

1 **Spatiotemporal connectivity of noise-derived seismic body waves with ocean waves and microseism excitations**

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5 **Abstract**

6 Everywhere on the Earth planet is experiencing tiny but incessant microseisms induced by formidable natural forces,
7 particularly, storm-driven ocean waves. Microseism noise was deemed a nuisance, but now can be turned into signals
8 via the emerging correlation technique. We use a P -type phase recently derived from double-array noise correlations
9 to study the links between global oceanic microseism sources and noise-derived seismic signals. The phase is termed
10 P_{dmc} in this study. The two correlated seismic arrays locate in the north hemisphere, while the effective sources
11 responsible for the construction of P_{dmc} signals lie in the south hemisphere. The temporal variations of P_{dmc} amplitudes
12 are highly correlated with those of the power of effective sources. The P_{dmc} amplitudes are also correlated with other
13 ineffective southern sources and anti-correlated with the northern sources. We ascribe the correlation with the
14 ineffective southern sources to the spatiotemporal connection of the southern sources. The anti-correlation with
15 northern sources is explained by the reverse seasonal patterns of the southern and northern sources, and that the
16 northern sources have adverse impacts on the signal construction. A successful signal construction relies on the
17 competition between effective and ineffective sources, not just the power of effective sources. In our case, the ocean
18 waves in the effective source region are dominated by wind waves, whereas the excitation of microseisms is largely
19 owing to swells by causal storms in surrounding regions.

20

21 **1 Introduction**

22 As early as the birth of seismometers in the later 19th century, the incessant background vibrations of Earth had been
23 observed (Bernard, 1990; Dewey and Byerly, 1969; Ebeling, 2012). They were termed “microseisms” due to their
24 feebleness. With more apparatus deployed worldwide, it was soon recognized that microseisms are ubiquitous and
25 irrelevant to seismicity. The observation of microseisms aroused interests from various disciplines. Researchers linked
26 the generation of microseisms to atmosphere processes and ocean wave activities. Meteorologists tried to employ land
27 observations of microseisms to track remote oceanic storms (e.g., Harrison, 1924). To the mid-twentieth century, it
28 has been known well that microseisms are excited by storm-driven ocean waves. The most energetic microseisms that
29 dominate the seismic noise spectra, namely, the so-called secondary microseisms at seismic periods around 7 s
30 (Peterson, 1993), are excited by the nonlinear interactions between nearly equal-frequency ocean waves propagating
31 in nearly opposite directions (Hasselmann, 1963; Longuet-Higgins, 1950). The periods of the excited secondary
32 microseisms are half those of the colliding ocean waves. By coupling the excitation theory of secondary microseisms
33 proposed by Longuet-Higgins (1950) with the ocean wave action model, Kedar et al. (2008) modeled the secondary
34 microseism excitations in the north Atlantic, and validated the numerical modeling by comparing with inland
35 seismological observations. After, more authors simulated the oceanic microseism sources and some reported the
36 consistency between predictions and observations (e.g., Ardhuin et al., 2011, 2015; Hillers et al., 2012; Nishida and
37 Takagi, 2016; Stutzmann et al., 2012).

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39 The seismic excitation by an oceanic microseism source has nothing special compared to that by an earthquake, in
40 that the seismic wavefield recorded at any point is a convolution of the source time function with the Green function
41 of the propagating medium between source and receiver. The main difference lies in the source process. The burst of
42 earthquake leads to an impulsive source time function. Isolated seismic phases are generally distinguishable from the
43 seismograms. In contrast, the excitation of microseisms is an incessant random process, so that the convolution mixture
44 signals are not directly discernible. With array beamforming (Rost and Thomas, 2002) or correlation technique
45 (Campillo and Paul, 2003; Shapiro and Campillo, 2004), specific phases from distant microseism sources have been
46 identified from seismic noise records (e.g., Euler et al., 2014; Gerstoft et al., 2008; Landés et al., 2010; Liu et al., 2016;
47 Meschede et al., 2017; Zhang et al., 2010). The correlation technique is advantageous in that, by correlating the noise
48 records at two receivers, explicit seismic signals can be derived. Noise-derived surface waves have been used to infer
49 the azimuthal and seasonal changes of noise sources (e.g., Stehly et al., 2006). Noise-derived body waves can provide
50 better constrains in imaging the noise sources (Landés et al., 2010). Recently, deep body waves that pass through
51 mantle and core have been constructed from ambient noise (e.g., Boué et al., 2013; Lin et al., 2013; Nishida, 2013;
52 Poli et al., 2015; Spica et al., 2017; Xia et al., 2016). The noise-derived body waves are valuable for surveying the

53 deep structure and for understanding the links between seismological observations and atmospheric/oceanographic
54 phenomena.

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56 Hillers et al. (2012) made the first systematic comparison between microseism sources derived from seismological
57 observations and oceanographic modeling. The seismologically derived data (time resolution: 13 days; spatial
58 resolution: 2.5° latitude \times 5° longitude), are the global back-projections of near-zero-lag P signals generated from the
59 cross correlations of microseism P waves at seismic array (Landés et al., 2010). The modeled data (time resolution:
60 3 hours; spatial resolution: 1° latitude \times 1.25° longitude), are a global extension of the numerical simulation by Kedar
61 et al. (2008). The two datasets are resampled to common resolutions for comparison. For the seismologically derived
62 data, the back-projection is based on the relationship between the source-receiver distance and the horizontal slowness
63 of teleseismic P wave. However, seismic phases that have common slownesses (e.g., P and PP waves), cannot be
64 discriminated in this method (Gerstoft et al., 2008; Landés et al., 2010). Thus, the imaged sources are somewhat
65 ambiguous. For the modeled data, coastal reflections of ocean waves, that can play a role in the ocean wave-wave
66 interactions at near-coast regions (Ardhuin et al., 2011; Longuet-Higgins, 1950), are neglected. Due to the resonance
67 of seismic waves in the water columns, bathymetry can have significant effect on the excitation of microseisms (Hillers
68 et al., 2012; Kedar et al., 2008; Longuet-Higgins, 1950). The importance to account for the bathymetric effect on the
69 microseism P -wave excitations has been addressed in several studies (e.g., Euler et al., 2014; Meschede et al., 2017).
70 Hillers et al. (2012) considered the bathymetric effect, but using the amplification factors derived by Longuet-Higgins
71 (1950) for surface waves.

72
73 There have been progress in several aspects since the study by Hillers et al. (2012). Rasclé and Ardhuin (2013)
74 published an oceanographic hindcast database that includes global oceanic secondary microseism sources (3-hour time
75 resolution and 0.5° spatial resolution). Coastal reflections were accounted for in the modeling (Ardhuin et al., 2011).
76 Gualtieri et al. (2014) proposed the formulae to compute the bathymetric effect on microseism body waves. Li et al.
77 (2019) developed double-array methods that can estimate the respective slownesses of the interfering waves at two
78 correlated seismic arrays, and thereby, provide better constrains on the ray paths of the correlated seismic phases. The
79 double-array configuration eliminates the ambiguity in determining the source area responsible for the noise-derived
80 signals (referred to as effective source region hereafter). In this study, we integrate these new progresses to investigate
81 the associations of noise-derived body waves to ocean wave activities and microseism excitations. Li et al. (2019)
82 observed a prominent P -type phase that originates from the interference between microseism P and $PKPab$ waves and
83 has no counterpart in either theoretical or real seismograms. We denote it as the P_{dmc} phase, referring to that the
84 correlated P and $PKPab$ waves transmit directly through the deep mantle and the outer core, respectively. In section
85 2 of this paper, we describe the main results on the P_{dmc} observations. In section 3, we estimate the temporal variations
86 of the P_{dmc} signals and refute the associations to seismicity as have been reported by some authors for coda-derived
87 core phases (e.g., Boué et al., 2014; Lin and Tsai, 2013). In section 4, we use correlation analysis to study the
88 spatiotemporal links between the P_{dmc} signals and global oceanic microseism sources. Last, we discuss the significance
89 of this study in seismology, oceanography and climate science.

