1 Application of a seismic network to baleen whale call detection and

2 localization in the Panama basin – a Bryde's whale example

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12 Abstract

13 Baleen whales use sounds of various characteristics for different tasks and interactions. This study focuses on recordings from the Costa Rica Rift, in the Eastern Tropical Pacific Ocean, made by 25 14 15 ocean-bottom seismographs and a vertical array of 12 hydrophones between January and February 16 2015. The whale calls observed are of two kinds: more commonly, repetitive 4-5 s long signals 17 separated into two frequency bands centered at ~20 and ~36 Hz; less commonly, a series of ~0.5-18 1.0 s long, lower amplitude signals with frequencies between 80 and 160 Hz. These characteristics 19 are similar to calls attributed to Bryde's whales which are occasionally sighted in this region. In this 20 study, the repetitive calls are detected using both the STA/LTA approach and a network empirical 21 subspace detector. In total, 188 and 1891 calls are obtained for each method, demonstrating the 22 value of the subspace detector for highly similar signals. These signals are first localized using a non-23 linear grid search algorithm and then further relocalized using the double-difference technique. The

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- 24 high-resolution localizations reveal the presence of at least seven whales during the recording period,
- 25 often crossing the instrument network from southwest to northeast.

27 I. INTRODUCTION

In water, the exchange of information by sight and smell is only possible over small 28 29 distances. Instead, many marine animals, and marine mammals in particular, have developed ways of 30 exchanging information using vibration and sound. Approaches include vocalizations, such as calls 31 and songs, and non-vocal sounds, for example generated by tail and fin slapping (Kavanagh et al., 32 2017). Marine mammals use these sounds for various purposes such as communicating, foraging and 33 socializing (e.g., Frankel, 2009; McDonald et al., 2006). Cetacean vocalizations, in particular, have a 34 wide range of signal characteristics, including waveform shape and frequency content (e.g., Richardson 35 et al., 2013; Frankel, 2009). The frequency characteristics of baleen whale vocalizations vary from 36 several Hz for Blue whales (Balaenoptera musculus) to more than 20 kHz for Humpback whales 37 (Megaptera novaeangliae), and call durations can range from a few tenths of a second to few tens of 38 seconds (Frankel, 2009).

39 In the Eastern Tropical Pacific Ocean, where the Costa Rica Rift (CRR) is located (Figure 1), various baleen whale species have been encountered, including Humpback whales, Bryde's whales 40 41 (Balaenoptera brydei and Balaenoptera edeni), Minke whales (Balaenoptera acutorostrata), Fin whales 42 (Balaenoptera physalus), Sei whales (Balaenoptera borealis), and Blue whales (Wade & Gerrodette, 1993; 43 Palacios et al., 2012). These species have different abundances and vulnerability levels in this region, 44 but experience the same threats corresponding to climate change, animal bycatch, ghost nets, ship 45 noise and ship strike (Palacios et al., 2023). Baleen whales are vocal in the frequency range of 46 seismometers, geophones and hydrophones which could, thus, provide a way of monitoring their 47 distribution and behavior. Such low frequency (<100 Hz) sound emissions can propagate over 48 several hundred kilometers, due to the relatively low sound attenuation in water and the presence of 49 a low-velocity waveguide in the water column. However, increase in ship noise may mask such

50 communications, and might induce changes in their vocalizations (McDonald et al., 2009; Hatch et al.,

51 *2012*).

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FIG. 1. (Color online). a) Bathymetric map of the Panama Basin, showing the position of the Costa
Rica Rift (CRR). GSC: Galapagos Spreading Center, IT: Inca Transform, ER: Ecuador Ridge, EFZ:
Ecuador Fracture Zone, PFZ: Panama Fracture Zone. b) Zoom-in of the grid of OBSs (numbered
black triangles) and on the VA (red triangle) deployed close to the CRR (white dashed box in (a)).
The construction of the VA is shown between maps. Hydrophone positions are indicated by yellow
shaded sections.

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Passive acoustic monitoring of whales using acoustic instruments informs knowledge of their
ecology and the possible impact of anthropogenic noise. Such approaches can also monitor, over the
long term, large portions of the oceans, but they are costly and large gaps in coverage remain
(*Wilcock & Hilmo, 2021*). Ocean-bottom seismograph (OBS) deployed for other purposes (e.g.,
seismic surveys, seismicity monitoring), can also provide complementary monitoring data (e.g.,

66 McDonald et al., 1995; Dunn & Hernandez, 2009; Brodie & Dunn, 2015). To contribute to the study of

67 whales in the Eastern Tropical Pacific Ocean region, we take advantage of the deployment of ocean-68 bottom seismographs and hydrophones close to the CRR in the Panama basin (Hobbs & Peirce, 2015) 69 (Figure 1). These instruments were originally deployed to study the lithospheric structure of the 70 CRR and its surroundings using mainly seismic data (e.g., Wilson et al., 2019; Robinson et al., 2020; 71 Funnell et al., 2021; Peirce et al., 2023). More specifically, we here focus on the analysis of whale calls 72 observed in the data of the North Grid (NG) where 25 closely-spaced OBSs and a vertical array 73 (VA) of 12 hydrophones were deployed in January-February 2015 (Figure 1). 74 Whale calls are generally detected using energy-based methods, spectrogram correlation (Širović et 75 al., 2004) and template matching (Socheleau et al., 2015; Matias & Harris, 2015). Here, we use the 76 STA/LTA (short-term average/long-term average) approach and, taking advantage of the high 77 whale call similarity, a network empirical subspace detector (De La Hoz et al., 2021) to detect the 78 whale calls. The STA/LTA algorithm is based on the ratio of the average of signal values calculated 79 for a short sliding window and a long sliding window. The subspace detector uses a basis of signal 80 vectors to detect other similar signals, by calculating the decomposition of a range of similar signals 81 projected onto the continuous data (Harris, 2006). We then localize the identified whale calls using a 82 non-linear grid search algorithm and the timing differences observed between different geophones 83 and hydrophones (Lomax et al., 2009). In our study the calls are highly similar and recorded by a 84 dense network of OBSs, thus, we can also refine the localizations, to high resolution, using the 85 double-difference method (Waldhauser & Ellsworth, 2000). We then analyze the call characteristics 86 and their spatio-temporal distribution, providing detailed information on signals and whale 87 behaviors, which is fundamental for their identification, management and conservation.

