simulations of earthquake rupture, seismo-acoustic waves and tsunami wave propagation. The strains from seismo-acoustic waves and static deformation near the earthquake source are orders of magnitude larger than the tsunami strain signal. We illustrate a data processing procedure to discern the tsunami signal. With enhanced low-frequency sensitivity on DAS interrogators (strain sensitivity $\approx 2 \times 10^{-10}$ at mHz frequencies), we find that, on seafloor cables located above or near the earthquake source area, tsunamis are expected to be observable with a sufficient signal-to-noise ratio within a few minutes of the earthquake onset. These encouraging results pave the way towards faster tsunami warning enabled by seafloor DAS.

Plain Language Summary

Detecting tsunami waves early is crucial to warn people in coastal areas to move to safety. However, monitoring tsunamis far from the coast remain challenging an expensive because it requires a lot of equipment spread out over large areas of the ocean. This research looks into a new method called Distributed Acoustic Sensing (DAS) to detect tsunamis. DAS uses the existing network of undersea fiber optic cables-cables that usually carry internet and telephone signals across oceans. By making slight adjustments to these cables, it is possible to use them to pick up strains in the cable caused by the passing of a tsunami wave. The study shows that by understanding and measuring these strains, DAS can potentially spot tsunamis soon after they start. This method is promising because it uses existing cables, making it a more affordable option for widespread tsunami detection.

1 Introduction

Although several Tsunami Early Warning Systems (TEWS) are in operation worldwide, this is yet to be the norm, mainly due to the high cost associated with the installation and operation of offshore instrumentation with sufficient spatial coverage, density, and real-time data availability. To circumvent these challenges, and as the majority of recorded tsunamis worldwide are directly ascribed to earthquakes (Reid & Mooney, 2023), many TEWS rely on seismic data for source characterization. However, the energy released by the earthquake often is not a sufficient predictor of the tsunami intensity, which may lead to inaccurate early warnings and false alarms (V. Titov et al., 2016; Katsumata et al., 2021). For this reason, the basic sensor package for earthquake and tsunami early warning is a seismometer/accelerometer to detect ground shaking and a pressure gauge to detect tsunami waves (Wilcock et al., 2016). An effective system should use local wave measurements to characterize the tsunami and to allow the detection of tsunamis from other sources such as submarine landslides.

To obtain reliable observations leading to faster tsunami confirmations, it is best to deploy instrumentation directly in the source region, including active subduction zones and areas prone to submarine landslides in volcanic systems. Such observation systems

69 are crucial for protecting the population and improving our understanding of tsunami
70 generation, which is less understood than the propagation process mainly due to scarce
71 offshore and in-situ observations. The detection of tsunamis in the open ocean, achieved
72 by monitoring variations in water pressure, is significantly challenged by the relatively
73 modest amplitudes of tsunamis — typically around a meter, even for the largest events
74 — and by the logistical complexities and expenses involved in maintaining instruments
75 in remote oceanic locations on a long-term/permanent basis. Additionally, ensuring dense
76 and extensive spatial coverage across all potential tsunami sources complicates this en-
77 deavor, making it difficult to rapidly and robustly estimate a tsunami’s potential impact.
78 Furthermore, little is known about spatial variations of tsunamis, because too few dense
79 array measurements of tsunamis are available so far (Kohler et al., 2020). As the pop-
80 ulation grows in coastal regions, the recurrence of tsunami tragedies underscores the ur-
81 gent need for better detection methods and early warning systems.

82 Direct tsunami observations are made mostly by coastal tide gauges and fixed moor-
83 ings or buoys located offshore, such as the Deep-ocean Assessment and Reporting of Tsunamis
84 (DART) system. The current network of 77 DART stations worldwide has a sparse sensor
85 density and a limited spatial footprint. Each DART station consists of a transmit-
86 ter surface buoy and an anchored seafloor pressure sensor (National Oceanic and Atmo-
87 spheric Administration (NOAA) Pacific Marine Environmental Laboratory (PMEL) Na-
88 tional Center for Tsunami Research (NCTR), 2023). With an approximate cost of US\$
89 0.5M per station and high maintenance and repair costs, compounded by the remote lo-
90 cation of the buoys and required ship operations, densifying DART instrumentation re-
91 quires a financial commitment which can be prohibitive for developing countries (Bernard
92 & Titov, 2015). Moreover, the DART system is primarily designed for providing fore-
93 casts in the case of transoceanic or far-field tsunamis, but not for regional and local tsunamis.
94 A confirmation of tsunami and an evaluation of its amplitude are obtained once the record-
95 ings from the closest station are analyzed, which may often take two to three hours (Mungov
96 et al., 2013). Tide gauges are typically located inside harbors and bays, and hence can
97 only have a limited contribution to early warning. Furthermore, the tsunami signal of
98 tide gauge records are often filtered or distorted by the shallow coastal water depth, which
99 makes it difficult to extract detailed information about a tsunami (Saito, 2019).

100 In rare cases, direct tsunami observations are made by cabled observatories: the
101 North-East Pacific Time-series Undersea Networked Experiments system (NEPTUNE)
102 in Canada (Heidarzadeh & Gusman, 2019), the Dense Oceanfloor Network System for
103 Earthquakes and Tsunamis (DONET) and S-NET systems in Japan. DONET consists
104 of 51 stations over a length of 800 km, whilst S-NET is the world’s largest seafloor ob-
105 servation network, consisting of 150 cable-linked seismic and tsunami sensors covering
106 an area of 1000 km x 300 km. Both are deployed on the slopes of subduction zone trenches
107 and in rupture-prone regions, such as those associated with the historical Tonankai and
108 Nankai earthquakes (Aoi et al., 2020). These observation systems have been used to de-
109 tect several tsunamis of various sizes based on changes in water pressure (Tsushima et
110 al., 2012; V. V. Titov et al., 2005), including the disastrous tsunami of the 2011 Tohoku-
111 oki earthquake (Saito et al., 2011; Maeda et al., 2011). Although capable of carrying ro-
112 bust, long-term observations, cabled observatories require a substantial financial invest-
113 ment, especially to achieve a wide and complete spatial coverage on the scale of subduc-
114 tion zones.

115 Distributed Acoustic Sensing (DAS) is an emerging technology that re-purposes
116 existing fiber optics in telecommunication cables into long and dense arrays of longitu-
117 dinal strain sensors (Zhan, 2020). DAS can record external inputs that deform fiber op-
118 tic cables in a broad frequency range (Paitz et al., 2020), with a current maximum dis-
119 tance range of around 150 km (Waagaard et al., 2021). These capabilities have motivated
120 a decade of applications in subsurface imaging and microseismicity monitoring for en-
121 ergy production and carbon sequestration. DAS arrays have recorded microearthquakes,

regional earthquakes, teleseisms, and infrastructure signals. Analysis of these wavefields is enabling earthquake seismology where traditional sensors were sparse, as well as structural and near-surface seismology. These studies have improved our understanding of DAS instrument response through comparison with traditional seismometers. More recently, DAS has been used to study cryosphere systems, marine geophysics, geodesy, and volcanology (Lindsey & Martin, 2021).

Several key requirements for TEWS are inherent attributes of DAS: low data latency, high spatial density, and relatively low cost, often dominated by the DAS interrogator equipment cost (\sim US\$ 150k). Currently commercially available DAS interrogators sense strain with meter-scale spatial resolution over cable spans of up to 150 km and deliver data on land in real-time (Williams et al., 2023; Fernandez-Ruiz et al., 2022; Sladen et al., 2019). Coupling these attributes with the existing (and still expanding) global coverage of the transoceanic telecommunication cable network (Brenne et al., 2024), positions DAS as a sensor package to consider for TEWS. It provides a cost-effective means of deploying instrumentation to monitor offshore locations such as subduction zones, and can also provide several measurements per tsunami wavelength to better study the characteristics of tsunamis.

The earliest reports on seafloor DAS for Earth science documented signals induced by the swell (Sladen et al., 2019; Williams et al., 2019; Lindsey et al., 2019). In particular, (Sladen et al., 2019) showed that the depth-dependence of the signal amplitude is consistent with the pressure depth-dependence from the linear theory of surface gravity waves. This demonstrated that DAS can record surface gravity waves, although at much shorter periods than tsunami waves. The first, and so far only reported detection of a tsunami with seafloor DAS (Xiao et al., 2024) made use of 60 km of cable off the northwestern US coast to detect a far-field tsunami, which produced a small tsunami wave height of \sim 6 mm at a DART station, and reaching \sim 1 cm by the time it reached the end of the monitored fiber. Although the signal-to-noise ratio of the detected tsunami phase was low, even after beamforming, these findings underscore the feasibility of DAS technology to detect tsunami waves.

Several aspects of DAS still need to be addressed to ascertain tsunami wave detection with sufficient signal-to-noise ratio to be integrated into TEWS. DAS is most sensitive to strain along the cable axis. Given that cables are typically laid or buried along the seafloor, they can primarily perform as horizontal arrays of sensors recording horizontal strain, which is not a quantity that has been considered before in tsunami studies. The question of sensitivity at low frequencies also merits attention given that DAS instrumentation usually has lower performance at long periods typical of a tsunami (>100 s) (R. Fernández-Ruiz et al., 2019).