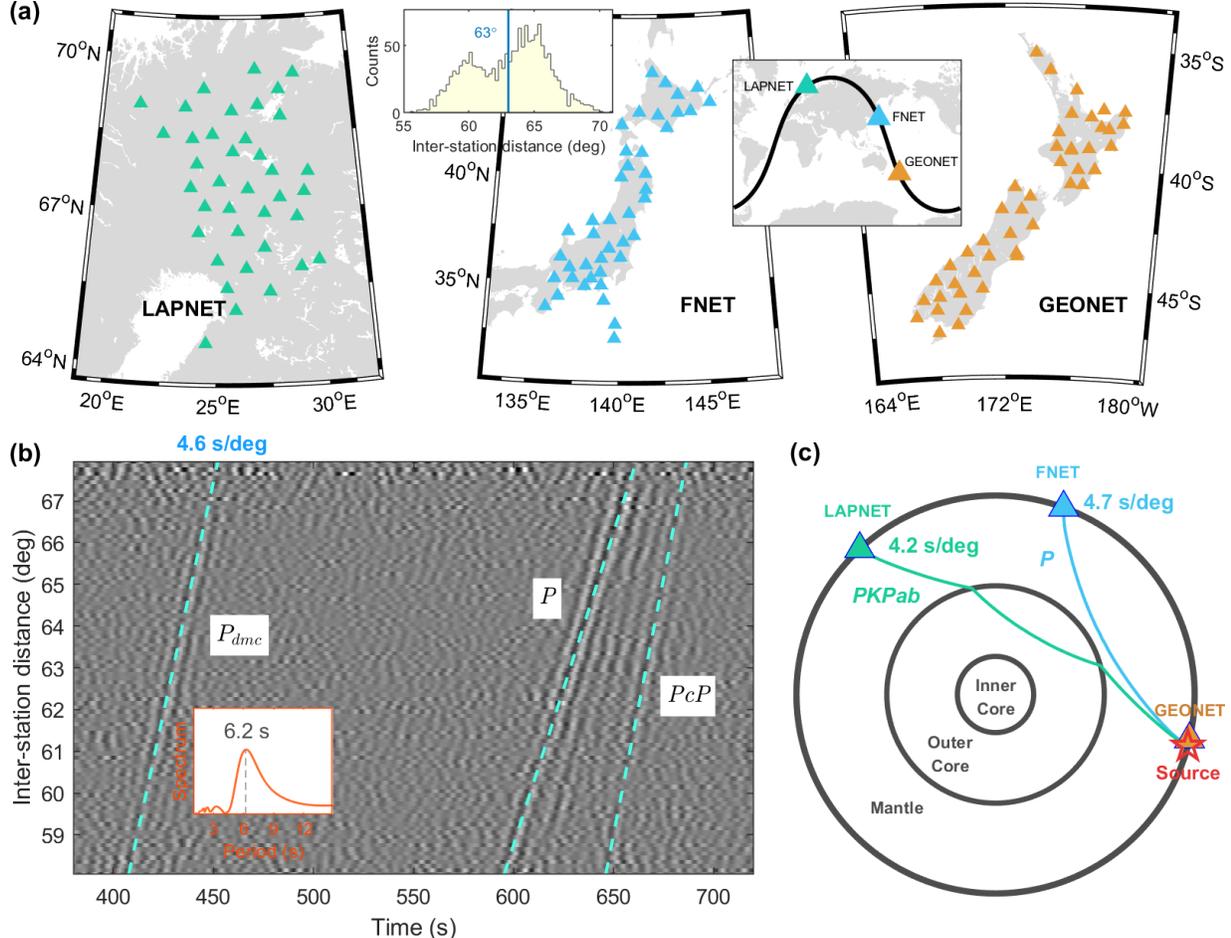
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91 **2 Noise-derived P_{dmc} phase**

92 Li et al. (2019) correlated the seismic noise records from two regional seismic networks at teleseismic distance: the
93 FNET array in Japan and the LAPNET array in Finland (Fig. 1a). From the vertical-vertical components of noise
94 correlations between the FNET-LAPNET station pairs, they observed coherent spurious arrivals (the P_{dmc} phase) that
95 emerged ~ 200 s earlier than the direct P waves (Fig. 1b). The P_{dmc} phase has a dominant period of 6.2 s, typical for
96 secondary microseisms. By estimating the respective slownesses of the interfering waves and their time delay, it is
97 unveiled that a quasi-stationary phase interference between the teleseismic P waves at FNET and the $PKPab$ waves
98 at LAPNET, emanating from noise sources in the ocean south of New Zealand (NZ), lead to the noise-derived P_{dmc}
99 phase (Fig. 1c). The quasi-stationary phase refers to that the interfering waves have no common path or slowness as
100 expected by the traditional stationary phase condition, but the stack of correlation functions over a range of sources
101 can still be constructive as an effect of finite frequency. The P_{dmc} phase has an apparent slowness of 4.6 s/deg, while
102 the slownesses of the interfering P and $PKPab$ waves are 4.7 s/deg and 4.2 s/deg, respectively. The observation of the
103 P_{dmc} phase is time-asymmetric (Fig. S1a). Its absence from the mirror side is ascribed to the faintness of the
104 corresponding source in the low-latitude Atlantic (Fig. S1b).

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106 There are several advantages to investigate the links between noise-derived signals and microseism sources with the

107 P_{dmc} phase. First, the correlated P and $PKPab$ waves are both prominent phases in the ballistic microseism wavefields.
 108 The P_{dmc} phase is easily observable from noise correlations, even between some single station pairs and on some single
 109 days (Fig. S2). Second, the isolation of P_{dmc} signals avoids potential bias caused by other prominent signals. Third,
 110 the effective source region is definite and unique. In contrast, noise-derived P wave can have multiple effective source
 111 areas (Boué et al., 2014). Fourth, the correlated FNET and LAPNET networks are next to the northern Pacific and
 112 Atlantic, respectively, while the effective source region locates in the southern Pacific. The northern oceans have
 113 consistent seasonal variation pattern distinct from (reverse to) that of the southern oceans (Hillers et al., 2012; Landés
 114 et al., 2010; Stutzmann et al., 2009). That will make the observations easier to interpret. Last, there happens to be a
 115 GEONET seismic array in NZ next to the effective source region for the P_{dmc} phase. The seismic data from GEONET
 116 provide extra support to our study.
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 119 Figure 1. (a) Three regional broadband seismic networks used in this study: left, the LAPNET array in Finland (38
 120 stations); center, the FNET array in Japan (41 stations); right, the GEONET array in New Zealand (46 stations). The
 121 histogram inset shows the distribution of the separation distances between the 1558 FNET-LAPNET station pairs.
 122 The center-to-center distance is 63° between LAPNET and FNET, and 85° between FNET and GEONET. The global
 123 inset shows the geographical locations of the three networks that are aligned on a great circle (dark line). (b) FNET-
 124 LAPNET noise correlations that are filtered between 5 s and 10 s and stacked in 0.1° bins. The spectrum inset indicates
 125 that the P_{dmc} phase has a 6.2 s peak period. (c) Ray paths of the interfering waves that generate the P_{dmc} phase. The
 126 effective source region is close to GEONET.
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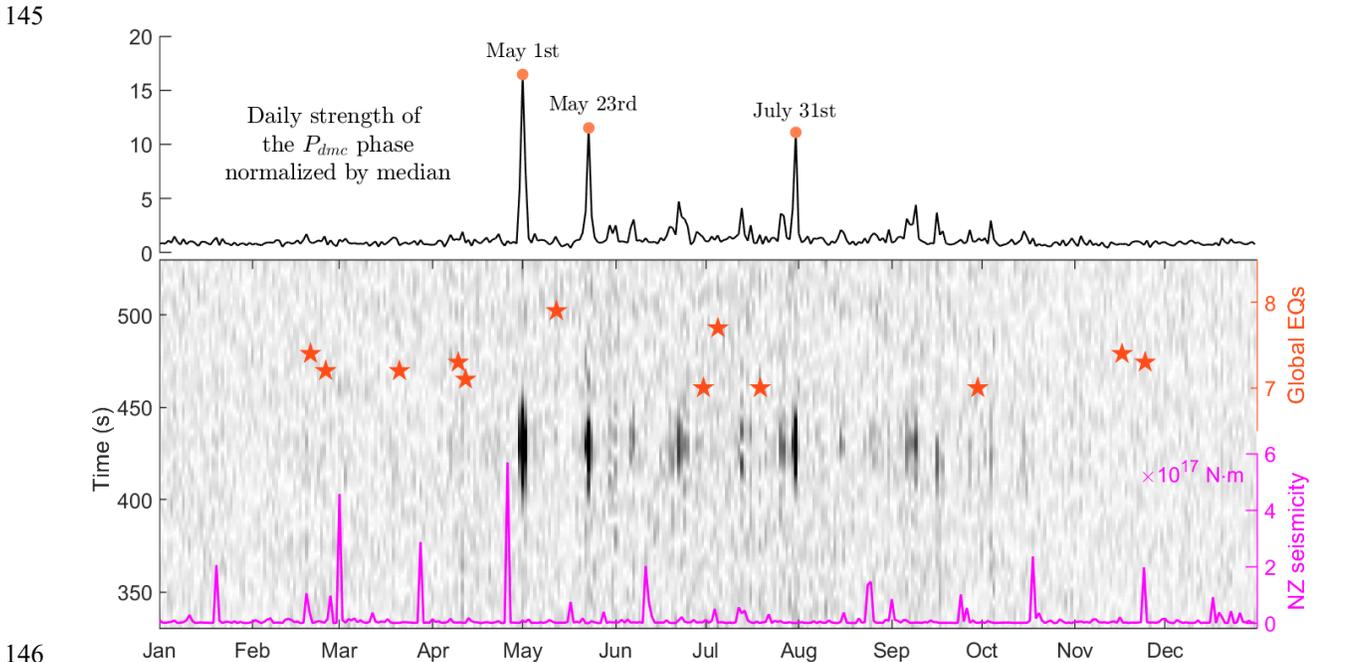
128 3 Temporal variations