89 II. DATA AND METHODS

90 A. Data acquisition

91 This study uses data recorded by OBSs and a vertical array of hydrophones provided by the 92 UK's OBIF (Ocean-Bottom Instrumentation Facility - see Acknowledgements) and deployed 93 during RRS James Cook cruise JC114 of the OSCAR research project (Hobbs & Peirce, 2015). The 94 primary objectives of this deployment were to image the oceanic crust around the CRR and study 95 fluid interactions between crust and the ocean through the acquisition of multi-channel seismic 96 reflection data and wide-angle seismic profiles (e.g., Gregory, 2018; Wilson et al., 2019; Robinson et al., 97 2020; Funnell et al., 2021; Peirce et al., 2023). The acquired data, thus, has periods with and without 98 active-source seismic operations. During the non-active shooting periods, the passive recordings 99 provide valuable information for seismic attenuation estimates (Vargas et al., 2018) and the degree of 100 the natural microseismicity (Lowell et al., 2020; Tary et al., 2021). In this study, we focus on the 101 detection and localization of the whale calls that were also recorded. 102 The OBSs were deployed \sim 5 km apart on the seafloor within an \sim 20 x \sim 20 km footprint, and 103 the VA was deployed in the southwest corner of this grid (Figure 1). The overall recording period ran between 26th January and 17th February 2015, although individual instruments have different 104 105 recording periods due to their respective recovery times. All but five OBSs (OBSs 4, 7, 14, 17, 24) were recovered on 1st and 2nd February. The remaining five OBSs as well as the VA were recovered 106 later on 16th and 17th February. The deepest hydrophone of the VA was positioned at 539 m above 107 108 the seafloor (at approximately 2858 m depth below sea surface), with the remainder spaced at \sim 5 m 109 intervals upwards through the water column (i.e., over 55 m). One of these hydrophones did not 110 record any useful data, most likely due to it being detached from the datalogger by strong lateral 111 surface currents during deployment (Hobbs & Peirce, 2015). Each OBS and the VA had High Tech 112 HTI-90-U hydrophones, with the OBSs also having, in addition, a three-component (3-C), shortperiod geophone set comprising Sercel L-28 4.5 Hz sensors. The data streams were sampled at 500
Hz. All instrument clocks were synchronized to GPS before deployment and on recovery, and any
drift was linearly corrected.

Although the airgun active-source seismic shots (GI, G, and Bolt array) hinder the detection ofnatural occurring signals such as the microseismicity, whale calls have different signatures (i.e.,

118 waveform and frequency content), and are generally observed by multiple stations given the dense

119 instrument distribution (Figure 2), improving the ability to reliably detect and localize these signals.

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1 **B.** Whale call detections

122 We initially analyzed the dataset to detect the microseismicity (Tary et al., 2021), using the 123 STA/LTA algorithm applied to band-pass filtered (between 5 and 120 Hz) geophone OBS data 124 (vertical component). Together with 1061 microearthquakes, we also detected whale vocalizations 125 with very similar signal characteristics (e.g., Figure 2). Considering the main signal frequency 126 components of these vocalizations, at ~20 and ~36 Hz, we re-applied the STA/LTA algorithm to 127 the same data, but instead filtered it between 20 and 45 Hz to remove as much background noise as 128 possible. To identify whale calls we used short and long windows of 3.0 and 15.5 s, respectively, and 129 recorded potential events when the STA/LTA ratio exceeded 3 at a minimum of three stations 130 within a time window of 10 s. The potential events were then manually reviewed to select the events 131 corresponding to whale calls.

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FIG. 2. Typical whale call recorded by the hydrophones of the vertical array (VA) and a selection
of OBS hydrophones (NG-), band-pass filtered between 10 and 45 Hz. P-wave time picks are
indicated by the black inverted triangles. The call origin time is 29th January 2015, at 15:30:24.

Taking advantage of the high signal similarity, we then applied the empirical subspace detector
to these detections (*Barrett & Beroza, 2014*). Contrary to the match filter where each previously
detected signal is used as a template to detect new events, the subspace detector employs the
previously detected signals to produce new signal subspace bases defined by a set of orthonormal

signal vectors U (*Harris, 2006*). The original signals are first separated into different clusters, which
are then used to calculate the subspace bases. The waveforms in each cluster in matrix A are aligned

145 on the P-wave arrival based on their cross-correlation coefficient, and the subspace base and its

146 orthonormal signal vectors are given by the singular value decomposition of matrix **A** as

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$$SVD(\mathbf{A}) = \mathbf{U}\mathbf{S}\mathbf{V}^{\mathrm{T}},\tag{5}$$

where U is also the matrix of left-singular vectors, and S and V are the singular values and
right-singular vectors, respectively. Then, only a small selection of left-singular vectors U_s, based on
the fraction of the total signal energy they represent for example (*Harris, 2006; Song et al., 2014*), is
used for detection. The different segments s of continuous data are then projected onto the matrix
U_s to define the detection metric of the subspace detector z following

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$$z = \frac{s^{\mathrm{T}} \mathbf{U}_{\mathbf{s}} \mathbf{U}_{\mathbf{s}}^{\mathrm{T}} s}{s^{\mathrm{T}} s}.$$
 (6)

154 Contrary to the subspace detector, the empirical subspace detector uses the stack of matrix **A** 155 and the first derivative of the stack with respect to time to populate $\mathbf{U}_{\mathbf{s}}$ instead of a selection of left-156 singular vectors. The stack and its first derivative closely resemble the first two left-singular vectors, 157 and using those instead of left-singular vectors can lead to a higher number of detections (Barrett & 158 Beroza, 2014). The detection is performed using the continuous hydrophone data, both from the 159 OBSs and the VA, band-pass filtered between 10 and 45 Hz. To take advantage of the dense array 160 of hydrophones, we use a network detection metric corresponding to the sum of the detection 161 metrics z over all stations in the network (De La Hoz et al., 2021). We sum the maximum of the 162 detection metrics z within a sliding window of 5 s to account of event moveout. The detection threshold γ_c was defined using correlations between the events in matrix **A** and random segments of 163 164 continuous data containing only noise (Song et al., 2014). This threshold was then adjusted by 165 reviewing detection results using representative data. We manually selected the detected events

166	corresponding to whale calls. The whale calls were picked automatically using the maximum of the
167	cross-correlation with the stack of matrix A , with the picks saved only if their cross-correlation
168	coefficients were greater than 0.65. These time picks were subsequently confirmed manually.
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C. Localization methods

171 *1. Absolute localizations*

Absolute event localizations were obtained using NonLinLoc (Lomax et al., 2001, 2009), which uses non-linear methods to search for the global maximum of the localization probability density

174 function. In most cases, the whale calls observed correspond to a P-wave travelling from the source

176 eikonal equation, which provides travel-times for the first, high-frequency wave to connect potential

to the instruments through the water. Theoretical P-wave travel-times were computed using the

177 sources and receivers. The crust was not included in the velocity model used for this calculation as

178 this leads the algorithm to calculate the faster crustal arrivals instead of water-borne arrivals. The

179 water column velocity model used was derived from a 1D average of conductivity-temperature-

180 depth (CTD) measurements in the vicinity of the CRR (Banyte et al., 2018a,b) (Figure 3). This 1D

181 velocity model extends to a water depth of ~3.12 km, and was then linearly extrapolated to a

182 maximum depth of 3.5 km. The final 3D localization model consists in a grid of 60 x 60 x 3.5 km

183 with node spacings of 0.1 km in all dimensions.