The remainder of the paper is organized as follows. In section 2, we review recently reported developments in the range and sensitivity of DAS at low frequencies, intended to enable improved accuracy over long distances in the mHz frequency range of interest in tsunami studies. They are attained by suppressing the $1/f$ instrument noise, specifically on the chirped-pulse (CP)-DAS architecture that employs linearly chirped laser pulses. Then, to determine the potential contribution of DAS instrumentation in the design of future TEWS, we present a theoretical analysis of the expected seafloor strains potentially induced by tsunami waves. In section 3, we consider two effects arising from the hydrostatic pressure due to the changing sea-surface level: transfer from radial to longitudinal deformation of the cable via Poisson’s effect and the deformation of the seafloor via the compliance effect. In sections 4 and 6, we consider the shear strain and temperature changes at the seafloor induced by the lateral water flow associated with tsunami waves. With the aforementioned hydrostatic and shear sources of strain, section 5 simulates the expected strain signal amplitude from a propagating tsunami wave. In section 7, we validate the proposed model by analyzing a 3-D fully coupled simulation of earthquake rupture, seismic, acoustic and tsunami wave propagation.

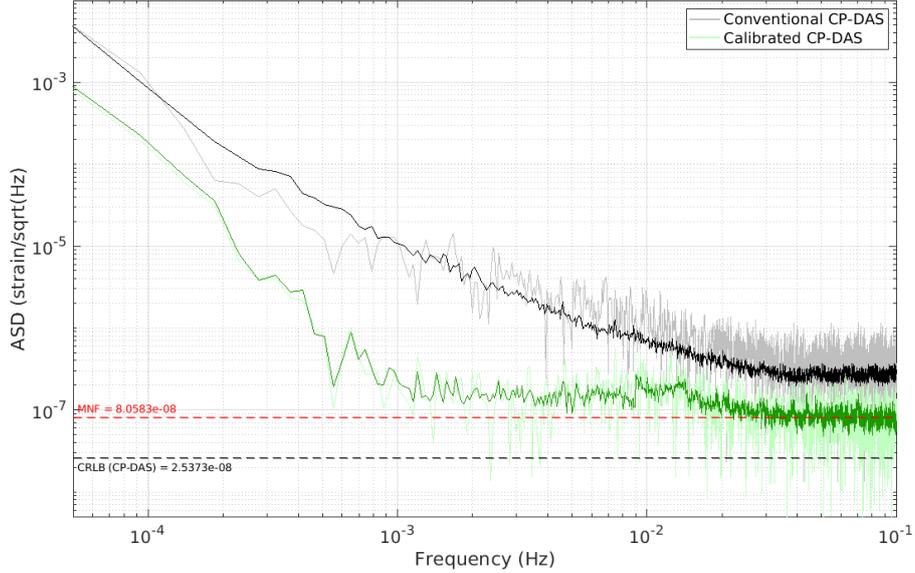


Figure 1. Median Amplitude Spectral Density (ASD) of the noise floor of a conventional chirped-pulse DAS (bold black line) and a DAS unit implementing the multi-frequency database calibration method (bold green line). Light-colored lines are the ASD of an example fiber point ($z = 4011$ m). The frequency of acquisition was 0.5 Hz. Both conventional and calibrated noise curves only differ in the processing, as it is the same physical measurement, using the same optical traces. The red dashed line is the median noise floor of the bold green line. The black dash line delineates the theoretical noise floor, the Cramer-Rao Lower Bound (CRLB) for this instrument configuration (i.e. frequency of acquisition, chirp bandwidth, signal-to-noise ratio of the optical trace).

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2 Sensitivity of Standard and Advanced DAS Instrumentation at Tsunami Frequencies

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DAS instrumentation exploits backscattered light from a laser pulse propagating along the optical fiber, due to Rayleigh scattering by inherent heterogeneities along the fiber. Deformation and temperature perturbations to the fiber due to environmental changes cause phase changes in the backscattered light, which are detected via optical interferometry by comparing a measured time-domain trace with a previous fiber reference state. The optical phase recovery methods employed can be complex and add stringent performance requirements on the optical hardware. An advantageous approach is to employ linearly chirped pulses. This allows a frequency-to-time mapping in the backscattered light intensity time-series, which enables direct detection of phase shifts without the need for frequency sweeping or phase unwrapping (Pastor-Graells et al., 2016, 2017). These raw optical measurements are integrated along discrete cable segments, the “gauge length”, and then related to position-resolved strain or strain-rate at each “DAS channel”.

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One of the main limitations of current DAS systems towards applications in tsunami monitoring is their limited sensitivity at low frequencies below 0.01 Hz. Relying on a reference measurement to evaluate the relative phase change of the backscattered light has the drawback that any large strain, temperature fluctuations, or laser frequency noise will cause deviations in the phase shift, potentially rendering the current reference trace

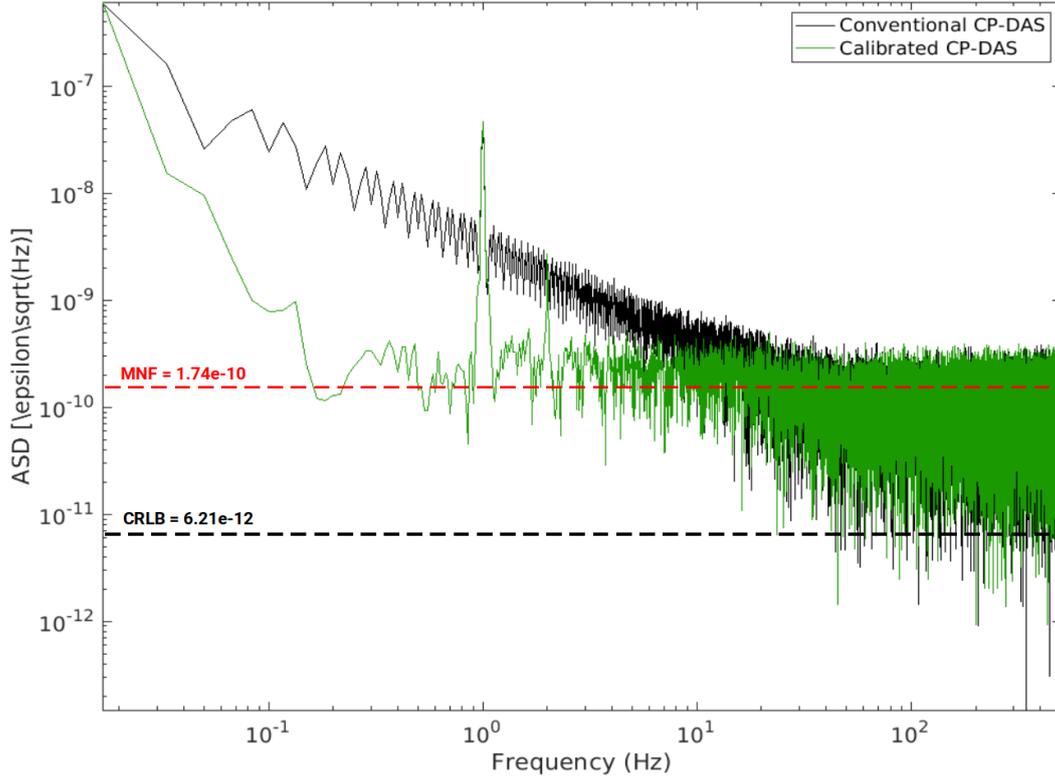


Figure 2. Median Amplitude Spectral Density (ASD) of the noise floor of a conventional chirped-pulse DAS (bold black curve) and a DAS unit implementing the multi-frequency database calibration method (bold green curve). The red dashed line is the median noise floor of the green curve. Further reduction of the noise floor (in comparison to Figure 1), is attained here by improving the stability of the laser and increasing the sampling frequency to 1 kHz. To clearly highlight the benefit of the calibration technique, we introduce a relatively strong 1 Hz modulation of the fiber strain with a piezoelectric transducer (PZT) in order to have regular reference updates in the normal CP-DAS measurement. Due to the high sampling frequency and limitations of our acquisition equipment, we could only resolve frequencies above 15 mHz. During the acquisition, the fiber was in a relatively uncontrolled environment, without temperature drift correction, and we observe an uptake in noise level below ≈ 0.2 Hz. The noise floor of the calibrated CP-DAS is flat down to at least 0.1 Hz. This noise floor is expected to extend to lower frequencies if temperature was controlled during acquisition.

195 invalid. This imposes the need to update such reference, and integrate the short-term
 196 variation measurements to obtain absolute strain or temperature variations (Fernandez-
 197 Ruiz et al., 2018). However, updating references unavoidably incorporates random noise,
 198 which introduces error accumulation over time (noise is integrated with each cumula-
 199 tive variation measurement). This translates into a $1/f$ noise component, which is detri-
 200 mental to the determination of very slow processes in the mHz frequency range or be-
 201 low.

202 The long-term stability of chirped-pulse phase-sensitive optical time-domain reflecto-
 203 metry (CP- Φ OTDR) can be improved with the acquisition of a multi-frequency database
 204 to carry out “calibrated” measurements in DAS along an unperturbed fiber (Vidal-Moreno
 205 et al., 2022). This approach has been found to suppress reference update-induced $1/f$ noise,

206 as the comparison in Figure 1 illustrates. Note that in order to attain sufficient tempo-
 207 ral resolution in the sub-mHz regime, and to circumvent limitations in the digital mem-
 208 ory of the data acquisition equipment employed in our facilities, the results in Figure 1
 209 have been obtained with a very low acquisition frequency of 0.5 Hz. The DAS noise floor
 210 can be decreased further by employing a laser source with an improved wavelength stabi-
 211 lity, and by increasing the frequency of acquisition. For fiber lengths up to 100 km, the
 212 acquisition frequency can be raised to 1 kHz, resulting in a mean noise floor of $\epsilon \approx 2 \times$
 213 10^{-10} as exemplified in Figure 2. Due to the relatively high acquisition trigger frequency,
 214 and the aforementioned limitations in the data acquisition equipment, we are only able
 215 to probe the fiber for about 1 minute, allowing to resolve frequencies above 15 mHz. Fur-
 216 thermore, the steady climb in noise levels at the low frequencies in both Figures 1 and
 217 2 is due to the relatively uncontrolled environment; no temperature control was imple-
 218 mented to mitigate in-lab temperature fluctuations, hence this is most likely tempera-
 219 ture drift. Despite these limitations, these acquisitions show a considerable improvement
 220 in canceling instrument-generated $1/f$ noise. Without temperature drifts, the noise floor
 221 limit obtained here at high frequencies likely extends to lower frequencies, and is per-
 222 haps limited by the temperature fluctuations inherent to ocean-bottom environments.