129 We beam the daily FNET-LAPNET correlations by

$$130 B(t) = \langle C_{ij}(t + (d_{ij} - d_0) \cdot p) \rangle, \quad (1)$$

131 with $\langle \cdot \rangle$ the assemble mean operator, C_{ij} and d_{ij} the correlation function and the distance between the i th FNET station
 132 and the j th LAPNET station, d_0 the reference distance (63°), p the apparent slowness of the P_{dmc} phase (4.6 s/deg),
 133 and t the time. Figure 2 shows the envelopes of daily beams computed from the Hilbert transform of Eq. (1). The
 134 strength of daily P_{dmc} signals varies strikingly. It is extremely strong on some single days (labeled dates for instances),
 135 but indiscernible on most other days.

136
 137 Considering that the region of effective source is tectonically active, one needs to investigate the plausible connection
 138 between P_{dmc} signals and seismicity. From Fig. 2, it is obvious that P_{dmc} is decorrelated with the NZ seismicity. Also,
 139 it shows no relevance with global large earthquakes as has been observed for coda-derived core phases at periods of
 140 20 to 50 s (Boué et al., 2014; Lin and Tsai, 2013). The P_{dmc} strength exhibits a pattern of seasonal variation, which
 141 does not favor a tectonic origin (seismicity shows no clear seasonal pattern), but prefers an oceanic origin (oceanic
 142 microseism excitations have seasonal variations; see e.g. Hillers et al., 2012; Landés et al., 2010; Stehly et al., 2006;
 143 Stutzmann et al., 2009). Next, we analyze the correlations between P_{dmc} signals and oceanographic data at a global
 144 scale.

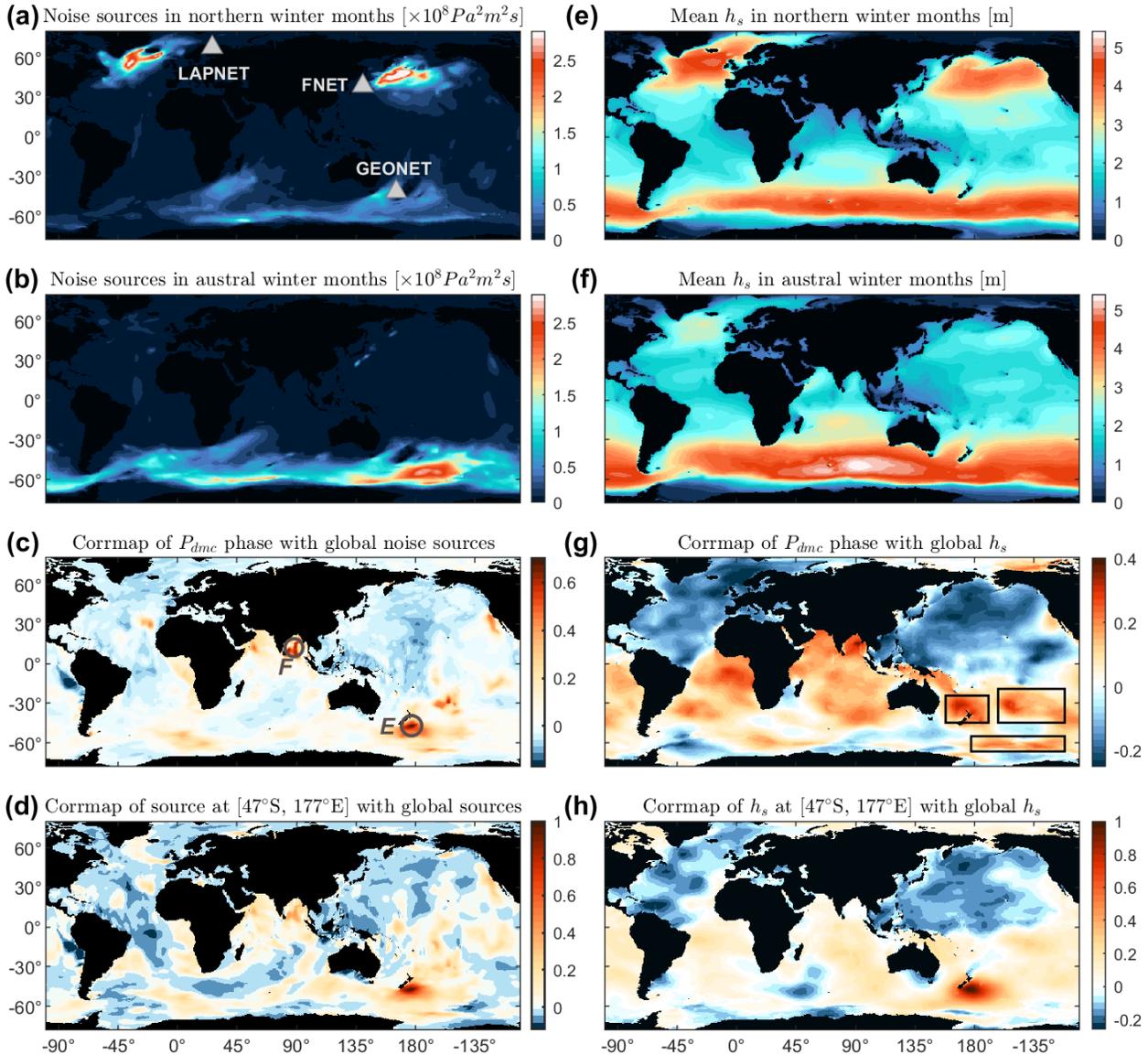


146 Figure 2. Temporal variations in the strength of daily P_{dmc} signals, in comparisons with daily cumulative seismic
 147 moments for magnitudes above 2.0 in NZ (line at bottom) and global large earthquakes (stars; magnitudes above 7.0).
 148 The background image is composed of columns of daily envelopes of beamed FNET-LAPNET noise correlations.
 149 Darker color represents larger amplitude. The top curve shows the daily P_{dmc} strength derived from the daily envelopes.
 150 Dates of the three largest peaks are labeled.