FIG. 3. Sound velocity profile in the water column calculated by averaging CTD measurements
obtained in the vicinity of the CRR (*Banyte et al., 2018a, 2018b*). This 1D profile was used to
construct the 3D localization velocity model.

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190 The calculated direct arrival travel-times may not correspond to the arrivals recorded by each 191 instrument due to strong bathymetric changes obstructing the direct wavepath, leading to multi-192 pathing (Dréo et al., 2019). This is especially the case for large source-receiver distances. In our case, 193 however, most of the OBSs recording the whale calls are located less than 10-15 km from the 194 whale's position. In addition, travel-time picks associated with large travel-time residuals are down-195 weighted during the localization procedure. This increases the robustness of the localization 196 procedure, focusing on high-quality travel-time picks closer to the source location. Time pick 197 uncertainties are also automatically attributed to each travel-time pick depending on the signal-to-198 noise ratio. After selecting events with a minimum of four phase picks, the localization is finally 199 determined using the equal-differential time likelihood function and the oct-tree importance

sampling method (*Lomax et al., 2009*). In the present case, including travel-time residuals as station
corrections to mitigate the effects of potentially unaccounted for velocity changes and ray paths, did
not improve the localizations. Spatial uncertainties were estimated using the dimensions of the 68%
confidence ellipsoid fitting the probability density function of each event localization (*Lomax et al., 2009*).

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2. Double-difference relative localizations

207 Using the travel-time picks and the absolute localizations obtained with NonLinLoc, we then 208 relocalize the events using the double-difference method implemented in hypoDD (Waldhauser & 209 Ellsworth, 2000), which assumes that similar events recorded by the same seismic station share similar 210 source mechanisms and wave propagation paths from their sources to the receiver(s). Such methods 211 have been successfully applied to the relocalization of Fin whale calls (Wilcock, 2012) and acoustic 212 sources (Tenorio-Hallé et al., 2017). The double-difference method particularly, which was developed 213 for clusters of similar events, assumes that small observed travel-time pick differences between 214 events at a station are due to small differences in their localizations, and not due to unmapped 215 velocity changes along their propagation paths or in their hypocentral area. This method also 216 benefits from the calculation of high-precision differential travel-time picks based on cross-217 correlating the event waveforms. In the present case, all observed calls were highly similar indicating 218 a similar source and little impact of the wave propagation on the frequency content of the observed 219 signals. As the travel-time picks had already been obtained using cross-correlation between a high 220 signal-to-noise ratio version of the signal (i.e., the event stack) and event waveforms, we used those 221 picks to calculate the differential timings and did not repeat the event waveform cross-correlation.

With the double-difference method, instead of localizing the events based on their travel-times,
events were relocalized relative to each other by calculating the difference in relative timings at the
stations (*Waldhauser & Ellsworth, 2000*), as

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$$dr_{k}^{ij} = (T_{k}^{i} - T_{k}^{j})^{obs} - (T_{k}^{i} - T_{k}^{j})^{calc},$$

where i and j are the indexes of two events, recorded at station k, and obs and calc refer to travel-times T observed and calculated, respectively. For events that are closely located with respect to variations in the velocity structure, these differences in relative timings can be related to changes in event locations as

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$$dr_{k}^{ij} = \frac{\partial T_{k}^{i}}{\partial \mathbf{m}} \Delta \mathbf{m}^{i} - \frac{\partial T_{k}^{j}}{\partial \mathbf{m}} \Delta \mathbf{m}^{j}$$

where **m** include the model parameters (x, y, z, and T_0 the origin time), $\frac{\partial T}{\partial \mathbf{m}}$ corresponds to slowness in the event region, and $\Delta \mathbf{m}$ corresponds to changes in event parameters. Finding the final set of model parameters **m**, which correspond to the final event locations, is an ill-posed problem due to the presence of errors (i.e., timings and initial locations), poorly linked events and outliers. This problem was, thus, stabilized using regularization, and solved using weighted, damped leastsquares as

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$$\left\| \mathbf{W} \begin{bmatrix} \mathbf{G} \\ \lambda \mathbf{I} \end{bmatrix} \mathbf{m} - \mathbf{W} \begin{bmatrix} \mathbf{d} \\ 0 \end{bmatrix} \right\|_{2} = 0$$

238 where *a priori* weights are included in **W**, **G** contains the slowness, the data vector **d** contains the 239 double differences, and λ is the regularization parameter.

The *a priori* weights represent the picking quality, corresponding to either the cross-correlation coefficients obtained during automatic picking or 1 for manual picks. The calls were first clustered using a maximum inter-event distance of 10 km and a minimum of six links with other calls. The inverse problem was then solved iteratively using the conjugate gradients method. We used two sets of 15 iterations, sufficient to converge to the final model parameters; the first set having no data re-weighting and the second using data re-weighting to reject and down-weight data associated with

246 large residuals (dynamic cutoff value of 6).

247 To work with hypoDD, the source configuration has to be adapted as all the sensors were

248 located deeper than the anticipated positions of the sources (whales) and the VA hydrophones were

249 located above the OBS hydrophones. Our solution was to flip the source–receiver configuration

250 upside down, setting the deepest station at 0 m and the others as if they were located in boreholes.

251 In this configuration, hypoDD uses a constant velocity to relocalize the events, and so we use the

average of the 1D water column velocity (1.5 km/s) shown in Figure 3.

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254 III. RESULTS

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A. Data and signal characteristics

256 We first calculated average Welch power spectral densities (PSD) to measure the average noise 257 levels over the entire recording period of the VA and its two nearest OBSs (Figure 4). Each 258 hydrophone's data were first converted to units of Pascal. For frequencies between approximately 1 259 and 30 Hz, noise levels are comparable between the OBSs and VA hydrophones, although variations 260 are observed between the hydrophones ($\sim \pm 5$ dB), with the OBS hydrophones generally being the 261 noisiest. OBS and VA hydrophones close to the seafloor also exhibit several spectral peaks between 262 1 and 5 Hz, the one at 1.8 Hz having the highest amplitude. Between 10 and 18 Hz, the VA 263 hydrophone closest to the seafloor (VA-1) has a higher noise level than all the other hydrophones 264 (+5 dB). For frequencies higher than 30 Hz, noise levels for VA-1 and the OBS hydrophones are 265 higher. VA-1 has a noise level similar to that of the OBS hydrophones at up to 60 Hz, approximately 266 10 dB higher than the other hydrophones, and then exhibits a lower average noise level of just a few

267 dB higher than the other hydrophones. This seems to indicate that environmental conditions close 268 to the seafloor contribute more to the instrument background noise, and that hydrophones located higher up in the water column do not experience higher noise levels induced by deep ocean currents 269 270 for example. Over the deployment period, average noise levels are higher at the beginning and end 271 (Figure 5). This is likely due to the presence of the research vessel either deploying instruments (25th and 26th January), carrying out the seismic survey (26th January to 1st February, 12th, 13th and 15th 272 February) or recovering the instruments (1st and 2nd February, 16th and 17th February). 273

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276 FIG. 4. (Color online). Welch averaged power spectral densities (PSDs) for the whole recording 277 period, for the hydrophones of the two OBSs (NG-17 and NG-24 - Figure 1) closest to the vertical array (VA) and the twelve channels of the VA (in dB re 1 Pa^2/Hz). The data used to calculate the 278 279

PSDs are in Pascal and unfiltered. We used segment lengths of 300 s with a Hamming window and

280 50% overlap.