223 Unlike phase-based DAS (Φ OTDR) which relies on retrieving the phase of backscat-
 224 tered light limited to values between $-\pi$ and π (Masoudi et al., 2013; Diaz-Meza et al.,
 225 2023); with the chirped-pulse architecture (CP- Φ OTDR), the measurable dynamic range
 226 is limited in the maximum strain variation between pulse and reference, $\Delta\epsilon_{max}$, which
 227 induces a temporal shift, such that it renders the reference measurement inaccurate (Bhatta
 228 et al., 2019). This is a function of the spectral characteristics of the chirped pulse (R. Fernandez-
 229 Ruiz et al., 2019). With the configuration illustrated in Figure 2, for the conventional
 230 chirped-pulse DAS (5 GHz chirp content at 1 kHz acquisition rate) this corresponds to
 231 $\Delta\epsilon_{max} \sim \pm 10^{-6}$ before the instrument resorts to updating the reference (and accumu-
 232 lating $1/f$ noise). For the calibrated version of the instrument this limit becomes $\pm 82.5 \times$
 233 10^{-6} (Vidal-Moreno et al., 2022).

234 In the following sections, we adopt as reference DAS sensitivity at all relevant fre-
 235 quencies the improved mean noise floor shown in Figure 2, $\epsilon \approx 2 \times 10^{-10}$.

236 3 Strain of Seafloor Optical Fibers from Water Pressure Loading

237 Here, we estimate the expected amplitude of seafloor horizontal strains that could
 238 be generated by the hydrostatic pressure variations induced by the changes in water height
 239 via two effects: the compliance effect of the seafloor and the Poisson’s effect of the cab-
 240 ble.

241 3.1 Compliance Effect

242 The water pressure perturbation induced by a tsunami wave acts on the seafloor
 243 and deforms it elastically. The problem of determining the strain of an elastic half-space
 244 caused by a normal force on its surface was first considered by J. V. Boussinesq. In this
 245 formulation, the solid Earth is approximated by a homogeneous, isotropic, linear elas-
 246 tic half-space. Neglecting the spherical geometry of Earth is adequate because we con-
 247 sider tsunami wavelengths that are short compared to Earth’s radius. Details of the so-
 248 lution are given by (Steketee, 1958), from which we take the expression for horizontal
 249 displacement parallel to the cable axis (\bar{u}_x) in the wavenumber domain (spatial Fourier
 250 transform):

$$\bar{u}_x = ik_x \frac{\bar{P}}{2\mu k^2} \left(1 - \frac{1}{\alpha} + kz \right) e^{-kz} \quad (1)$$

251 where z is the vertical coordinate, defined as pointing upwards,

$$\alpha = \frac{\lambda + \mu}{\lambda + 2\mu}, \quad (2)$$

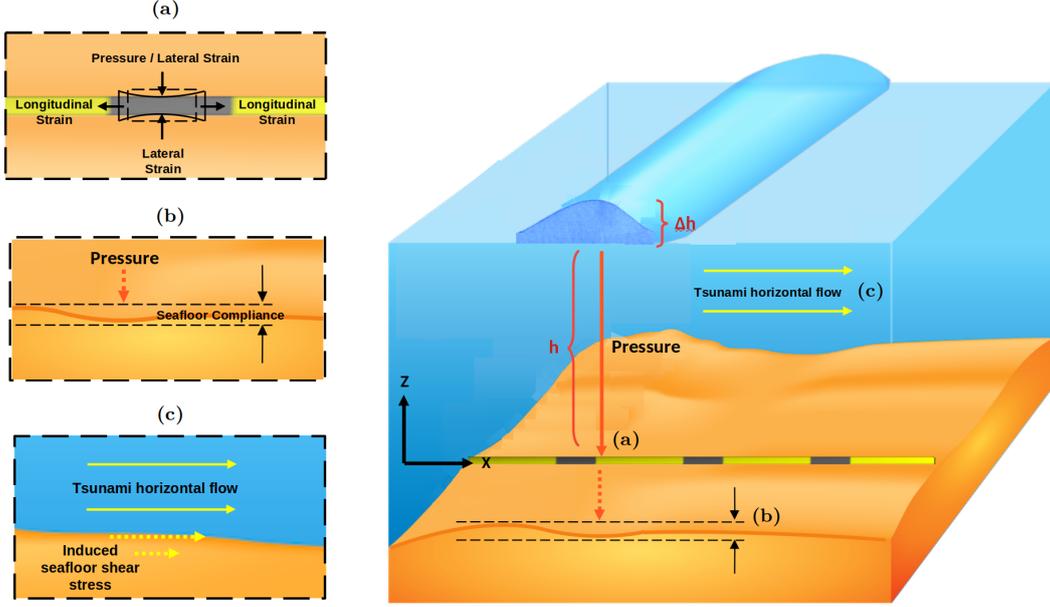


Figure 3. Illustration of a submarine cable in the presence of a propagating tsunami wave. Vertical pressure loading due to changes in the water column, Δh give rise to (a) longitudinal elongation of the cable due to the Poisson effect and (b) deformation of the seafloor (compliance). The horizontal flow motion beneath the tsunami wave induces (c) shear stress on the seafloor surface. Also shown is the coordinate system convention and definition of relevant variables.

252 λ and μ are Lamé's elastic moduli, and $k = \sqrt{k_x^2 + k_y^2}$ is the amplitude of the wavenum-
 253 ber vector (k_x, k_y) along the horizontal plane. The horizontal displacement at the seafloor
 254 ($z = 0$) is

$$\bar{u}_x = ik_x \frac{\bar{P}}{2\mu k^2} \left(1 - \frac{1}{\alpha}\right) = -ik_x \frac{\bar{P}}{2k^2(\lambda + \mu)} \quad (3)$$

255 The strain along the cable axis (x -axis), which is the quantity measured by DAS, is $\epsilon_x =$
 256 $\frac{du_x}{dx}$. Since its Fourier transform is $\bar{\epsilon}_x = ik_x \bar{u}_x$, we obtain:

$$\bar{\epsilon}_x = \frac{k_x^2}{k^2} \frac{\bar{P}}{2(\lambda + \mu)} \quad (4)$$

257 Defining θ as the angle between the wavenumber vector (the tsunami wave propagation
 258 direction) and the cable axis, we have $k_x/k = \cos \theta$ and $\bar{\epsilon}_x = \frac{\bar{P} \cos^2 \theta}{2(\lambda + \mu)}$. Taking the in-
 259 verse Fourier transform of $\bar{\epsilon}_x$, the amplitude of seafloor strain oscillation induced by a
 260 single tsunami plane wave is

$$\epsilon_x = \frac{P \cos^2 \theta}{2(\lambda + \mu)} \quad (5)$$

261 We obtain a generic estimate of the amplitude of this effect by taking the average
 262 over all possible tsunami wave directions:

$$\epsilon_{xavg} = \frac{P}{4(\lambda + \mu)} \quad (6)$$

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3.2 Poisson's Effect

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Another potential contribution to the strain signal detected by DAS is from Poisson's effect: water pressure perturbations cause radial compression of the submarine cable, which in turn induces longitudinal elongation of the cable through Poisson's effect. We assume that water pressure acts isotropically on the whole circumference of the cable, neglecting the fact that the cable might be partially buried. Representing the cable as an effective homogeneous medium with effective Young's modulus E and Poisson's ratio ν , its longitudinal strain due to Poisson's effect is given by (Tatekura et al., 1982):

$$\epsilon_x = 2 \frac{1 - 2\nu}{E} P \quad (7)$$

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The cable's sensitivity to pressure is highly dependent on the cable design. For an optical submarine cable assembly, $E \sim 5 - 50$ GPa and $\nu \sim 0.2 - 0.25$ (Kojima et al., 1982; Tatekura et al., 1982). The lower values of E and larger values of ν correspond to less armored cables. For a submarine cable with mid-range elastic properties ($\nu = 0.23$ and $E = 25$ GPa) the Poisson's effect sensitivity is $\epsilon_x \sim 4.3 \times 10^{-11} \Delta P$, which falls within expectations based on previously reported theoretical estimates (Budiansky et al., 1979; Haavik, 2022).

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3.3 Expected Strain Signal from Hydrostatic Pressure

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Here we combine the two pressure effects and estimate their total expected contribution to seafloor DAS strain as a function of tsunami wave height and frequency. We stay within the linear regime of the tsunami theory. This framework assumes that the wave amplitude (typically less than 10 meters) is much smaller than the wavelength. This assumption is generally violated only during the final stage of wave breaking near the coast, or under extreme tsunami generating conditions. We neglect the feedback of seafloor compliance on tsunami waves, which is significant at much longer periods and propagation distances than considered here (Tsai et al., 2013); we thus adopt the conventional tsunami theory with a rigid seafloor. We assume an incompressible, homogeneous, and non-viscous ocean, subject to a constant gravitational field, with uniform water depth h .