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 153 **4 Correlation analysis**

154 The sea state is composed of ocean waves at various frequencies and propagation directions. The nonlinear interaction
 155 between nearly equal-frequency ocean waves traveling in nearly opposite directions is equivalent to a vertical random
 156 pressure applied to the ocean surface (Hasselmann, 1963; Longuet-Higgins, 1950), so that microseisms are generated.
 157 Figure 3(a) shows a global map of average Power Spectral Density (PSD) of the equivalent surface pressure for a
 158 seismic period of 6.2 s, during the northern winter months of 2008. The most energetic microseism excitations occur
 159 in the northern Atlantic south of Greenland and Iceland (near LAPNET), and in the northern Pacific between Japan
 160 and Alaska (near FNET). Figure 3(b) shows the map for the austral winter months, with the strongest excitations
 161 occurring between NZ and Antarctic (near GEONET). The seasonal pattern of oceanic microseism excitations results
 162 from the same pattern of global wave climate (Figs 3e-f). The seasonal pattern of the P_{dmc} strength agrees with that of
 163 the microseism excitation and wave climate in the effective source region south of NZ.
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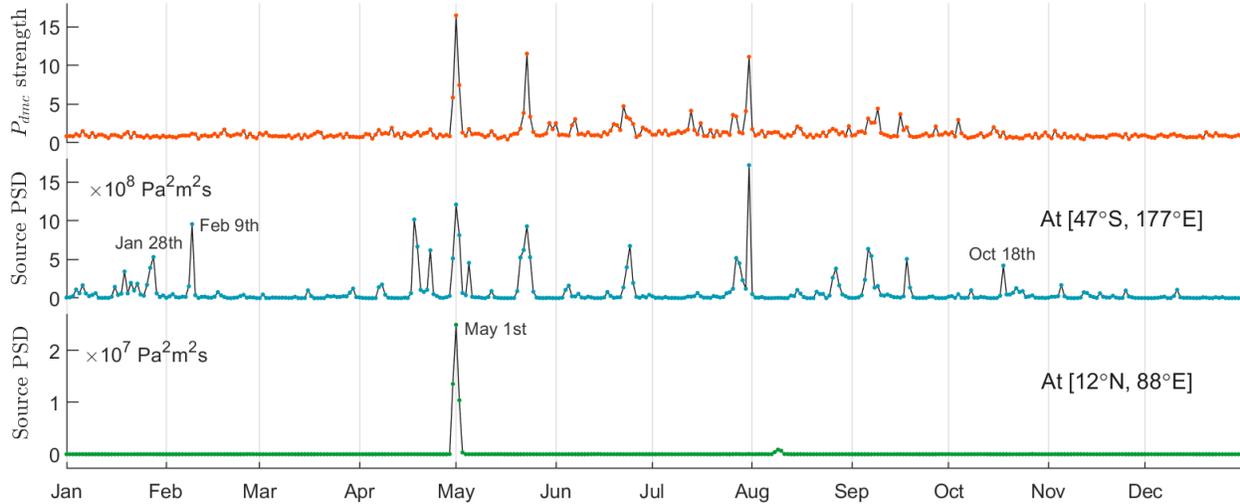
165 We compute the correlation coefficient (denoted as r) between the P_{dmc} strength and the source PSDs at each grid
 166 point, and thereby obtain a global correlation map (Fig. 3c). As for a freedom degree of 364 for 366 samples and a p -
 167 value of 0.05 for null hypothesis, r values above 0.1 are statistically significant. The highest r arises at $[47^\circ\text{S}, 177^\circ\text{E}]$
 168 in the effective source region (E in Fig. 3c). The corresponding time series of daily source PSDs is plotted in Fig. 4,
 169 in parallel with the P_{dmc} strength. Large peaks in the P_{dmc} series have good correspondence with large peaks in the
 170 source PSD series. From Fig. 3(c), one can observe a broad region of positive r values (red colors; roughly, south
 171 Atlantic, south Pacific, and Indian ocean). However, the positive correlation does not imply a causality between the
 172 P_{dmc} phase and the sources outside the effective region E . We ascribe the apparent positive correlation to the spatial
 173 connectivity of the time-varying microseism excitation in region E with global sources (Fig. 3d). We also notice there
 174 are high- r regions that may not be fully explained by the spatial connectivity. These regions are characterized by low
 175 intensity of microseism excitations. A striking example is around $[12^\circ\text{N}, 88^\circ\text{E}]$ in the Bay of Bengal (F in Fig. 3c).
 176 From Fig. 4, it can be seen that the source PSD series for $[12^\circ\text{N}, 88^\circ\text{E}]$ is dominated by a single peak around May 1st,
 177 coincstantaneous with the largest P_{dmc} peak. This coincidence leads to a fake high correlation. Figures 3(g-h) show the
 178 correlation maps for h_s , which will be discussed later.
 179



180 Figure 3. (a) Global map of average PSD of oceanic microseism sources in 2008 northern winter months (Jan. to Mar.
 181 and Oct. to Dec.), for a seismic period of 6.2 s. (b) Similar to (a) but for 2008 austral winter months (Apr. to Sep.). (c)
 182 Correlation map (corrmap) for the P_{dmc} strength and global microseism sources. Circles mark two regions with highest
 183

184 correlation coefficients: *E*, effective source region surrounding [47°S, 177°E] south of NZ; *F*, fake highly-correlated
 185 region surrounding [12°N, 88°E] in the Bay of Bengal. (d) Correlation map for the source at [47°S, 177°E] and global
 186 sources. (e) Mean significant wave height (h_s ; four times the square root of the zeroth-order moment of ocean-wave
 187 frequency spectrum) in northern winter months. (f) Similar to (e) but for austral winter months. (g) Correlation map
 188 for the P_{dmc} strength and global wave heights. (h) Correlation map for wave heights at [47°S, 177°E] and global wave
 189 heights. The oceanographical hindcast data are provided by the IOWAGA products (Rascle and Ardhuin, 2013).

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192 Figure 4. True correlation ($r = 0.73$) between the P_{dmc} strength and the source at [47°S, 177°E] in the effective source
 193 region (*E* in Fig. 3c), and false correlation ($r = 0.71$) between P_{dmc} and the source at [12°N, 88°E] in the Bay of Bengal
 194 (*F* in Fig. 3c).

195

196 As shown in Fig. 4, prominent peaks in the P_{dmc} series have correspondence in the source PSD series for the effective
 197 source at [47°S, 177°E]. However, there are some peaks in the latter without correspondence in the former (see the
 198 labeled dates for examples). To verify if this is due to the spreading of the effective source region has been neglected,
 199 we compare the P_{dmc} series with the source PSD series for the full effective source region, rather than at a single point.
 200 The bathymetric effect should be considered when evaluating the overall microseism excitation in the effective source
 201 region. Using the equations proposed by Gualtieri et al. (2014) and the bathymetry around NZ (Fig. 5a), we compute
 202 the bathymetric amplification factors for *P* waves (Fig. 5b). The P_{dmc} phase has varying sensitivities to the sources in
 203 the effective region. The weights are obtained by back-projecting the beam power of noise correlations onto a global
 204 grid (Fig. 5c; see Supplementary for technical details). Figure 5(d) shows the map of annually averaged source PSDs
 205 surrounding NZ and Fig. 5(e) shows the map after the modulation of the bathymetric amplification factors in Fig. 5(b).
 206 The spatial patterns are altered significantly, indicating the importance to account for the bathymetric effect. The final
 207 source imaging that has been weighted by Fig. 5(c), is plotted in Fig. 5(f), which agrees well with the effective source
 208 region *E* determined from the correlation map in Fig. 3(c). Replacing the annual PSD map in Fig. 5(d) with daily PSD
 209 maps, we obtain maps like Fig. 5(f) for each date. Averaging over the map leads to the time series of daily effective
 210 source intensity (Fig. 6). The new series has almost the same peaks as the series for [47°S, 177°E] in Fig. 4, suggesting
 211 that the observed disparities between peaks of the P_{dmc} strength and the effective source intensity are caused by other
 212 reasons.