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FIG. 5. (Color online). Welch averaged PSDs vs date in 2015 (in dB re 1 Pa²/Hz) for the
hydrophones of OBS NG-17 and NG-24, and VA-1 and VA-12. Over the deployment period,
average noise levels are higher at the beginning and end and are thought to be related to the seismic
survey (26th January to 1st February, 12th, 13th and 15th February), and deployment (25th and 26th
January) and recovery (1st and 2nd February, 16th and 17th February) operations of instruments within
the grid.

Whale calls recorded by the VA and OBS hydrophones, although located less than a few kilometers away, display significantly different signals after ~1 s of the first arrival (Figure 2). These differences are most likely due to significantly different wave propagation paths and interferences even for these closely located instruments. For example, the call reflection from the seabed will be part of the received wavelet for the OBSs on the seafloor but will interfere with the direct arrival through the water column at different lag times for the VA hydrophones. Figure 6 shows the timefrequency representation of whale calls with high signal-to-noise ratio using the short-time Fourier 297 transform (e.g., Tary et al., 2014). The total signal length is ~4-5 s, and comprises two main parts. The first part starts with a low-amplitude "snout" of ~ 0.5 -1 s, which is only visible for high signal-298 299 to-noise ratio signals, followed by a higher-amplitude monochromatic wave packet of ~ 1 s duration. 300 The frequency of this first part is \sim 35.7 Hz, together with its overtone at \sim 71.5 Hz, and the snout 301 potentially also has a frequency closer to 40 Hz (Figure 6). The second part of the signal has large 302 variation in amplitude (i.e., below noise level through to higher amplitude than the first part). Its 303 duration and frequency are roughly 2-3 s and ~20 Hz, respectively. For some of the high-amplitude 304 signals, a high-frequency content up to approximately 150 Hz is also observed in addition to the 305 other two main frequency components (Figure 7). Its magnitude is, however, much lower than those 306 of the main signal components. For these main components, we also observe low-amplitude trails 307 after the main signal (e.g., Figure 7), lasting up to 50 s, likely corresponding to wave multi-pathing 308 within the water column and seafloor.





FIG. 6. (Color online). The same whale call, band-pass filtered between 10 and 45 Hz, recorded bythe hydrophones of OBS NG-25 (left) and NG-16 (right, see Figure 2) and their time-frequency

representations calculated using the short-time Fourier transform (Hann window of 0.5 s with 90%overlap).

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316 The whale calls are observed in sequences with variable periodicities ranging from tens of 317 seconds to 5-10 min (e.g., Figures 7 and 8), repeated multiple times during the whole survey. At a 318 single instrument, different whale calls can display differences in amplitude reflecting differences in 319 signal strength emitted by the source. However, larger differences are observed between close and 320 distant whale calls, with higher-amplitude calls corresponding to those located close to the 321 instrument and lower-amplitude calls corresponding to answers originating further away. This was 322 verified as several vocal whales were located at different positions within the instrument network 323 during the same time frame (Figure 8). While most of the signals recorded during the survey are 324 similar to those shown in Figures 2 and 6, other signals of higher frequency are also observed 325 (Figure 7 and 9). These are generally clumped in time and with frequency ranging from 80 to 160 326 Hz. Individually, they are of low amplitude, have a signal-to-noise ratio lower than 3, and a duration 327 of ~ 0.5 -1 s. Despite their lower amplitude, they are still simultaneously recorded by nearby OBSs 328 and some of the VA hydrophones.



FIG. 7. (Color online). Continuous data from the hydrophone of OBS NG-24 (Figure 1) starting
at 00:43:20 (top) and 02:30:00 (bottom) on the 14th February 2015. The time series data are bandpass filtered between 5 and 240 Hz, and the time-frequency representations are calculated using the
short-time Fourier transform (Hann window of ~4.1 s with 90% overlap).



FIG. 8. (Color online). Continuous data from the hydrophones of OBS NG-25 (top) and NG-8
(bottom) (Figure 1) starting at 14:50:50 on the 29th January 2015, band-pass filtered between 5 and
100 Hz. Their time-frequency representations are calculated using the short-time Fourier transform
(Hann window of ~4.1 s with 90% overlap). At least three whales are acoustically active (W1, W2
and W3).



FIG. 9. (Color online). High-frequency whale calls, band-pass filtered between 5 and 240 Hz
(top plot), recorded by the hydrophone of OBS NG-24 (zoom-in of Figure 7 between 11 min 50 s
and 12 min 13 s) and their time-frequency representation calculated using the short-time Fourier
transform (bottom plot, Hann window of ~1 s with 90% overlap).

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B. Detection and localization results

350 Carrying out the detection using the STA/LTA on the vertical geophone component OBS data, 351 we obtained 1177 detections and, from those, manually selected 188 events corresponding to whale 352 calls. The other detections mainly correspond to other coherent signals such as seismic shots and seismicity. By far the majority of these calls were observed on the 29th January, the others being 353 354 observed during seven other days throughout the whole survey (Figure 10). We then picked the 355 same arrivals using the hydrophone data for both OBSs and VA, resulting in a total of 2134 P-wave 356 picks from the different instruments that recorded the calls. We encountered additional difficulties 357 in manually picking these calls as they share very similar waveform characteristics, whatever their 358 position and distance from the instrument (i.e., similar source and propagation material). Often,

359 more than one arrival moveout at a time is observed, and the emergent low-amplitude snout could 360 be below the noise level. Out of the 188 events, 183 events were localized with average horizontal 361 error, vertical error and RMS travel-time residual of ± 2.1 km, ± 0.7 km and 0.2 s, respectively (Figure 11). Focusing on the localizations from the 29th January, at least three animals or groups of 362 363 animals are present simultaneously (see also Figure 8); one stays in the northeast corner of the 364 network, while another passes through the southern half of the network from west to east, and the 365 last moves through the southeast corner of the network to join up with the second at the eastern 366 edge of the network.

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369 FIG. 10. Temporal distribution of the number of whale calls detected using the STA/LTA (black

370 bars) and the network empirical subspace detector (white bars).



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FIG. 11. (Color online). Whale call localizations, laterally and with depth, for the events detected using the STA/LTA and the empirical subspace detector, and localized using NonLinLoc (NLL) and hypoDD (hDD). Only events with horizontal uncertainties lower than ±5 km are shown. The resulting number of events is annotated for each method combination. The dot colors indicate the date after 26th January 2015. Dashed line and black triangles show the location of the CRR and the instruments (OBSs and VA), respectively. Arrows indicate whale trajectories inferred from call timings.