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Within this framework, the frequency ω and wavenumber k of a sea surface gravity wave satisfy the dispersion relation (Watada, 2013)

$$\omega^2 = gk \tanh(kh) \quad (8)$$

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where g is the gravitational acceleration. For a given frequency within the range of interest in tsunami studies, and for sea depths h ranging from 500 meters to 6 km, we solve the above relation for the corresponding kh . The amplitude of the pressure induced by a sea surface height anomaly (SSHA) of amplitude Δh , at a vertical position z relative to the seafloor, is

$$P(z) = \rho g \Delta h \frac{\cosh(kz)}{\cosh(kh)} \quad (9)$$

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The seafloor pressure amplitude (at $z = 0$) is thus

$$P = \frac{\rho g \Delta h}{\cosh(kh)} \quad (10)$$

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The cosh term induces a low-pass filter effect: only long wavelengths generate significant seafloor pressure changes. As a validation of this linear theory of surface gravity waves with DAS data, (Sladen et al., 2019) showed that the depth-dependence in equation 10 is consistent with the decay with depth of the amplitude of DAS signals from swell.

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We calculate seafloor pressure for values of Δh (SSHA) of 0.1 and 1 meter representative of intermediate and large-size tsunamis in the open ocean. We use the resulting pressure to calculate the longitudinal cable strain due to the compliance effect and

305 Poisson's effect through Equations (5) and (7), respectively. We add the two effects to
 306 obtain the total DAS strain due to pressure perturbations. To evaluate the compliance
 307 effect, for a given frequency, we consider elastic parameters λ and μ from the Prelimi-
 308 nary Reference Earth Model (PREM) (Dziewonski & Anderson, 1981) averaged down
 309 to a depth proportional to the tsunami wavelength, namely $1/k$ (Crawford, 2004). For
 310 wavelengths between 10 and 100 km, Equation 5 yields a sensitivity to pressure via the
 311 compliance effect of $\epsilon_x \sim (5.5 \times 10^{-11} - 4.8 \times 10^{-12}) \Delta P$.

312 With the elasticity parameters considered in this study for the cable assembly and
 313 those from the PREM model, the sensitivity to Poisson's (estimated in the previous sec-
 314 tion) and compliance effects are very similar. One feature that might help distinguish
 315 their relative contributions in field data is that only the compliance effect depends on
 316 the wave propagation direction (θ is involved in Equation (5) but not in Equation (7)).

317 The sum of Poisson's and compliance effects yields an order-of-magnitude theoret-
 318 ical estimate of pressure sensitivity of $\epsilon_x/\Delta P \sim 10^{-10} \text{ Pa}^{-1}$. This is similar to or some-
 319 what smaller to the observations made by Meulé et al. (2024), and recent empirical es-
 320 timates: $\epsilon_x/\Delta P \sim 5 \times 10^{-10} \text{ Pa}^{-1}$ in (Williams et al., 2023), $\epsilon_x/\Delta P \sim 10^{-9} \text{ Pa}^{-1}$ in
 321 (Glover et al., 2024). The latter estimate was obtained in shallow water (≤ 15 m) and
 322 for relatively short wavelengths associated to frequencies in the 0.04-0.3 Hz band, thus
 323 the difference can be partly attributed to the lower stiffness of shallow sediments and
 324 to larger contributions from shoaling and seafloor shear effects. An additional order-of-
 325 magnitude verification is based on DAS signal amplitudes reported by (Sladen et al., 2019)
 326 for surface gravity waves recorded at shallow depths (< 100 m). Considering their ob-
 327 served strain rates (extrapolated) at zero depth $\dot{\epsilon} \sim 2 \times 10^{-7} \text{ s}^{-1}$, their dominant pe-
 328 riod ~ 10 s, and $\epsilon_x/\Delta P \sim 10^{-10} \text{ Pa}^{-1}$, yields $\Delta P \sim 10^4 \text{ Pa}$. This pressure corresponds
 329 to $\Delta h \sim 1$ m, which agrees in order-of-magnitude with reported wave heights in the
 330 same region (Guerin et al., 2022).

331 Figure 4a,b shows the resulting strains as a function of frequency, along with the
 332 mean noise floor illustrated in Figure 2 for the calibrated version of a chirped-pulse DAS
 333 instrument. The strain amplitudes yield signal-to-noise ratios (SNR) larger than 100 for
 334 a wave height of 0.1 m, and $\text{SNR} > 1000$ for a wave height of 1 m. These results sug-
 335 gest that DAS could reliably detect signals from the pressure induced by tsunami waves.
 336 The stacking of multiple DAS channels could further improve detectability; with typ-
 337 ical DAS gauge lengths ~ 10 m, it would be feasible to stack 100 channels along cable
 338 segments of ~ 1 km length that are still much shorter than a typical tsunami wavelength,
 339 potentially improving SNR by a factor of 10.

340 4 Seafloor Strain from Shear Stress Beneath a Tsunami Wave

341 The displacement of water masses by a tsunami wave can induce sea-bottom shear
 342 strain. The horizontal free-stream velocity beneath a tsunami wave of frequency ω can
 343 be estimated from the particle velocities given by linear wave theory (Dean & Dalrym-
 344 ple, 1991):

$$v_x(x, t) = A \frac{\cosh(kz)}{\sinh(kh)} \sin(kx - \omega t)$$

$$v_z(x, t) = A \frac{\sinh(kz)}{\sinh(kh)} \cos(kx - \omega t)$$
(11)

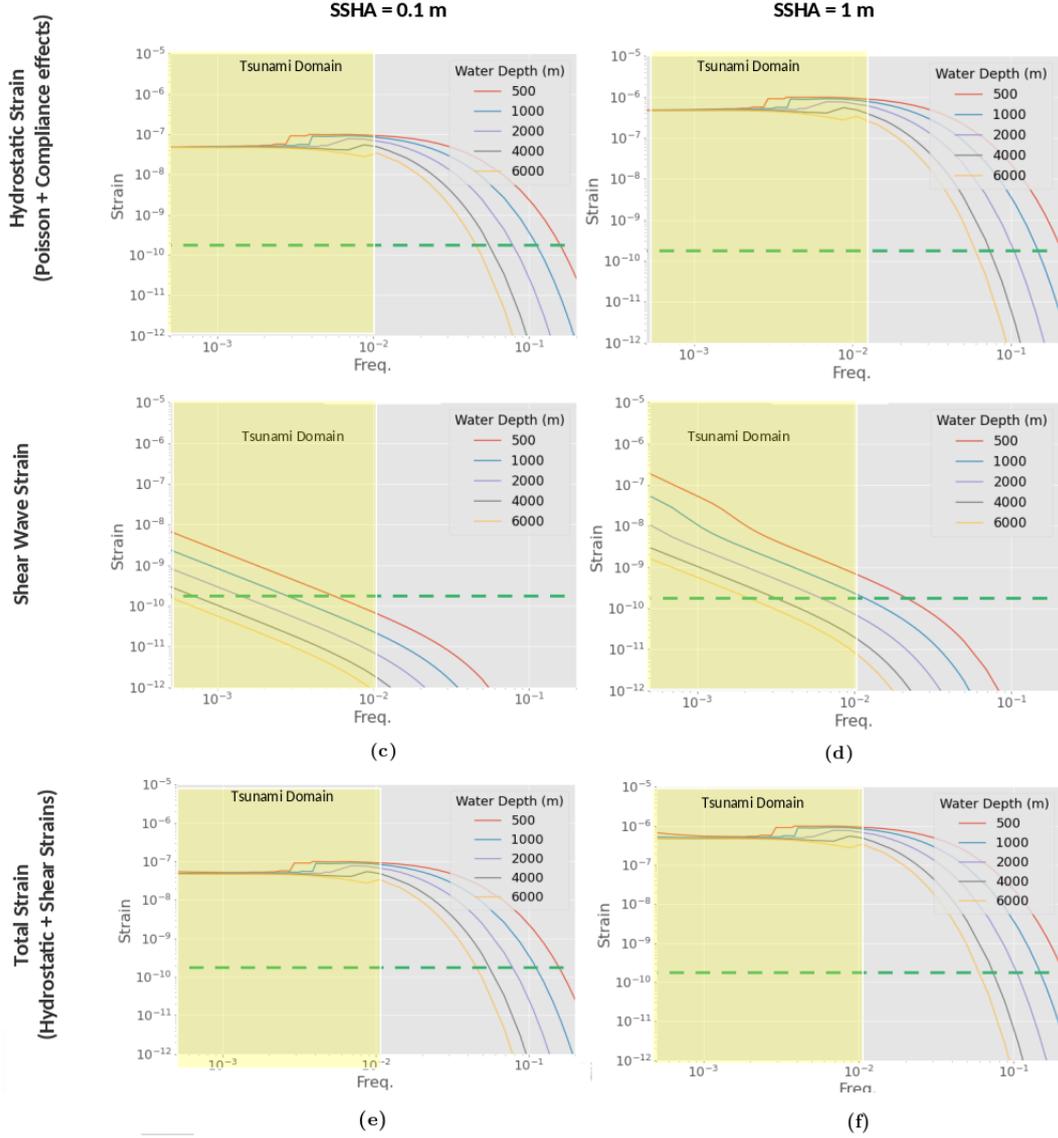


Figure 4. Expected strain amplitudes for tsunami heights (sea surface height anomalies, SSHA) of 0.1 m and 1 m, as indicated in the column headers. Figures in rows correspond to the strain from hydrostatic pressure at the seafloor (a-b), the bottom strain due to the shear stress from the horizontal flow motion beneath a tsunami wave (c-d), and the combined strain from the aforementioned sources of strain (e-f). The different colored curves represent different sea depths h as indicated on each legend; the green dashed line represents the noise floor for the calibrated DAS unit. The yellow-shaded region encompasses the range of frequencies corresponding to tsunami waves. For the cable assembly, nominal values of $E=25$ GPa and $\nu=0.23$ were used for Young’s modulus and the Poisson’s ratio, respectively. The seafloor is modeled as a stratified solid with elastic properties based on the PREM model. The ocean has a density $\rho = 1030$ kg/m³, the ocean has an acoustic wave speed of $c_p = 1500$ m/s and the water particle viscosity $\nu_s = 1E-6$ m²/s.