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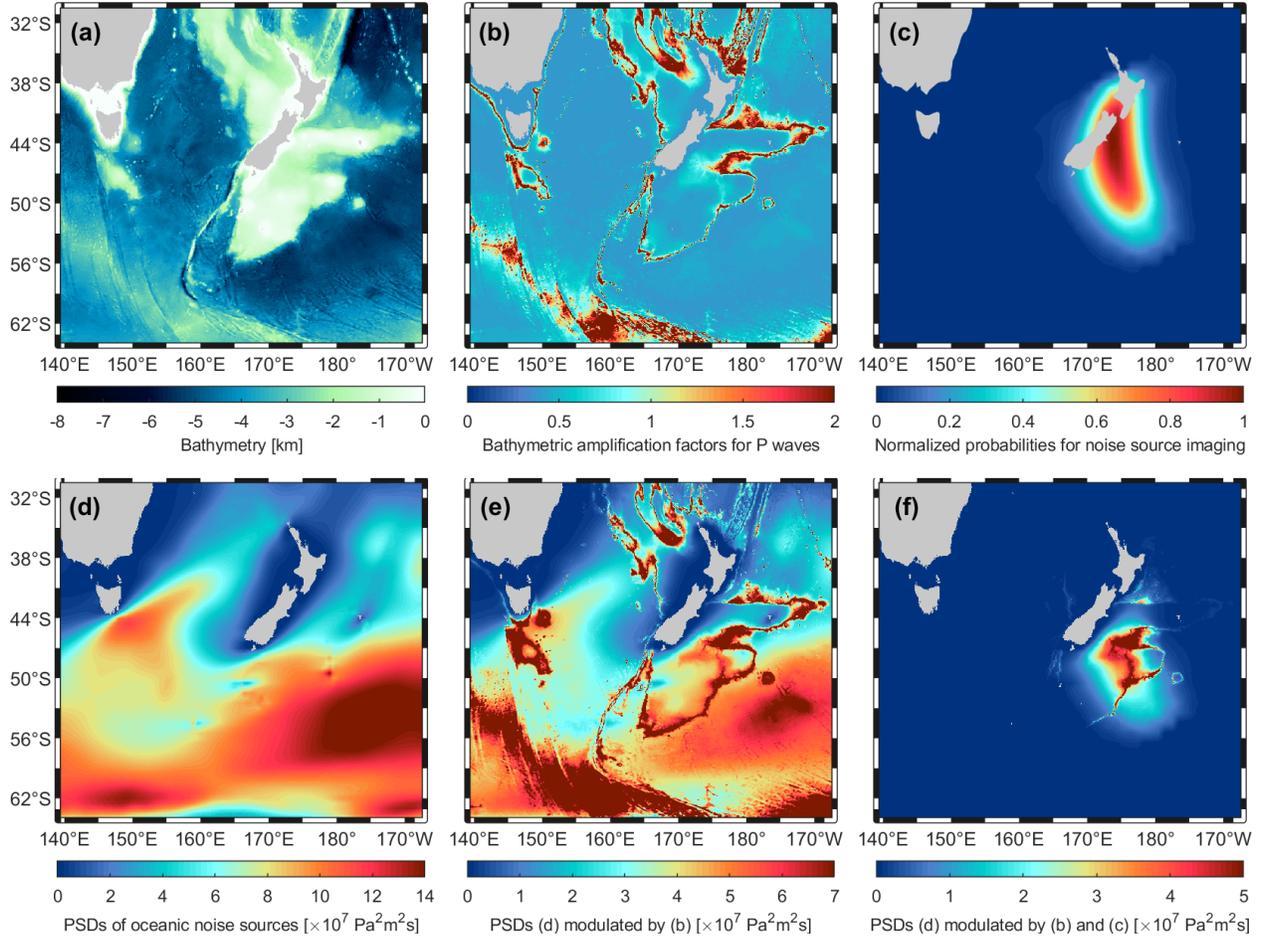
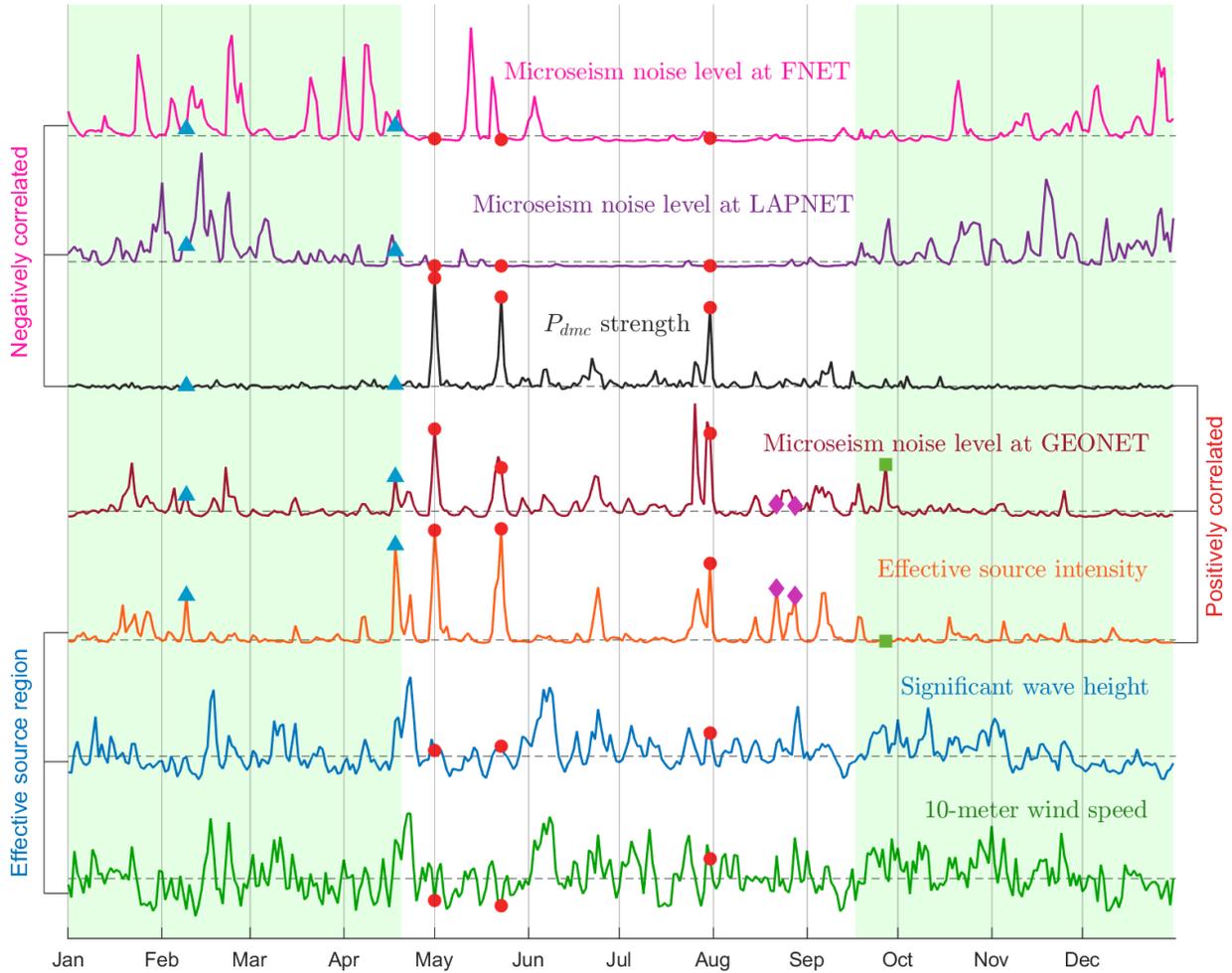


Figure 5. (a) Bathymetry around NZ. (b) Geometric average of bathymetric amplification factors for 6.2 s period P -type waves with slownesses of 4.2 s/deg and 4.7 s/deg. (c) Weights for oceanic microseism sources obtained from the back-projection of the FNET-LAPNET noise correlations. (d) Annual average of source PSDs in 2008. (e) Source PSDs in (d) modulated by the factors in (b). (f) Source PSDs in (e) further modulated by the weights in (c).