381	For the empirical subspace detector, we used signals with high signal-to-noise ratio recorded by
382	OBS NG-9 to calculate the stack and its first derivative. Applying the detector to the continuous
383	data we obtain 3100 events for the entire period. We then manually reviewed these detections and
384	selected the 1891 events corresponding to whale calls, including 113 events also detected by the
385	STA/LTA approach. This represents approximately 10 times more events selected using the
386	empirical subspace detector than the STA/LTA. The temporal distribution of the selected events
387	shows that whales are present during the whole survey (Figure 10), with some periods either with
388	more individuals or of higher vocal activity (e.g., 29th–30th January, 2nd–8th February, 9th–11th
389	February, 13 th –15 th February). Using the high similarity between events, we then automatically
390	picked the newly detected events with the stack. These automatic picks were then manually reviewed
391	resulting in 7156 picks. Out of the 1891 events, 938 events were localized with average horizontal
392	error, vertical error and RMS travel-time residual of ± 8.7 km, ± 0.9 km and 0.1 s, respectively
393	(Figure 11). Most of the events localized only have a low number of travel-time picks, which lead to
394	larger horizontal uncertainties in comparison with the events detected by the STA/LTA. Focusing
395	on the events localized with uncertainties smaller than ± 5 km, these event localizations are similar to
396	those obtained using the events detected with the STA/LTA. The main difference is in the inclusion
397	of more events for days either side of the 29th January. For example, whales are detected on 26th
398	January to the west and in the south half of the network, and on the 2 nd February and between the
399	13 th and 15 th February mainly in the southwest part of the network. On the 2 nd February, an
400	individual or a single group is moving through the network from the southwest to the northeast,
401	ceasing their vocalizations when they reach the center of the network. On the 26th January and
402	between the 13th and 15th February, however, the call localizations are gathered at specific locations.

403 We then relocalized the events detected by the STA/LTA and the empirical subspace detector 404 using the double-difference method. Most calls were sufficiently linked to other calls to be 405 relocalized due to the high number of instruments deployed and their proximity to the whales. The 406 resulting event relocalizations show that, due to the high linkage between events, more events are 407 retained (i.e., with NonLinLoc only 291 events have uncertainties smaller than ± 5 km, Figure 11). In 408 the case of the original events detected by the STA/LTA, 155 events are relocalized with a final 409 average RMS difference time residual of 0.05 s, and preserving ~84% of the differential times in the 410 final relocalizations. Relocalizing all events led to 419 events being relocalized and a final average 411 RMS difference time residual of 0.09 s, and preserving ~45% of the differential times in the final 412 relocalizations. The relocalized events were better clustered both laterally and in depth. Indeed, 413 while some event scatter remains in depth, most events are relocalized at depths shallower than 1 414 km. On the other hand, some scattered events outside the network couldn't be associated to a 415 cluster and were not kept during the relocalization procedure. The trajectories of each whale, or groups of whales, during 29th January can be determined 416 417 (Figure 11). From south to north, estimations of the whale swimming speed for these three main

418 whale tracks are 2.8, 5.0 and 3.1 km/hr (see supplementary material for an animation of whale call

419 localizations during 29th January). In addition, events from 26th January and 2nd February, and at the

420 end of the survey (13th-16th February), are mostly localized in the southwestern part of the network.

- 421 Those of the 26^{th} January and 2^{nd} February also show a southwest-northeast orientation, which is not
- 422 the case for those from the 13^{th} - 16^{th} February.
- 423

424 IV. DISCUSSION

425

5 A. Signal characteristics and whale identification

- 426 The main characteristics of the low frequency whale calls we observe during this survey can be
- 427 summarized as follows:
- 428 i) calls have a total duration of 4-5 s with two main parts;
- 429 ii) these two main parts have a duration and monochromatic frequency content of \sim 1-2 s and
- 430 \sim 36 Hz and \sim 2-3 s and \sim 20 Hz, respectively; and
- 431 iii) the calls repeatedly occur over long periods without a precise periodicity.

432 These characteristics fall within the general range of characteristics for pulsed calls generated by

433 whales (e.g., Richardson et al., 2013; McDonald et al., 2009). Different baleen whale species inhabit the

434 Eastern Tropical Pacific Ocean including Blue whales, Humpback whales, Bryde's whales, Fin

435 whales, and Minke whales (Wade & Gerrodette, 1993; Palacios et al., 2012; Palacios et al., 2023). Low

436 frequency sounds such as those observed in this study could potentially be associated with some of

437 these baleen whales, such as Blue whales, Humpback whales, Bryde's whales and Fin whales

438 (*Richardson et al., 2013*; *Oleson et al., 2003*).

439 Similar signals and time-frequency characteristics are reported for Bryde's whales in the 440 Eastern Tropical Pacific Ocean (signals Be1 and Be6 of Oleson et al., 2003). Be1 signals show two 441 parts with average frequencies of 21.2 Hz and 36.6 Hz, and an average total duration of 2.7 s. Be6 442 signals are also very similar to the higher frequency signals observed (e.g., Figures 7 and 9). Other 443 signals attributed to Bryde's whales were reported by Oleson et al. (2003), but they were not observed 444 during our survey. Sightings of Bryde's whales in this region are common even though the 445 abundance is low (Wade & Gerrodette, 1993; Palacios et al., 2012). Bryde's whales are found all year 446 round within warm and temperate oceans, such as the Eastern Tropical Pacific Ocean, some also 447 migrate within these waters but their actual movements are still largely unknown (Constantine et al.,

448 2018). We conclude that the most likely origin of the observed signals are Bryde's whales dwelling in449 this region.

450 Oleson et al. (2003) suggested that Be1 calls correspond to counter-calling between different 451 whales located relatively close to each other. Our observations with calls alternating between 452 different localizations are consistent with the view that this call is used by the whales to briefly 453 communicate with each other. Moreover, on the 29th January, three whales were simultaneously 454 localized within the network footprint. All the whales show a general trajectory from the southwest 455 to the northeast, with the whale to the far northeast of the network also zigzagging, possibly waiting 456 for the other two individuals. The other two whales finally meet at the eastern edge of the network. 457 The calls could then correspond to the different whales communicating their general location to the 458 other individuals. Even though the calls continue after the two whales met, they are situated outside 459 the network and hence cannot be as precisely localized. Examining the final localizations presented 460 in Figure 11, at least seven Bryde's whales passed through the network at different times, although 461 some whales might have passed through the network more than once. These movements should not 462 be associated with a particular migration, as there is no evidence to support Bryde's whale seasonal 463 migration (Kato & Perrin, 2009), but instead a year-round presence is noted in the areas where they 464 have been observed (Heimlich et al., 2005; McDonald, 2006; Širović et al., 2014). They might however 465 expand their habitat range seasonally (Kerosky et al., 2012; Ferreira et al., 2021).