345 where v_x and v_z are the horizontal and vertical particle velocities, respectively, z
 346 is vertical position relative to the seafloor, and

$$A = \frac{\Delta h}{2} \sqrt{\frac{g}{h}} \quad (12)$$

347 The vertical velocity peaks at the ocean surface and decays exponentially to zero
 348 at the seafloor. In contrast, the horizontal velocity remains relatively constant through
 349 the water column.

350 At the seafloor ($z = 0$):

$$v_x(x, t) = \frac{\Delta h}{2} \sqrt{\frac{g}{h}} \frac{1}{\sinh(kh)} \sin(kx - \omega t) \quad (13)$$

351 The resulting bottom shear stress is

$$\tau_w = \frac{\rho}{2} f v_x^2 \quad (14)$$

352 where f is the friction coefficient. The associated shear strain is

$$\epsilon_{shear} = \frac{\tau_w}{\mu} \quad (15)$$

353 where μ is the elastic shear modulus of the seafloor.

354 The friction coefficient that should be used under a tsunami wave is determined
 355 by the flow regime on the bottom boundary layer. According to the analysis by (Tinh
 356 & Tanaka, 2019), for most of the propagation phase, the tsunami-induced bottom bound-
 357 ary layer shows an unsteady behavior and resembles that induced by wind-driven waves,
 358 even under long-period wave motion; with the transition from wave to steady-motion typ-
 359 ically occurring only a few meters from shore. The bottom boundary layer at the tsunami
 360 source is within the laminar regime and subsequently, a transition occurs to smooth tur-
 361 bulance during the shoaling process, with a transition from smooth to rough turbulent
 362 region at shallow depths. In this study, the full-range equation proposed by (Tanaka et
 363 al., 2020, Eq.18) for the wave-current combined motion, is implemented to compute the
 364 wave friction coefficient under shoaling tsunami, f_w , given that it yields a smoothly in-
 365 terpolated value for each of the aforementioned flow regimes including the transitional
 366 domain. The full-range equation is a function of the wave-induced velocity v_x , the an-
 367 gular frequency of the wave, ω , the roughness of the seabed material (sand grain diam-
 368 eter), and the viscosity of the fluid.

369 Figure 4c,d shows the estimated shear strain levels as a function of frequency. Fig-
 370 ure 4e,f summarises the total expected strain signal at the seafloor from Poisson's effect,
 371 seafloor compliance, and bottom shear strain. The two first effects are largely dominant.
 372 The shear strain has an appreciable impact only at low frequencies, at shallow depth,
 373 and high SSHA (the case $h = 500$ m and $\Delta h = 1$ m in Figure 4f).

374 5 Strain from a Propagating Tsunami Wave

375 To study the strain signal as a tsunami wave propagates, we consider a tsunami
 376 wave traveling from its source towards the coast, up to a water depth of $h = 100$ m. In
 377 the source area, the wave height is Δh_0 and the water depth is h_0 . To estimate the evo-
 378 lution of strain during the shoaling process, we approximate the spatial variation of wave
 379 height Δh using Green's law:

$$\Delta h = \Delta h_0 \left(\frac{h_0}{h} \right)^{1/4} \quad (16)$$

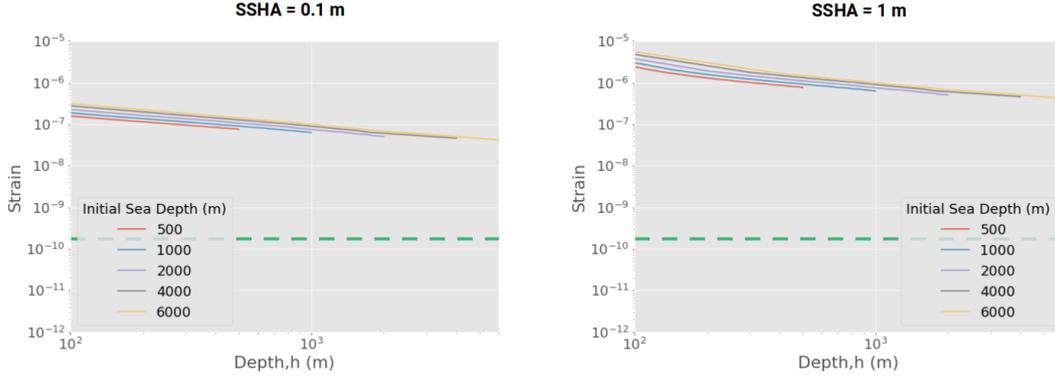


Figure 5. Calculated strain amplitudes from traveling tsunami waves generated at various sea depths h (indicated on the legend) and propagating through decreasing water depths towards the coast, up to a depth of 100 m. Results are for initial tsunami heights - sea surface height anomalies (SSHA) of (a) 0.1 m and (b) 1 m. These results are the average from 0.5 mHz to 11 mHz, and represent the total combined strain from hydrostatic seafloor pressure and the bottom shear stress due to the horizontal flow motion beneath a tsunami wave. The green dashed line represents the noise floor of the calibrated DAS unit.

380 For each water depth value h , the corresponding wave height Δh is calculated. These
 381 values are then used to calculate the wavenumber k through equation (8), for a range
 382 of frequencies between 0.5 mHz up to 11 mHz. This is then used to compute the total
 383 strain resulting from the three effects considered so far.

384 Once again, we observe that the strain from hydrostatic pressure is the dominant
 385 signal for most of the tsunami trajectory. The contribution from shear strain becomes
 386 appreciable only for the cases with initial SSHA = 1 m (Figure 5b), when the tsunami
 387 wave reaches shallower depths. This manifests as a slight increase of the slope of each
 388 curve at shallow depths in comparison to Figure 5a.

389 The tsunami strain signal yields an SNR of at least 200, reaffirming the feasibil-
 390 ity of tsunami wave detection with DAS instrumentation. Towards implementation, in
 391 the following sections, we consider signals that may overlap with tsunami signals, po-
 392 tentially affecting detection capabilities.

393 6 Influence of Temperature Perturbations

394 Temperature and strain signals co-exist in DAS data. Both phenomena affect the
 395 optical path along a fiber: strain perturbations change the fiber length and temperature
 396 fluctuations change the refraction index through the thermo-optical effect. Temperature
 397 perturbations ΔT produce a proportional apparent DAS strain perturbation $\Delta \epsilon$. The
 398 proportionality factor is a function of the refractive index, thermo-optic and elasto-optic
 399 responses of the fiber silica (Fernandez-Ruiz et al., 2022; Haavik, 2022; Hartog et al., 1979).
 400 A representative estimate is:

$$\Delta \epsilon \approx 10^{-5} \times \Delta T \quad (17)$$

401 The mean noise floor $\epsilon \approx 2 \times 10^{-10}$ of the DAS instrument under consideration corre-
 402 sponds to a temperature perturbation of approximately 20 μK . Given such fine sensi-
 403 tivity, it is necessary to evaluate whether thermal effects may overlap with the mechani-
 404 cal strain signals induced by tsunami waves. Here we consider thermal effects due to
 405 (1) background temperature fluctuations and (2) tsunami-induced fluid advection.

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6.1 Background Temperature Fluctuations

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Long-term oceanographic observations in deep water indicate that temperature fluctuations at the ocean bottoms generally have dominant periods ranging from hours to weeks. Their frequency content is highly dependent on site characteristics (depth, geographic location, bathymetry, etc). For our purposes, such long-period signals can be filtered out if they do not overlap with the tsunami period range.

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Frequency content and water depth can serve as a basis to distinguish strain from temperature signals in DAS seafloor data. At shallow water depths (< 200 m) and at short periods ($10^{-2} - 10^2$ s), mechanical strains from seismic, acoustic or surface gravity waves dominate over thermal fluctuations driven by the oceanic environment. This differentiation has allowed seismic phase detection with ocean-bottom DAS (e.g., Lior et al. (2020); Ugalde et al. (2022)) and observations of coastal currents (Lindsey et al., 2019; Williams et al., 2019). In deep water or at long periods, thermal and mechanical DAS signals may have comparable amplitudes.

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Recent studies suggest ocean-bottom DAS can record temperature. (Ide et al., 2021) interpreted DAS signals at tidal periods and with apparent propagation speeds of 0.5 m/s, recorded along a cable offshore Japan, as temperature fluctuations of a few Kelvins due to water advection driven by tides. (Williams et al., 2023) observed, along a cable off the island of Gran Canaria, the propagation of thermal fronts associated with non-linear internal tides, with apparent velocities in the 0.1 - 0.5 m/s range.

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In general, thermal anomalies driven by oceanographic processes are larger in shallow waters. Ocean-bottom DAS data confirms that temperature fluctuations are weaker with increasing depth. Off the coast of Toulon, France, (Pelaez Quiñones et al., 2023) found temperature fluctuations in the 0.1 - 1 mHz band reaching 0.1 K at depths shallower than 100 m and about 0.01 K at further depths down to 2000 m, with apparent velocities close to 0.1 and 0.01 m/s, respectively. Off Gran Canaria, (Williams et al., 2023) observed perturbations up to about 2 K at 1.4 km depth and 0.2 K at 2.5 km depth.