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The microseism source PSD data are simulated from the hindcast data of ocean wave directional spectra base on the excitation theory of Longuet-Higgins (1950) and Hasselmann (1963), which have no constrains from seismological observations. One should consider the accuracy of the simulation: the peak disparities in Fig. 4 can be ascribed to the simulation error or not? The seismic noise records from the GEONET array adjacent to the effective source region provide the opportunity to validate the simulation. To obtain the daily microseism noise levels at GEONET, we apply the Hampel filter, a variant of the classic median filter, to the continuous seismograms to discard earthquakes and anomalous impulses. The filter replaces outliers with the medians of the outliers' neighbors and retains the normal samples. Technical details are provided in the Supplementary. The resultant GEONET noise level exhibits a good correlation with the effective source intensity ($r = 0.7$). We thus affirm that the numerical simulation is statistically reliable. When the effective source intensity is high, the GEONET noise level should also be high (see the peaks marked by dots in Fig. 6 for examples). However, due to the great spatiotemporal variability of noise sources in the effective region and the complexity of seismic waves propagating from ocean to land (Gualtieri et al., 2015; Ying et al., 2014), a larger peak in the source intensity series does not necessarily imply a larger peak in the noise level time series (e.g., comparing the peaks marked by diamonds in Fig. 6 with the beside peaks). We also emphasize that a high GEONET noise level does not need to always have a correspondence in the source intensity (see squares in Fig. 6 for example), because the GEONET stations record microseisms emanating from noise sources all around, not only from the effective source region



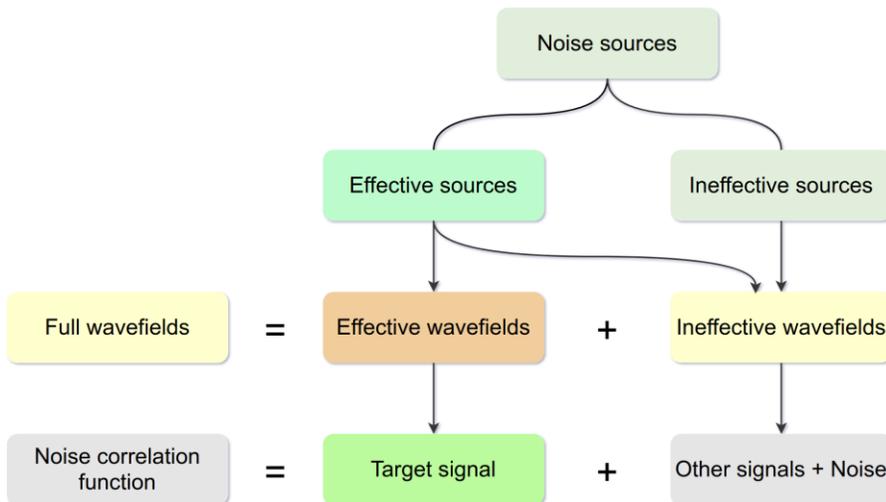
238 Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec
 239 Figure 6. Temporal variations of daily P_{dmc} strength, microseism noise levels at three networks, and average wind
 240 speeds, wave heights and microseism excitations in the effective source region. The curves are normalized by their
 241 own maximums. Dashed horizontal lines denote their respective medians. Symbols mark some dates cited in the main
 242 text. When computing the effective source intensity, factors in Fig. 5(b) and weights in Fig. 5(c) are used. When
 243 computing the average wind speeds and wave heights, weights in Fig. 5(c) are used.
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245 The above analysis validates the observed disparities between P_{dmc} strength and effective source intensity. From Fig.6,
 246 one can see that the disparities primarily emerge in the shaded period when dominant microseism sources shift to the
 247 north hemisphere. The shading roughly separates the northern winter from the austral winter. The correlation between
 248 P_{dmc} strength and effective source intensity is low in the shaded period ($r = 0.16$; outside the shading: $r = 0.74$). Large
 249 P_{dmc} peaks always emerge on dates during the austral winter when the effective source intensity is much higher than
 250 its median, and meanwhile, noise levels at FNET and LAPNET are below their respective medians (see dots in Fig. 6
 251 for examples). The seasonal variations of oceanic sources in the south hemisphere is less strong than in the north
 252 hemisphere (Fig. 3). On some dates (see triangles in Fig. 6 for examples), the effective source intensity can be
 253 considerable, but relevant P_{dmc} peaks are still missing. We notice that the corresponding microseism levels at FNET
 254 and LAPNET are obviously above their medians. Intensive ocean activities and microseism excitations in the north
 255 Pacific and Atlantic, lead to increased microseism noise levels at FNET and LAPNET. The P_{dmc} strength is anti-
 256 correlated with microseism noise levels at FNET ($r = -0.12$) and LAPNET ($r = -0.18$). We hereby conjecture that the
 257 microseism energy from the distant effective source region is dwarfed by the energetic microseisms excited by the
 258 oceanic sources closer to the correlated FNET and LAPNET arrays, and consequently, P_{dmc} signals are overwhelmed by the
 259 background noise in the FNET-LAPNET cross-correlations.
 260

261 **5 Discussions and conclusions**

262 In this study, we analyzed the spatiotemporal correlations between noise-derived P_{dmc} signals and global oceanic
 263 microseism sources. In our case, the correlated seismic networks are located in the north hemisphere, while the
 264 effective source region is in the south hemisphere. Ideally, we expect a correlation map with these features: positive
 265 correlation with sources in the effective region; negative or trivial correlations with other sources. Positive correlation
 266 indicates contributing to the construction of P_{dmc} signal from noise correlations, negative correlation implies an
 267 adverse effect, and trivial correlation means a neglectable effect on the signal construction. However, we obtained a
 268 correlation map roughly showing that, the P_{dmc} signal is correlated with the southern sources and anti-correlated with
 269 the northern sources. The correlation with southern sources outside the effective region can be interpreted with the
 270 spatiotemporal connections of microseism sources in the southern oceans. The anti-correlation with northern sources
 271 can partly be explained by the reverse seasonal patterns of oceanic microseism excitations in the south and north
 272 hemispheres. Another important reason is that compared to the remote effective sources in the south hemisphere, the
 273 northern sources closer to the correlated stations have larger impacts on the microseism noise levels at stations. Strong
 274 energy flux from the northern sources outshines the microseism energy coming from the distant effective sources.
 275 That deteriorates the construction of the P_{dmc} phase. The noise-derived P_{dmc} signals are primarily observable in the
 276 austral winter. That can be, on one hand, attributed to the stronger effective source excitation during that period, and
 277 on the other hand, to the relative tranquility in the north oceans.

279 We generalize the above discussions to any noise-derived signal (termed as target signal for convenience). The noise
 280 correlation function is composed of the target signal, other signals and background noise. A source or a wave is called
 281 effective if it contributes to the construction of the target signal from noise correlations. Otherwise, it is called
 282 ineffective. The construction of the target signal is exclusively ascribed to the interferometry between the effective
 283 wavefields emanating from effective sources. Stronger effective sources (relative to ineffective sources) lead to a
 284 larger portion of the effective wavefields occupying in the total wavefield, and thereby, a better quality for the noise-
 285 derived target signal. Note that not all waves emanating from the effective sources, but only those following specific
 286 ray paths, are effective. In the case of the P_{dmc} phase, the effective waves are those following the paths of P and $PKPab$.
 287 Weak phases in the ballistic wavefield, like the PcS-PcPPcP correlation as discussed by Li et al. (2019), have fewer
 288 contributions to the noise-derived signals. Figure 7 summarizes the classification of noise sources, the decomposition
 289 of wavefields, together with the associations to the constituents of the inter-receiver noise correlation function.
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 292 Figure 7. Sketch explanation for the relationships between noise sources and noise-derived signals.
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294 The effective source region E for the P_{dmc} phase is successfully identified from the correlation map in Fig. 3(c). It is
 295 consistent with that determined from seismological back-projection (Fig. 5). The correlation map provides an easy
 296 way to identify the effective sources for noise-derived seismic signals. The double-array back-projection is superior
 297 to the classic single-array back-projection in reducing the ambiguity in the detection and location of microseism events.
 298 A catalogue of microseism events would be promising in teleseismic body-wave tomography (Boué et al., 2013;
 299 Nishida and Takagi, 2016; Zhang et al., 2010). The good consistency of the temporal variations in the P_{dmc} strength,
 300 the effective source intensity and the NZ microseism noise level (Fig. 6), provides extra supports to the analysis of the

301 P_{dmc} observations and the quasi-stationary phase arguments by Li et al. (2019), and also validates the numerical
302 modeling of oceanic microseism sources by Ardhuin et al. (2011) and Rascle and Ardhuin (2013).