The higher frequency Be6 signal is generally of lower amplitude and is hence recorded only by nearby OBSs, which suggests that it only propagates over shorter distances. This seems to support the suggestion of *Oleson et al. (2003)*, that these signals were produced by closely spaced animals while heading in a specific direction. The examples shown in Figures 7 and 9 correspond to calls produced on the 14th February (red dots in Figure 11), and their different amplitudes observed at nearby instruments seems to indicate they were generated by more than one animal. On this day,

111 low-frequency calls were localized, mostly in a cluster centered around the position of OBS
NG-24 (Figure 1). Their spatial distribution as a compact cluster might, however, be a consequence
of the limited number of OBSs remaining on the seabed at this time (five OBSs: 4, 7, 14, 17 and 24
- Figure 1).

- 476
- 477 B. Methodological aspects

478 Whale calls are generally detected in the time-frequency domain owing to their stable, 479 recognizable time-frequency features (e.g., Mellinger & Clark, 2000; Munger et al., 2005; Baumgartner & 480 Mussoline, 2011). Such detectors can use pre-defined time-frequency templates to target particular 481 signals or scan spectrogram portions for any high-magnitude features. Matched filtering techniques 482 do not take into account potential variations due to changes in time-frequency features (i.e., over 483 time and/or across a population), background noise, or site-specific noise levels for example. 484 General time-frequency and time-domain detectors (e.g., the STA/LTA) are more flexible at the 485 expense of their sensitivity, and require additional steps to select and classify the signals (Baumgartner 486 & Mussoline, 2011). For the STA/LTA, high amplitude noise, even outside the frequency band of 487 interest can mask the signals to be detected. On the other hand, subspace detectors take into 488 account some signal variability embedded within the original signal dataset and translate it to the 489 vector basis used to perform the detection (Harris, 2006). The selectivity of this vector basis allows 490 the detection of signals with lower signal-to-noise ratio compared to other general time-domain 491 detectors. Different factors determine the final performance of this technique such as using a 492 representative set of signals to build the vector basis, using an appropriate number of singular 493 vectors, and adequately setting the detection threshold. Further studies would be needed to compare 494 the performance of time-domain subspace detectors with those of frequency-domain template 495 matching detectors.

496 Even though we used the STA/LTA approach on OBS geophone data, we managed to detect 497 high-amplitude whale calls. However, using a larger number of OBS hydrophones and the VA 498 hydrophones as well as the empirical subspace detector, we obtained roughly 10 times more events. 499 This shows the benefit of both using hydrophones and subspace detection techniques to study 500 whale vocalizations. The observation of whale calls during the whole recording period shows that 501 they are continuously present at one position or another (Figure 10). The number of detections 502 obtained using the STA/LTA and the empirical subspace detector also strongly depends on other 503 factors such as the presence of seismic shots and/or intense noise episodes in the data (e.g., on the 504 2nd February), and the number of stations available at different times. Most of the OBSs and the VA 505 are deployed until the 2nd February whereas, after that date and until the end of the survey, only five 506 OBSs and the VA remain deployed in the network. Nevertheless, some calls were detected during 507 the whole survey.

508 Despite the high number of OBSs at the beginning of the survey, the average horizontal uncertainty for the calls detected using the STA/LTA and the empirical subspace detector remain 509 510 relatively high (i.e., at 2.1 and 8.7 km, respectively). This likely mainly arises from the challenges of 511 the travel-time picking discussed earlier. This could also indicate discrepancies between the observed 512 travel-times and the theoretical arrival times of the first P-wave calculated using the eikonal equation 513 (Lomax et al., 2009). In the case of strong bathymetry variations, such as those encountered around 514 mid-ocean ridges, ray paths between whales and instruments on the seafloor might not be direct but 515 involve reflections from the seabed or within the water column (Dréo et al., 2019). This can be 516 particularly important for longer propagation paths. Call localizations in depth are well-constrained 517 for calls within the instrument network until the partial instrument recovery on the 1st February, 518 2015. However, calls localized deeper than 0.5-1 km are systematically associated with larger vertical 519 uncertainties (e.g., 0.5-1 km) and generally a lower number of time picks as well. For calls occurring

after the 2nd February, their uncertainty increases due to the lower number of instruments remaining
deployed and their linear deployment configuration. This can be counterbalanced for some calls
using the double-difference method when event clustering is high and, hence, due to the high
connectivity between nearby events. Ultimately, although localization error remains, our approach
has resulted in a general improvement in ability to reconstruct the general trend of the whale
movement trajectories, estimate of their number, and interpret of their behavior.

526

527 V. CONCLUSION

528 In this study we analyze and compare continuous data acquired by 25 OBSs and a VA of 12 529 hydrophones deployed around the CRR, Eastern Tropical Pacific Ocean, in January and February 530 2015. The noise levels observed for the hydrophones closer to the seafloor are on average 10 dB 531 higher than for those situated higher up in the water column. Calls and seismic events are generally 532 localized using instruments either close to the sea surface or at the seafloor. The VA provides a 533 larger aperture in depth that could, in theory, improve the localization accuracy in this dimension. In 534 this study, however, depth uncertainties remained high (e.g., $\sim \pm 0.8$ km) even for calls including 535 travel-time picks from the VA data.

We observe two different signal types produced by whales during this deployment. The first
is present during the whole deployment with different amplitudes and recurrences. It displays two
parts with consistent frequency content; the first is a 1-2 s long signal of high amplitude and
frequency ~36 Hz, and the second is a 2-3 s long signal of generally lower amplitude and frequency
~20 Hz. The second type of signal is a 0.5-1.0 s long signal of low amplitude and higher frequency
content ranging between 80 and 160 Hz. Both types of signals are most likely made by Bryde's
whales, a whale species thought to inhabit and is occasionally sighted in this region. The first signal

543 type seems to correspond to whales counter-calling between each other to provide their

544 approximate location, and eventually meeting up to form a group. The second could not be

545 associated with specific conditions or type of behavior.

546 Using highly similar whale calls previously detected using the STA/LTA approach, we employed 547 the network empirical subspace detector and obtained approximately 10 times more events (i.e., 188 548 vs 1891, respectively). These events were then absolutely localized and finally relocalized using the 549 double-difference technique. The combination of the subspace detector with the double-difference 550 technique results in an increase in events localized and a better delineation of the whale trajectory. 551 This is particularly effective for highly similar events observed by our dense instrument network as 552 that leads to highly linked events. These precise relocalizations show that at least seven (group of) 553 whales passed through the network at different times, most of them from the southwest to the 554 northeast (i.e., from an area closer to the Galapagos Islands towards the southern shores of 555 Panama). At the end of the survey period, one or more whales stayed within the area for up to a few 556 days. This could, however, be a localization artifact due to the limited number of instruments 557 remaining deployed at the end of the survey and their generally, north-south linear orientation. 558 Regardless, this study illustrates the value of applying advanced detection and localization methods 559 in the passive monitoring of baleen whales, particularly their ways of communicating and ecology, 560 which is crucial to their identification, conservation and the mitigation of the potential impacts of 561 human activities.