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The thermal DAS signals due to internal waves and tides may overlap with the mechanical DAS signals due to tsunami waves in the mHz range. From (17), fluctuations larger than 0.01 K will correspond to strain signals above 10^{-7} , which may exceed tsunami signal amplitudes (Figure 4e,f). Yet, these temperature anomalies propagate about 3 orders of magnitude slower than tsunami waves, thus it should be possible to filter them out based on their apparent speed $C_x = C / \cos(\theta)$, where C is the phase velocity and θ is the angle between the direction of propagation of the temperature front and the cable orientation. Although oblique incidence tends to increase the apparent velocity, only a very narrow range of incidence angles, within 0.2° from normal incidence, would make the temperature signal velocity comparable to tsunami velocities. Hence, frequency-wavenumber filtering based on signal speed is expected to be a viable approach.

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If the cable is not straight, the wavenumber decomposition cannot be obtained by Fourier transform, and alternate strategies need to be devised. This may consist of in situ calibration of the instrument by characterizing temperature variations that may be pervasive and endemic to a given offshore location; to consider cable construction and burial, and to make key observations of the environment. Temperature fluctuations on seafloor DAS recordings may also be mitigated by burying the cable. Indeed, (Pelaez Quiñones et al., 2023) and (Williams et al., 2023) reported a lack of temperature signals along cable sections that are buried. While buried sections along telecom cables are rare, they could be a design consideration for fiber optic cables tailored for geophysical monitoring.

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6.2 Tsunami-Induced Temperature Fluctuations

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In principle, seafloor temperature perturbations can arise from advection of the thermally stratified water column by a tsunami wave. However, such signals have not been observed yet. Temperature data during tsunamis has been fortuitously recorded in association with seafloor pressure observations, as part of the system that compensates for the thermal drift of pressure transducers (Eble et al., 1989; Joseph, 2011). One ocean bottom station in the 2011 $M_w=9.0$ Tohoku earthquake source area, at a sea depth of 1.1 km, recorded a water temperature increase of 0.19 °C about 3 h after the earthquake, lasting for several hours (Inazu et al., 2023). This temperature transient was attributed to a tsunami-generated turbidity current (Arai et al., 2013). During the 2003 $M_w=8.3$ Tokachi-oki earthquake and tsunami, a CTD (Conductivity-Temperature-Depth) station located in the source area recorded a temperature perturbation at least two hours after the mainshock, which was attributed to a benthic storm (Mikada et al., 2006). Furthermore, the temperature data at two seafloor pressure stations found no significant change in the tsunami pressure signal after applying a temperature correction (Inazu & Hino, 2011). As the temperature measurements from quartz crystal transducers are primarily intended to compensate for the thermal drift of the pressure gauge, their temperature resolution is limited. In the absence of well-resolved observations of tsunami-induced seafloor temperature changes, we turn next to back-of-the-envelope theoretical analysis.

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An order-of-magnitude estimate shows that temperature changes due to deep water advection by a tsunami wave could be recorded by seafloor DAS, but would be much smaller than the strain mechanically induced by the tsunami. Considering a SSHA of 10 cm, the horizontal water particle displacement at the sea bottom is of at least 1 m (Ward, 2003). Assuming this same horizontal advection follows along a typical slope of 5% between a continental shelf and a subduction trench, it displaces water vertically by 0.05 m along the slope. Considering a generic vertical temperature gradient for the open ocean of 0.002 K/m at depths exceeding 1 km (Talley, 2011), the vertical water advection carries a temperature change of ~ 0.1 mK. This value corresponds to a DAS strain $\sim 10^{-9}$, which is above the calibrated DAS noise floor of 2×10^{-10} , but about two orders of magnitude below the expected strain mechanically induced by tsunami waves (Figure 4e).

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7 Comparison to Fully Coupled Earthquake and Tsunami Simulations

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With the improvements to DAS instrument sensitivity at low frequencies illustrated in Figure 2, our analysis of the expected seafloor strain due to tsunami waves (Figures 4 and 5) points to the feasibility of tsunami detection with seafloor DAS. To further demonstrate the potential contribution of DAS to TEWS, we analyze here a synthetic data set from a fully physics-based 3D simulation of earthquake dynamic rupture and tsunami generation.

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7.1 Coupled Ocean-solid Earth Simulation Setup

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We consider an idealized megathrust earthquake on a low-angle planar thrust fault, with subsequent tsunami generation and propagation across a compressible ocean layer with constant depth. The model setting is based on a benchmark scenario for linking earthquake and tsunami simulations established in (Madden et al., 2021; Krenz et al., 2021). The fault strikes North, is 200 km wide along the strike, extends from the surface to 35 km depth, and dips eastward with a 16° dip angle (Figure 6). The ocean has a water depth of 2 km, acoustic wave speed of $c_p = 1500$ m/s and density $\rho = 1000$ kg/m³. The solid half-space has homogeneous elastic properties representative of oceanic crust: P-wave speed $c_p = 7639.9$ m/s, shear wave speed $c_s = 4229.4$ m/s and density $\rho = 3775$ kg/m³.

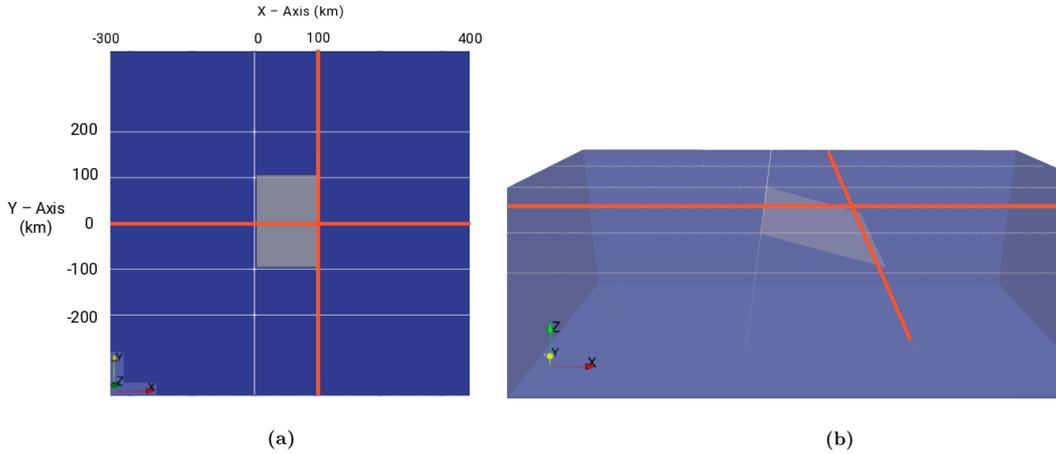


Figure 6. Benchmark simulation scenario showing location and dimensions of the planar subducting plate relative to the receiver matrix. The simulation space encompasses an area of approximately $750 \text{ km} \times 750 \text{ km}$, with a uniform 2 km water layer atop. Receivers along each linear array are arranged in pairs separated by 50 m, respectively, which emulates commonly used DAS gauge lengths; the next adjacent receiver pair is placed 450 m apart. The simulation consisted of two identical matrices, one buried 10 cm below the seafloor, whilst the other matrix is located 10 cm below the sea surface. Given the orientation of the subducting plate, increasing positive values move towards the coast along the X-axis, whilst negative values go seaward. X and Y coordinate values are given in meters. The highlighted arrays at $Y = 0 \text{ km}$ and $X = 100 \text{ km}$ are used in this analysis.

503 Because this setup ignores the presence of shallow compliant layers, the resulting seafloor
 504 compliance signal should be understood as a lower bound.

505 The simulation employs a fully coupled 3-D earth and ocean model of earthquake
 506 dynamic rupture, seismic and acoustic wave propagation, whilst simultaneously solving
 507 for the tsunami (gravity) wave propagation, implemented in the simulation package Seis-
 508 Sol (www.seissol.org). The tsunami is modeled through linearized equations, derived by
 509 combining mass balance with a linearized equation of state and momentum balance, with
 510 gravity acting as a restoring force and an initial small perturbation about the hydrostatic
 511 rest state of the ocean (Lotto & Dunham, 2015). The simulation self-consistently com-
 512 puts the full wavefield in 3-D, which comprises seismic, acoustic, and surface gravity
 513 waves in elastic (earth) and acoustic (ocean) media.

514 SeisSol is based on the arbitrary high-order derivative Discontinuous Galerkin (ADER-
 515 DG) method and optimized for modern high-performance computing infrastructure (e.g.,
 516 Heinecke et al. (2014); Uphoff et al. (2017); Krenz et al. (2021)). Here, we use a fifth-
 517 order accurate scheme and an unstructured tetrahedral mesh with a minimum element
 518 size of 66 m (on the fault), consisting of 29.5 million elements. We simulate 10 minutes
 519 of combined earthquake dynamic rupture, tsunami generation, and tsunami propagation
 520 on the supercomputer SuperMUC-NG, which requires a total of $\sim 460,000$ CPUh using
 521 512 nodes (24,576 cores) for ~ 18.7 hours.