303
304 In the $P_{\text{dmc}}-h_s$ correlation map in Fig. 3(g), the largest r values do not fall in region E as in Fig. 3(c), but in surrounding
305 regions with moderate to high ocean wave activities (the bounded areas). We speculate that these regions could be the
306 birthplaces of the colliding swells that incite the secondary microseisms in region E , or the ocean waves in these
307 regions are driven by the same storms as the colliding waves in region E (see the spatial connectivity of h_s from Fig.
308 3h and supplementary movie). From Fig. 6, one can observe a high correlation between wind speed and wave height
309 in region E ($r = 0.74$). It indicates that the ocean waves in the region E are likely dominated by the wind waves forced
310 by local winds. The correlation between wave height and microseism excitation is low ($r = 0.25$), implying a dominant
311 role of swells in exciting the microseisms. Extreme sea state does not guarantee strong microseism excitation. It is not
312 surprising according to the microseism excitation theory (Hasselmann, 1963; Longuet-Higgins, 1950). In lack of
313 equal-frequency waves coming from opposite directions, even extreme wave climate cannot incite strong secondary
314 microseisms. In contrast, for large peaks in the microseism excitation, the corresponding wave heights are generally
315 moderate (e.g., on May 1st and 23rd). On these two dates, the low wind speeds but moderate wave heights in region E
316 suggest that the ocean waves are dominantly the freely travelling swells from elsewhere (see the supplementary movie).
317 Oppositely propagating equal-frequency swells collide with each other and incite strong microseisms. Obrebski et al.
318 (2012) reported similar observations in eastern Pacific. There are also some examples that wind waves can play a role
319 in the excitation of microseisms (e.g., around July 31st).

320
321 We have described above the implications of this study in seismology and in understanding the process of microseism
322 excitation. Now, we discuss the significance in oceanography and climate science. Microseisms are induced by storm-
323 driven ocean waves (Ardhuin et al., 2015; Hasselmann, 1963; Longuet-Higgins, 1950). Seismograms have registered
324 the imprint of climate (Aster et al., 2010; Stutzmann et al., 2009). Instrumental observation of microseisms has a
325 history over a century, much earlier than the modern observations of ocean waves and storms. Researchers expect that
326 past seismic records can be used to recover undocumented historical ocean storms and wave climate, which would be
327 valuable in improving our understanding about climate change and global warming (Ebeling 2012). The correlation
328 analysis in this study suggests that it is practicable to detect remote microseism events with land observation of
329 microseisms. The event detection could be effective when there are no strong sources near the station, otherwise it
330 could be missed. Stations at low latitudes where wave climate and microseism excitation are relatively mild, or inland
331 stations far from oceans, should have better performance in remote detection. Powerful microseism excitation does
332 not need extremal *in situ* wave height, and extremal wave climate does not necessarily produce energetic microseisms,
333 suggesting that secondary microseism events are not perfect proxy for extremal *in situ* wave climate. However, it does
334 not mean the long-lasting attempt to monitor remote sea state and ocean storms with land observation of secondary
335 microseisms is unavailing. The emergence of microseism events affirms the existence of causal storms, despite the
336 storms could be distant from the events. Last, we mention that from the correlation maps in Fig. 3, we observe an
337 interesting phenomenon that the microseism excitations and wave climates in part of the west coast of Africa and of
338 US in the north hemisphere are connected to those in the south hemisphere.

339

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344 were provided by the European Centre for Medium-Range Weather Forecasts (<https://www.ecmwf.int>). The hindcast
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 437

Supplementary

FNET-LAPNET noise correlations

The correlation function C_{AB} between two seismograms (S_A and S_B) is given by

$$C_{AB}(\tau) = \frac{\sum_i S_A(i)S_B(i-\tau)}{\sqrt{\sum_i S_A^2(i) \sum_i S_B^2(i)}}$$

The resultant C_{AB} consists of an acausal part and a causal part, that correspond to the negative lags ($\tau < 0$) and the positive lags ($\tau > 0$), respectively. For efficiency, it is routine to compute the correlation function with the Fast Fourier Transform:

$$C_{AB}(\tau) = \frac{\mathcal{F}^{-1}[\mathcal{F}(S_A)\mathcal{F}^*(S_B)]}{\sqrt{\sum_i S_A^2(i) \sum_i S_B^2(i)}}$$

Figure S1(a) shows the acausal and causal sections of FNET-LAPNET noise correlations that are filtered between 5 s and 10 s and binned in distance intervals of 0.1° . The acausal section is flipped to share the time axis with the causal section. The expected locations of the acausal and causal noise sources are marked by stars on the maps of global microseism source PSDs and ocean wave heights in Fig. S1(b). The ocean wave activities and microseism excitations at the acausal source region are intense, while those in the causal source region are fainter. Consequently, the P_{dmc} phase is only observable from the acausal noise correlations.

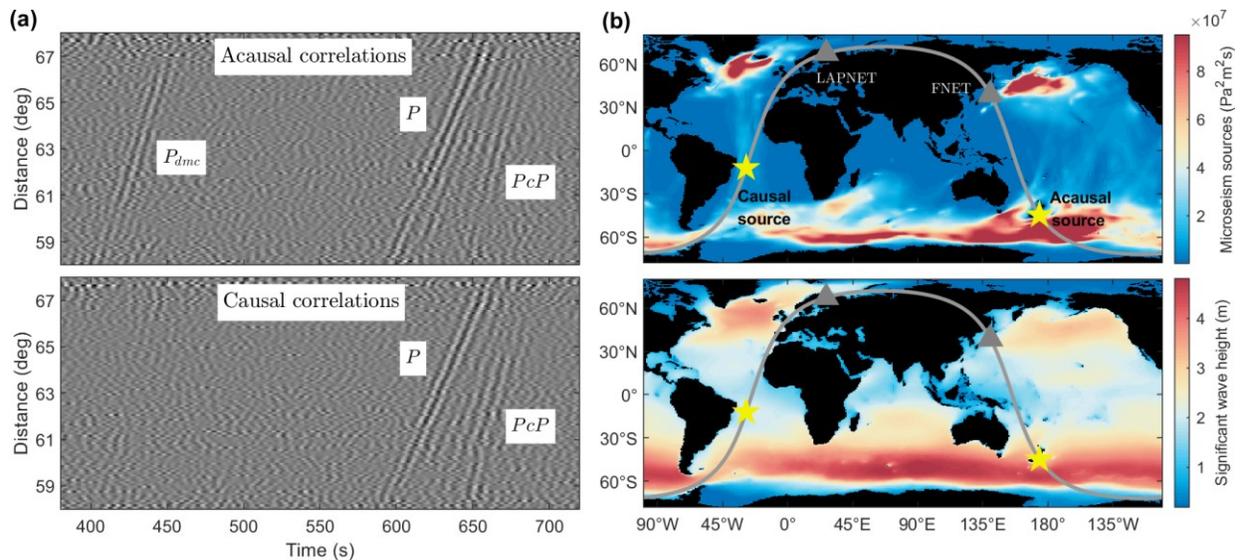


Figure S1. (a) Acausal and causal sections of FNET-LAPNET noise correlations. (b) Global maps of 6.2 s period secondary microseism sources and significant wave heights in 2008.