562

563 SUPPLEMENTARY MATERIAL

See supplementary material at [URL] for an animation of the final whale call localizations
obtained on the 29th January, 2015, between 2pm and 10pm.

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584

585 DATA AVAILABILITY

586 Data from cruises JC113, JC114 and SO238 are archived at the NERC's British Oceanographic Data

587 Centre and are available from the following links:

588 JC113 – www.bodc.ac.uk/resources/inventories/cruise_inventory/report/15035/,

589 JC114 – www.bodc.ac.uk/resources/inventories/cruise_inventory/report/15036/, and

590 SO238 – www.bodc.ac.uk/resources/inventories/cruise_inventory/report/15045/.

592 **REFERENCES**

- 593 Banyte, D., Morales Maqueda, M.A., Smeed, D.A., Hobbs, R., Megann, A., and Recalde, S. (2018a).
- 594 "Geothermal heating in the Panama Basin: 1. Hydrography of the basin", J. Geophys. Res.
 595 Oceans 123(10), 7382-7392.
- 596 Banyte, D., Morales Maqueda, M.A., Smeed, D.A., Megann, A., Hobbs, R., and Recalde, S. (2018b).
 597 "Geothermal heating in the Panama Basin. Part II: Abyssal water mass transformation", J.
 598 Geophys. Res. Oceans 123(10), 7393-7406.
- **599** Barrett, S.A., and Beroza, G.C. (**2014**). "An empirical approach to subspace detection", Seismol Res.
- **600** Lett. **85**(3), 594-600.
- Baumgartner, M. F., and Mussoline, S. E. (2011). "A generalized baleen whale call detection and
 classification system", The Journal of the Acoustical Society of America 129(5), 2889-2902.
- Brodie, D. C., & Dunn, R. A. (2015). "Low frequency baleen whale calls detected on ocean-bottom
 seismometers in the Lau basin, southwest Pacific Ocean", The Journal of the Acoustical
 Society of America 137(1), 53-62.
- 606 Constantine, R., Iwata, T., Nieukirk, S.L., and Penry, G.S. (2018). "Future directions in research on
 607 Bryde's whales", Frontiers in Marine Science 5, 333.
- 608 De La Hoz, C., Tary, J. B., and Lomax, A. (2021). "Empirical subspace detection applied to triggered
 609 seismicity by the July 25, 2011, M_w 5.0 earthquake in the Sea of Marmara, Turkey",
- **610** Computers & Geosciences **152**, 104738.
- 611 Dréo, R., Bouffaut, L., Leroy, E., Barruol, G., and Samaran, F. (2019). "Baleen whale distribution
- and seasonal occurrence revealed by an ocean bottom seismometer network in the Western
- 613 Indian Ocean", Deep Sea Research Part II: Topical Studies in Oceanography 161, 132-144.

614	Dunn, R. A., and Hernandez, O. (2009). "Tracking blue whales in the eastern tropical Pacific with an
615	ocean-bottom seismometer and hydrophone array", The Journal of the Acoustical Society of
616	America 126 (3), 1084-1094.

- 617 Ferreira, R., Dinis, A., Badenas, A., Sambolino, A., Marrero-Pérez, J., Crespo, A., and Alves, F.
- 618 (2021). "Bryde's whales in the North-East Atlantic: New insights on site fidelity and
- 619 connectivity between oceanic archipelagos", Aquatic Conservation: Marine and Freshwater
 620 Ecosystems 31(10), 2938-2950.
- Frankel, A.S. (2009). Sound production. In "Encyclopedia of marine mammals" (Academic Press,
 Cambridge), pp. 1056-1071.
- Funnell, M.J., Robinson, A.H., Hobbs, R.W. and Peirce, C. (2021). "Evolution and properties of
 young oceanic crust: constraints from Poisson's ratio", Geophysical Journal International
 225(3), 1874-1896.
- Harris, D. (2006). Subspace Detectors: Theory, Lawrence Livermore Natl. Lab. Rep. UCRL- TR222758. Lawrence Livermore National Laboratory, Livermore, California.
- 628 Heimlich, S. L., Mellinger, D. K., Nieukirk, S. L., and Fox, C. G. (2005). "Types, distribution, and
- seasonal occurrence of sounds attributed to Bryde's whales (Balaenoptera edeni) recorded in
- 630 the eastern tropical Pacific, 1999–2001", The Journal of the Acoustical Society of
- 631 America 118(3), 1830-1837.
- 632 Gregory, E.P.M. (2018). The seismic characterisation of layer 2 in oceanic crust around ODP
- 633 borehole 504B (PhD thesis). University of Durham, UK.
- Hatch, L.T., Clark, C.W., Van Parijs, S.M., Frankel, A.S., and Ponirakis, D.W. (2012). "Quantifying
 loss of acoustic communication space for right whales in and around a US National Marine
 Sanctuary", Conservation Biology 26(6), 983-994.
 - 34

- 637 Hobbs, R., and Peirce, C. (2015). RRS James Cook JC114 Cruise Report. Online Report.
- 638 <u>https://www.bodc.ac.uk/resources/inventories/cruise_inventory/reports/jc114.pdf.</u>
- Kato, H., and Perrin, W. F. (2009). "Bryde's whales: Balaenoptera edeni/brydei", in Encyclopedia of
 marine mammals (pp. 158-163). Academic Press.
- Kavanagh, A. S., Owen, K., Williamson, M. J., Blomberg, S. P., Noad, M. J., Goldizen, A. W., ... and
 Dunlop, R. A. (2017). "Evidence for the functions of surface-active behaviors in humpback
- 643 whales (Megaptera novaeangliae)", Marine Mammal Science 33(1), 313-334.
- 644 Kerosky, S. M., Širović, A., Roche, L. K., Baumann-Pickering, S., Wiggins, S. M., and Hildebrand, J.
- 645 A. (2012). "Bryde's whale seasonal range expansion and increasing presence in the Southern
- 646 California Bight from 2000 to 2010", Deep Sea Research Part I: Oceanographic Research
 647 Papers 65, 125-132.
- Lomax, A., Zollo, A., Capuano, P., and Virieux, J. (2001). "Precise, absolute earthquake location
 under Somma-Vesuvius volcano using a new three-dimensional velocity model", Geophys. J.
 Int. 146(2), 313–331.
- Lomax, A., Michelini, A., and Curtis, A. (2009). Earthquake Location, Direct, Global-Search
 Methods, in Encyclopedia of Complexity and System Science, vol. 5. Springer, New York,
 pp. 2449-2473.
- 654 Lowell, R.P., Zhang, L., Maqueda, M.A.M., Banyte, D., Tong, V.C.H., et al. (2020). "Magma-
- hydrothermal interactions at the Costa Rica Rift from data collected in 1994 and 2015",
 Earth Planet. Sci. Lett. 531, 115991.
- Matias, L., and Harris, D. (2015). "A single-station method for the detection, classification and
 location of fin whale calls using ocean-bottom seismic stations", The Journal of the
 Acoustical Society of America 138(1), 504-520.