522 The earthquake source evolves spontaneously during the earthquake dynamic rup-
 523 ture simulation governed by linear slip-weakening friction (e.g., Harris et al. (2018)). The
 524 earthquake hypocenter is located in the southeast corner of the fault at 26 km depth.
 525 While the fault does reach the seafloor, the rupture itself is buried; higher fault strength

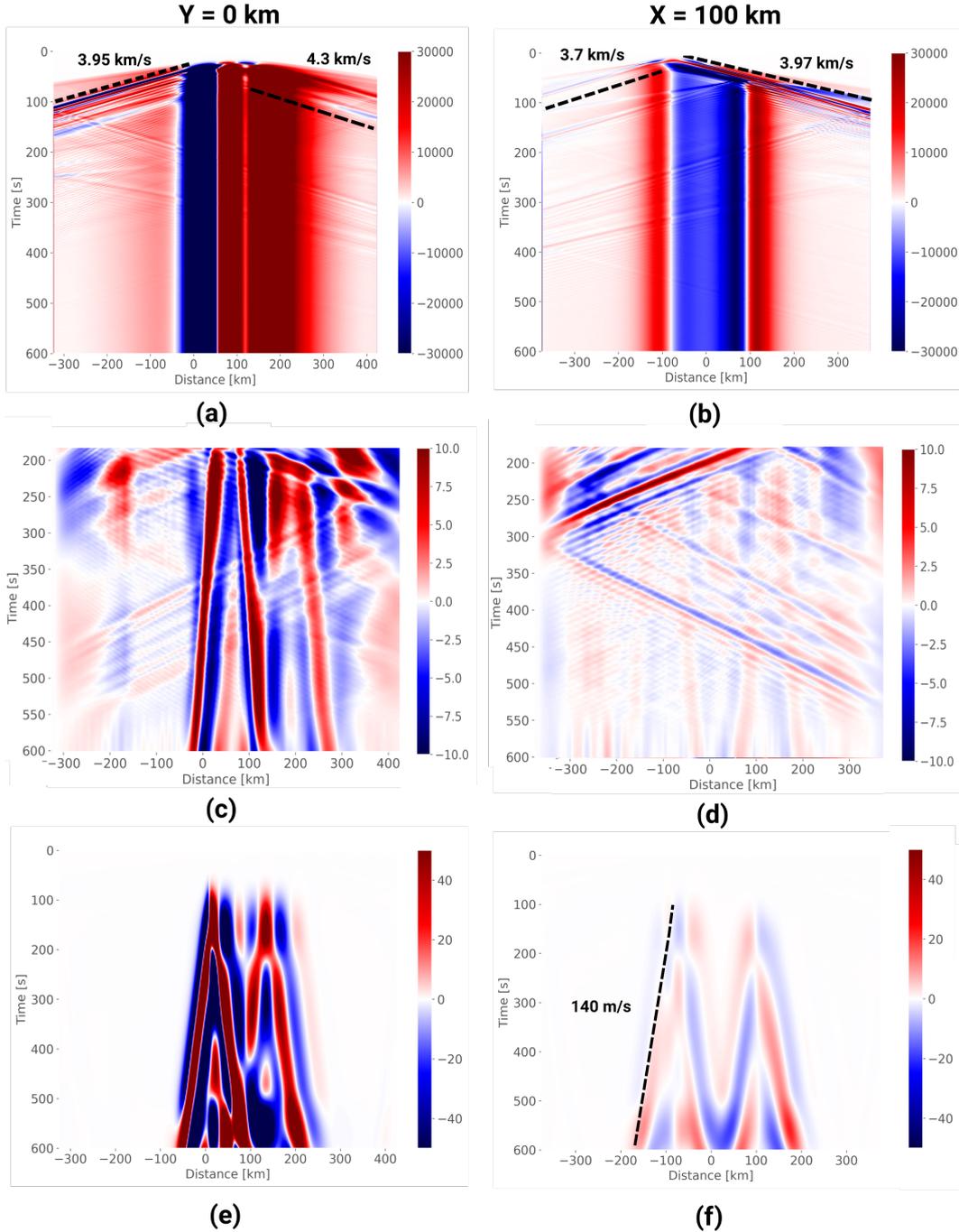


Figure 7. Strain signal corresponding to a trench-parallel array located at $y = 0$ km (left column) and a trench-orthogonal array at $x = 100$ km (right column). Top (a, b): Broadband strain recorded on seafloor-coupled receivers. Dashed black lines indicate seismic and surface wave velocities. Middle (c, d): same after removing the initial 180 secs, subtracting the mean of the final 180 secs (static seafloor deformation) from each receiver and F-K filtering seismic surface waves by bandpass filtering between 3 - 10 mHz onto the range of wavelengths (11 - 66 km) corresponding to propagating velocities between 110 - 200 m/s. Bottom (e, f): Estimated strain signal due to seafloor pressure (seafloor compliance and the cable’s Poisson effect), derived by subtracting the vertical seafloor displacement from the sea surface height anomalies (SSHA, recorded by receivers near the top of the water column), F-K filtered as in (c, d). Strain is expressed as a signal-to-noise ratio (SNR) relative to the mean noise floor of the DAS instrument shown in Figure 2. The dashed black line in (d) indicates the tsunami phase velocity.

526 at shallow depth smoothly stops the rupture as it approaches the surface. The earth-
527 quake magnitude is $M_W = 8.5$ and the rupture propagates at 3.5 km/s on average.

528 Simulation outputs were recorded on 42000 receivers, slightly buried below the seafloor.
529 Each receiver records velocity and stress at a 100 Hz sampling rate. They are arranged
530 by pairs, with an intra-pair separation of 50 m, comparable to a commonly used DAS
531 gauge length, and an inter-pair separation of 500 m. The receivers are arranged in 7 lin-
532 ear arrays, either parallel or orthogonal to the trench (Figure 6). The trench-parallel and
533 trench-normal linear arrays, across and in the vicinity of the earthquake source area, rep-
534 resent end-member configurations of a seafloor cable.

535 Based on our previous analysis, we focus on longitudinal seafloor strain induced
536 by the tsunami wave only from hydrostatic pressure, namely the compliance and Pois-
537 son effects (equations 6 and 7, respectively). We calculate the seafloor strain taking the
538 finite difference between pairs of seafloor receivers as $\epsilon = \frac{u_1 - u_2}{d}$, where u_i is the dis-
539 placement component parallel to the direction of the linear array at receiver i , and $d =$
540 50 m is the intra-pair receiver distance. The displacement is obtained by time integra-
541 tion of the velocity data.

542 To record SSHA, an additional set of receivers was placed near the top of the wa-
543 ter column, directly above the seafloor receivers. We extract the SSHA from the verti-
544 cal displacement of these receivers. Seafloor pressure is calculated from equation 10 based
545 on the effective water column height, obtained by subtracting vertical seafloor displace-
546 ment from SSHA. We then compute the estimated strain signal due to seafloor compli-
547 ance and the cable's Poisson effect by using equations 6 and 7.

548 7.2 Enhancing the Tsunami Signal

549 The resulting seafloor strain fields for two profiles, one perpendicular to the fault
550 ($y = 0$ km) and the other along the buried tip of the fault ($x = 100$ km), are shown
551 in Figure 7 (a) and (b), respectively. The strain field is largely dominated by seismic and
552 acoustic waves while the rupture propagates across the fault for ~ 60 s. The strain field
553 continues to change after this time until it reaches final values at $t \sim 80$ s (Madden et
554 al., 2021). From then on, the seafloor strain is dominated by the static deformation above
555 and around the rupture area. Static displacements reach ~ 3 m along the $y = 0$ km
556 line and 30 cm along the $x = 100$ km line.

557 The simulation data also contains spurious reflections originating from the bound-
558 aries of the computational domain. Thanks to the (non-perfect) absorbing boundary con-
559 ditions, the amplitudes of these reflections are very small compared to the physical seismo-
560 acoustic waves. However, they are large compared to the predicted tsunami-induced strain
561 signals. These artifacts could be reduced with a simulation over a larger domain at higher
562 computational cost or by employing more advanced boundary conditions, which is a non-
563 trivial problem (e.g., Duru et al. (2019)).

564 Through data processing, we managed to visualize the tsunami signal in the sim-
565 ulation results. The simulated tsunami has a dominant wavelength of ≈ 57 km, which
566 is much larger than the water depth. Thus, the shallow-water tsunami regime holds and
567 we should be looking for tsunami signals propagating at speed $v \approx \sqrt{gh} = 140$ m/s. Within
568 the initial 10 minutes simulated here, as the tsunami only propagates within the vicini-
569 ty of the tsunamigenic region, detection requires techniques to discern the tsunami sig-
570 nals which can be orders of magnitude below the seismo-acoustic wave and static defor-
571 mation signals. As exemplified in the raw broadband seafloor strain (Figure 7a,b), the
572 tsunami signal is not readily visible, but becomes observable after the initial 180 seconds
573 thanks to post-processing (Figure 7c,d). We first subtract on each channel an estimate
574 of the static strain, the mean over the final 180 seconds. We then apply a window taper.
575 We finally apply a frequency-wavenumber (F-K) filter in the frequency range of 3

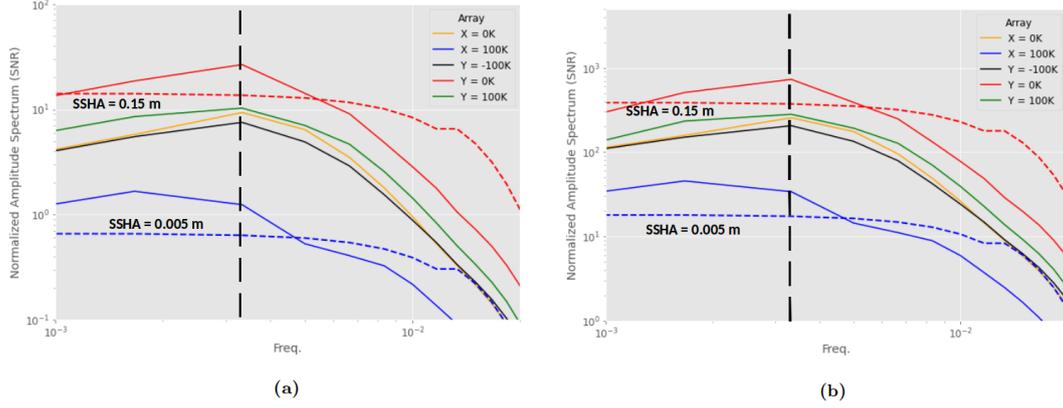


Figure 8. Amplitude Spectral Density (ASD) of the synthetic tsunami signal from the simulation due to (a) seafloor compliance and (b) the full hydrostatic pressure (Poisson and seafloor compliance). Pressure is derived from the effective vertical displacement, by subtracting the vertical displacement at the seafloor from the SSHA (receivers just below the sea surface). The PSD is calculated for each receiver and then averaged along each array. Power is converted to amplitude. The signal is normalized relative to instrument noise (SNR). Red and blue dashed lines follow the color assignment on the legend, where the labeled SSHA values correspond to the maximum displacement (after F-K filtering) observed on the given array. Such SSHA is used in equations 5 and 7 to define the predicted signal from the derived model. The dashed black line marks the dominant tsunami wave frequency.