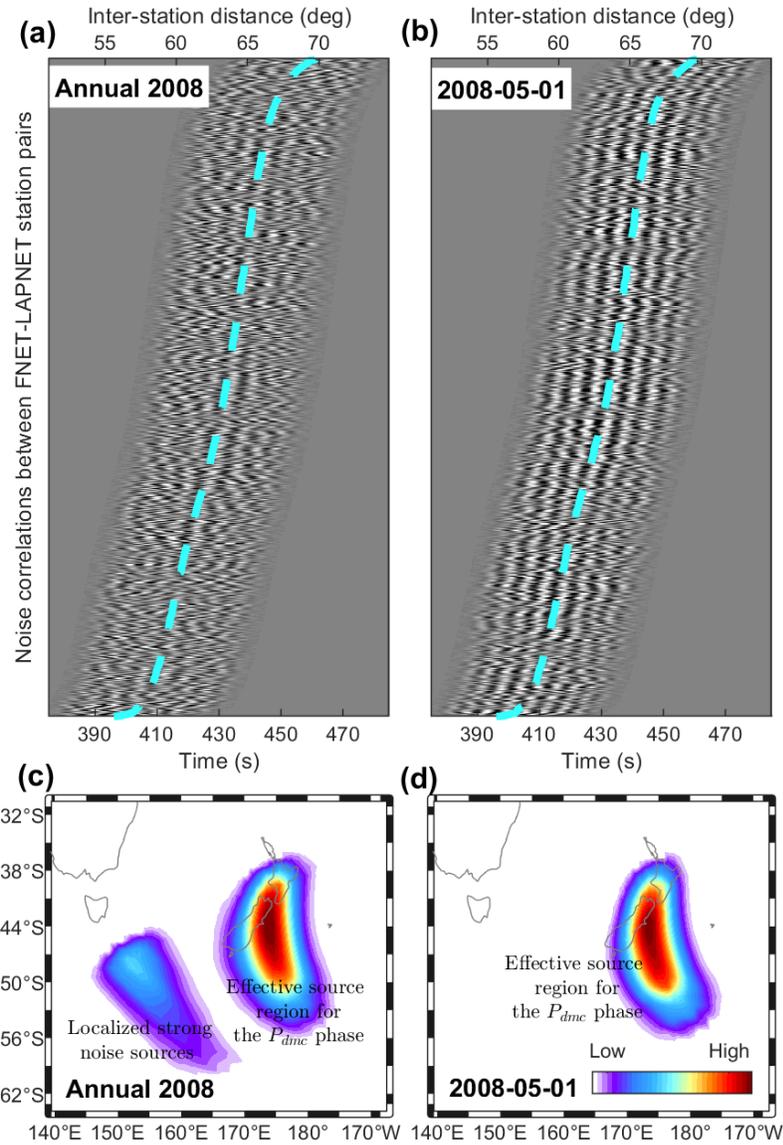
22 **Noise source imaging by back-projection**

23 Assuming the interferometry between P waves at FNET and PKPab waves at LAPNET, we
24 image the effective noise sources through the back-projection of the FNET-LAPNET noise
25 correlations. We beam the FNET-LAPNET noise correlations and assign the beam power

26
$$P_s = \langle \langle C_{ij}(t + t_{si} - t_{sj}) \rangle_{ij}^2 \rangle_t, \quad (\text{S1})$$

27 to a $0.5^\circ \times 0.5^\circ$ grid as the probabilities of noise sources on the global surface. In the above
28 equation, $\langle \cdot \rangle_x$ means the average over x , C_{ij} is the correlation function between the i -th FNET
29 station and the j -th LAPNET station, t_{si} is the traveltime of the P wave from the s -th grid point
30 to the i -th FNET station, and t_{sj} is the traveltime of the PKPab waves from the s -th source to the
31 j -th LAPNET station. The inter-station noise correlations are windowed before the beamforming
32 (Fig. S2a). The noise source imaging for the annually stacked noise correlations is plotted in Fig.
33 S2(c). Only the region surrounding NZ is shown. Outside the region, hardly can the P wave
34 reach FNET or the PKPab waves reach LAPNET. Besides a well-focused imaging of the
35 expected source region in the ocean south of NZ, we notice a secondary spot to the west. In
36 comparisons with the power map of oceanic microseism noise sources in Fig. 5(e), we ascribe it
37 to the strong microseism excitation in the ocean south of the Tasmania island of Australia. We
38 also back-project the daily noise correlations on 2008-05-01 (Fig. S2b), when the P_{dmc} phase
39 reaches the largest strength through the year (Fig. 2). As shown in Fig. S2(d), an exclusive
40 source region is imaged, which agrees with the dominant spot in Fig. S2(c).

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48 **Microseism noise levels at seismic networks**

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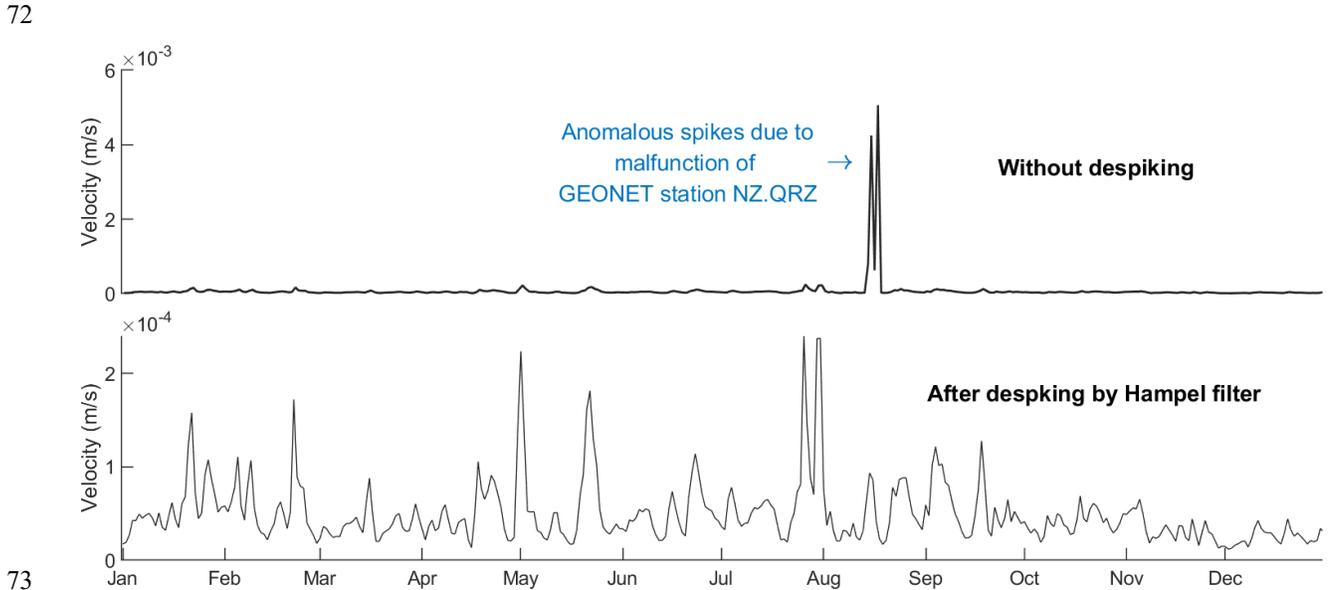
55

Figure S2. Inter-receiver noise correlations for all FNET-LAPNET station pairs: (a) stacked over the year of 2008; (b) on single day of 2008-05-01. The waveforms are windowed around the P_{dmc} phase. Dashed lines indicate inter-station distances. Back-projection imaging of noise sources: (c) using data from (a); (d) using data from (b).

The continuous seismograms record not only the background vibrations of Earth, but also ground motions induced by seismicity or other events. Instrumental malfunction also leads to anomalous (e.g., nearly vanishing or extremely large) amplitudes in the records. These extreme amplitudes (outliers) could bias the estimates of microseism noise power. It is necessary to get rid of them from the ambient noise records before the computation of noise power. Mean and median filters are commonly-used tools for this task. However, they modify all the samples. Here, we prefer to use a variant of the median filter called Hampel filter. In contrast to the median filter that replace

56 all samples with local medians, the Hampel filter detects outliers by compare a sample with the
57 neighboring samples. A sample is replaced by the local median if it deviates k times of the
58 median absolute deviation (MAD) from the local median, or else, it is unchanged.

59
60 We filter the vertical components of the continuous seismograms around 6.2 s period. The
61 seismograms are then divided into 15-min segments and the power of segments are computed.
62 We apply the Hampel filter to the time series of noise power recursively. For each sample, we
63 compute the local median and MAD of its eight neighbors (four before and four after). A sample
64 is replaced by the median if it deviates from the median over three times of the MAD. The
65 despiked time series is resampled from a 15-min interval to a 1-hour interval, by averaging over
66 every four samples. Then, we apply the Hampel filter again and resample the time series to a 24-
67 hour interval. The averaging of noise levels over all stations of a seismic network leads to the
68 time series of array noise level. Before the averaging, the Hampel filter is applied again, to
69 discard possible anomalous values at some stations (see Fig. S3 for the example of GEONET).
70 The final time series of microseism noise levels for networks FNET, LAPNET and GEONET are
71 shown in Fig. 6.



74 Figure S3. Comparison between the time series of daily GEONET noise levels with (lower) and
75 without (upper) despiking using the Hampel filter.

76