660	Mellinger, D. K., and Clark, C. W. (2000). "Recognizing transient low-frequency whale sounds by
661	spectrogram correlation", The Journal of the Acoustical Society of America 107(6), 3518-
662	3529.
663	McDonald, M. A., Hildebrand, J. A., and Webb, S. C. (1995). "Blue and fin whales observed on a
664	seafloor array in the Northeast Pacific", The Journal of the Acoustical Society of
665	America 98 (2), 712-721.
666	McDonald, M.A., Hildebrand, J.A., and Mesnick, S. (2009). "Worldwide decline in tonal frequencies
667	of blue whale songs", Endangered species research 9(1), 13-21.
668	McDonald, M.A., Mesnick, S.L., and Hildebrand, J.A. (2006). "Biogeographic characterization of
669	blue whale song worldwide: using song to identify populations", Journal of cetacean research
670	and management 8 (1), 55-65.
671	Minshull, T.A., Sinha, M.C., and Peirce, C. (2005). "Multi-disciplinary, sub-seabed geophysical
672	imaging", Sea Technol. 46 , 27-31.
673	Munger, L. M., Mellinger, D. K., Wiggins, S. M., Moore, S. E., and Hildebrand, J. A. (2005).
674	"Performance of spectrogram cross-correlation in detecting right whale calls in long-term
675	recordings from the Bering Sea", Canadian Acoustics 33(2), 25-34.
676	Oleson, E.M., Barlow, J., Gordon, J., Rankin, S., and Hildebrand, J.A. (2003). "Low frequency calls
677	of Bryde's whales", Marine Mammal Science 19(2), 407-412.
678	Palacios, D.M., Herrera, J.C., Gerrodette, T., Garcia, C., Soler, G.A., Avila, I.C., Bessudo, S.,
679	Hernandez, E., Trujillo, F., Forez-Gonzalez, L., and Kerr, I. (2012). "Cetacean distribution
680	and relative abundance in Colombia's Pacific EEZ from survey cruises and platforms of
681	opportunity", J. Cetacean Res. Manage. 12(1), 45-60.
682	Palacios, D., Felix, F., Montecinos, Y., Najera, E., Kelez, S., Samaniego, J., Velasquez, P., Zapata, L.,
683	Lancaster, M., Friedlaender, A.S., Castro, C.A., Quintana, E., Bermudez-Rivas, C., Cazas, JJ.,

684	Villamil Echeverri, C., Rojas Bracho, L., Sepulveda, M., Medrano-Gonzales, L., Santillan, L.,
685	Aguilar Arakaki, R., Pineda, L., Reisinger, R., and Johnson, C.M. (2023). Blue Corridor of the
686	Eastern Pacific: Opportunities and Actions for the protection of Migratory Whales. A
687	Technical Report. World Wide Fund for Nature (WWF).
688	Peirce, C., Tedd, J. C., and Hobbs, R. W. (2023). "Structure and dynamics of the Ecuador Fracture
689	Zone, Panama Basin", Geophys. J. Int. 235(2), 1519-1540.
690	Richardson, W. J., Greene Jr, C. R., Malme, C. I., and Thomson, D. H. (2013). Marine mammals and
691	noise (Academic press, Cambridge).
692	Robinson, A.H., Zhang, L., Hobbs, R.W., Peirce, C., and Tong, V.C.H. (2020). "Magmatic and
693	tectonic segmentation of the intermediate-spreading Costa Rica Rift – a fine balance
694	between magma supply rate, faulting, and hydrothermal circulation", Geophys. J. Int. 222,
695	132-152.
696	Širović, A., Bassett, H. R., Johnson, S. C., Wiggins, S. M., and Hildebrand, J. A. (2014). "Bryde's
697	whale calls recorded in the Gulf of Mexico", Marine Mammal Science 30(1), 399-409.
698	Širović, A., Hildebrand, J.A., Wiggins, S.M., McDonald, M.A., Moore, S.E., and Thiele, D. (2004).
699	"Seasonality of blue and fin whale calls and the influence of sea ice in the Western Antarctic
700	Peninsula", Deep Sea Research Part II: Topical Studies in Oceanography 51(17-19), 2327-
701	2344.
702	Socheleau, FX., Leroy, E., Carvallo Pecci, A., Samaran, F., Bonnel, J., and Royer, JY. (2015).
703	"Automated detection of Antarctic blue whale calls", J. Acoust. Soc. Am. 138(5), 3105-3117.
704	Song, F., Warpinski, N.R., Toksöz, M.N., and Kuleli, H.S. (2014). "Full-waveform based
705	microseismic event detection and signal enhancement: an application of the subspace
706	approach", Geophys. Prospect. 62 (6), 1406-1431.

- Tary, J.B., Herrera, R.H., Han, J., and van der Baan, M. (2014). "Spectral estimation—What is new?
 What is next?", Reviews of Geophysics 52(4), 723-749.
- 709 Tary, J. B., Hobbs, R. W., Peirce, C., Lesmes, C. L., and Funnell, M. J. (2021). "Local rift and
- 710 intraplate seismicity reveal shallow crustal fluid-related activity and sub-crustal
- faulting", Earth and Planetary Science Letters **562**, 116857.
- 712 Tenorio-Hallé, L., Thode, A. M., Sarkar, J., Verlinden, C., Tippmann, J., Hodgkiss, W. S., and
- Kuperman, W. A. (2017). "A double-difference method for high-resolution acoustic tracking
 using a deep-water vertical array", The Journal of the Acoustical Society of America 142(6),
 3474-3485.
- 716 Vargas, C.A., Pulido, J.E., and Hobbs, R.W. (2018). "Thermal structure of the Panama Basin by
 717 analysis of seismic attenuation", Tectonophysics 730, 81–99.
- Wade, P.R., and Gerrodette, T. (1993)." Estimates of cetacean abundance and distribution in the
 eastern tropical Pacific", Report of the International Whaling Commission 43, 477-493.
- 720 Waldhauser, F., and Ellsworth, W.L. (2000). "A double-difference earthquake location algorithm:
- Method and application to the northern Hayward fault, California", Bulletin of the
 seismological society of America 90(6), 1353-1368.
- Wilcock, W. S. (2012). "Tracking fin whales in the northeast Pacific Ocean with a seafloor seismic
 network", The Journal of the Acoustical Society of America, 132(4), 2408-2419.
- 725 Wilcock, W. S., and Hilmo, R. S. (2021). "A method for tracking blue whales (Balaenoptera
- musculus) with a widely spaced network of ocean bottom seismometers", Plos one 16(12),
 e0260273.
- Wilson, D.J., Robinson, A.H., Hobbs, R.W., Peirce, C., and Funnell, M.J. (2019). "Does intermediate
 spreading-rate oceanic crust result from episodic transition between magmatic and magma-

- 730 dominated, faulting-enhanced spreading?–the Costa Rica Rift example", Geophys. J. Int.
- **218**, 1617-1641.



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