576 - 10 mHz and wavelength range of 11 - 66 km. The wavelength range is derived from the
 577 frequency range by dividing by a range of wave velocities around the expected tsunami
 578 speed, namely 110 - 200 m/s. For the $y = 0$ km array, it was necessary to remove a single
 579 receiver in the vicinity of the trench which showed a spatially abrupt change in dis-
 580 placement polarity. Upon post-processing, we readily observe the tsunami wavefronts
 581 on the trench-parallel array (Figure 7c) and faintly on the trench-orthogonal array (Fig-
 582 ure 7d). The seismo-acoustic and static signals that dominate the broadband seafloor
 583 strains (Figure 7a,b) are about three orders of magnitude larger than the tsunami strain
 584 signals (Figure 7e,f). Making abstraction of the spurious reflections from the simulation
 585 boundaries, this analysis confirms the presence of the tsunami signal in the simulated
 586 seafloor strain data.

587 Real-time data processing, which is a key requirement for TEWS, could entail work-
 588 ing with strain rate instead of strain. An example of this approach is included in the sup-
 589plementary materials. After applying the same F-K filtering we remove the static sig-
 590nal. However, the boundary reflections from the simulation remain strong, which obfus-
 591cates the tsunami phase in this simulation.

592 7.3 Comparison to DAS Sensitivity and to Theoretical Model

593 We compute the expected DAS SNR by taking the ratio between the filtered strain
 594 and the mean noise floor of the DAS instrument shown in Figure 2. Because this sim-
 595 ulation did not model a seafloor cable, the contribution from Poisson's effect is not in-
 596 cluded in the simulation data shown in Figure 7c,d. But both seafloor compliance and
 597 Poisson's effect are included in Figure 7e,f, which is derived from the water column height
 598 simulation data. The SNR reaches values higher than 10 during the whole time window

599 along the trench-parallel array (Figure 7-c) and of order 1 along the trench-orthogonal
600 array (Figure 7d).

601 Among the DAS array configurations examined here, we find that the best suited
602 for tsunami detection is located close to the source area and perpendicular to the trench.
603 We compare the simulated seafloor compliance signals over all receiver arrays, to the strain
604 predicted by the theoretical model. We calculate the power spectral density of the es-
605 timated strain due to seafloor compliance at each receiver, based on their SSHA and ver-
606 tical seafloor displacement, then average along each array, and take the square root to
607 convert power to amplitude. The results are shown in Figure 8(a). The arrays at $y =$
608 ± 200 km are ignored because the tsunami does not reach them during the 10 minutes
609 of simulation. The expected tsunami signal from the full hydrostatic effect (Poisson ef-
610 fect and seafloor compliance) is shown in Figure 8(b). The strongest signal is recorded
611 on the $y = 0$ km array, which is aligned with the prevalent propagation direction of the
612 tsunami. Overall, the theory gives an adequate order-of-magnitude estimate of the sig-
613 nal across all arrays. The two dashed curves in Figure 8 are the estimated signal for two
614 selected values of SSHA (see labels), corresponding to the maximum SSHA for the ar-
615 ray with the largest SSHA ($y = 0$ km) and for the array with the lowest SSHA ($x =$
616 100 km). The theoretical and the simulation amplitudes agree in order of magnitude.
617 Their differences are attributable to the simplifying assumptions in the theoretical es-
618 timate: the same SSHA value at all frequencies, a single tsunami propagation direction.

619 8 Conclusion

620 Motivated by the need to advance tsunami warning systems and the fundamental
621 understanding of tsunami processes, we have evaluated, through theoretical analysis and
622 numerical simulations, the potential of Distributed Acoustic Sensing on seafloor fiber op-
623 tic cables for direct observation of tsunami waves.

624 We derived first-order estimates of the seafloor DAS signals potentially generated
625 by tsunamis. We considered two mechanisms through which the hydrostatic pressure per-
626 turbation of tsunami waves can induce longitudinal strain on a cable: the elastic defor-
627 mation of the seafloor (compliance) and the Poisson effect within the cable. We also quan-
628 tified two mechanisms by which the sub-horizontal deep water flow, induced by tsunami
629 waves, can generate DAS signals: seafloor shear and temperature advection. However,
630 we found them to have a relatively minor potential contribution in deep waters. Seafloor
631 shear contributes significantly only at shallow depths (< 100 m) and for relatively large
632 initial SSHA (> 0.1 m). The presence and amplitude of background temperature fluc-
633 tuations can be assessed a priori and, if problematic, they can be selectively filtered out
634 based on their slow propagation speed.
635 If the seafloor compliance and the Poisson effect on the cable represent the primary mech-
636 anisms through which DAS systems are anticipated to record the passage of tsunami waves;
637 in order to achieve a sensitivity similar to that of a DART station (1 mm amplitude at
638 4 km water depth) (Mungov et al., 2013), the minimum amplitude of the strain signal
639 to be resolved is $\epsilon \approx 5 \times 10^{-10}$ at frequencies in the order of tens of mHz. The cur-
640 rent values of single-channel strain noise ASD of conventional DAS equipment operat-
641 ing at kHz sampling frequencies suggest that these water column changes will not be eas-
642 ily resolved without using the information from the full array. However, beamforming
643 methods allowing fairly large processing gains, may probably lead to enough sensitiv-
644 ity, as shown in recent related efforts (Xiao et al., 2024). The strain noise ASD of con-
645 ventional DAS systems in this frequency range is dominated by $1/f$ noise caused by ref-
646 erence updates and laser frequency drifts. Methods for improving the sensitivity of DAS
647 instrumentation at these frequencies, such as the multi-frequency calibration technique
648 (Vidal-Moreno et al., 2022), together with laser frequency stabilization, have already demon-
649 strated a way to strongly mitigate this noise component and reach the low-frequency per-
650 formance necessary for tsunami detection at almost single-channel level.

651 To further evaluate the use of DAS for TEWS, we carried out a full-physics sim-
 652 ulation representative of a large earthquake rupture in a subduction zone. This simu-
 653 ulation allowed to assess our ability to separate the tsunami contribution from the static
 654 and dynamic earthquake components, depending on their relative distance and the ori-
 655 entation of the cable. We found that the preferred cable orientation is aligned with the
 656 horizontal displacement most likely to occur (i.e. perpendicular to the trench), as the
 657 tsunami energy and propagation will mirror such a pattern and because DAS records lon-
 658 gitudinal strain. While less optimal, cables perpendicular to the horizontal displacement
 659 are still relevant for tsunami detection. To optimize early warning times, it is best to de-
 660 tect the tsunami in the generating region, but cables directly over the subducting slab
 661 will also record the quasi-static seafloor displacement. The latter is orders of magnitude
 662 larger than the former and both have similar spatio-temporal characteristics. We show
 663 that it is still possible to retrieve the tsunami signal after appropriate filtering and af-
 664 ter a few minutes of signal propagation. Furthermore, as seismo-acoustic waves can in-
 665 duce seafloor strains exceeding the maximum strain variation limit of the instrument,
 666 $\Delta\epsilon_{max} \approx \pm 82.5 \times 10^{-6}$, these signals will momentarily saturate the instrument. This
 667 would delay the tsunami detection by a few minutes (3 minutes in our simulation), but
 668 this delay remains compatible with the design of local TEWS as tsunami waves usually
 669 take more than 10-15 minutes to reach the nearby coasts. Possible strategies to reduce
 670 this delay may consist of customized signal processing, monitoring for the tsunami sig-
 671 nal from strategically selected segments of the array, away from the strongest seismic sig-
 672 nals, or by selecting a different array placement altogether.

673 The signal strength derived in the simulation can be considered a conservative esti-
 674 mate as the model consisted of a rigid solid earth model, representative of the oceanic
 675 crust. A more compliant seafloor due to the presence of sediments and less rigid layers
 676 near the surface amplifies low-velocity strain signals. This signal can be optimized by
 677 the degree of mechanical ground-cable coupling; this is especially the case when mon-
 678 itoring the compliance of low-confined upper layers containing large amounts of soft soils.
 679 Further enhancement can be attained by utilizing more compliant cables.

680 In summary, the theoretical and numerical considerations presented in this work
 681 point to the real possibility of detecting tsunami waves via fiber optic cable. With the
 682 vast existing network of submarine telecom cables, DAS provides an unprecedented op-
 683 portunity to monitor and anticipate tsunamis. The advantages of DAS over other ex-
 684 isting tsunami observation technologies include relatively low maintenance costs, wide
 685 and dense coverage, proximity to tsunamigenic sources (earthquakes or other), data avail-
 686 ability in real time. With ongoing improvements in sensitivity at low frequencies, DAS
 687 could soon be poised as a key technology to enhance regional TEWS and to advance our
 688 understanding of tsunami processes.

689 Open Research Section

690 The synthetic dataset described in section 7 and accompanying Python routine to
 691 process and generate Figure 7; as well as Python routine to simulate the strain model
 692 outlined in sections 3 through 5 and generate Figures 4 and 5 is accessible via (Becerril,
 693 2024) hosted by Recherche Data Gouv.

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