

Coral reef species assemblages are associated with ambient soundscapes

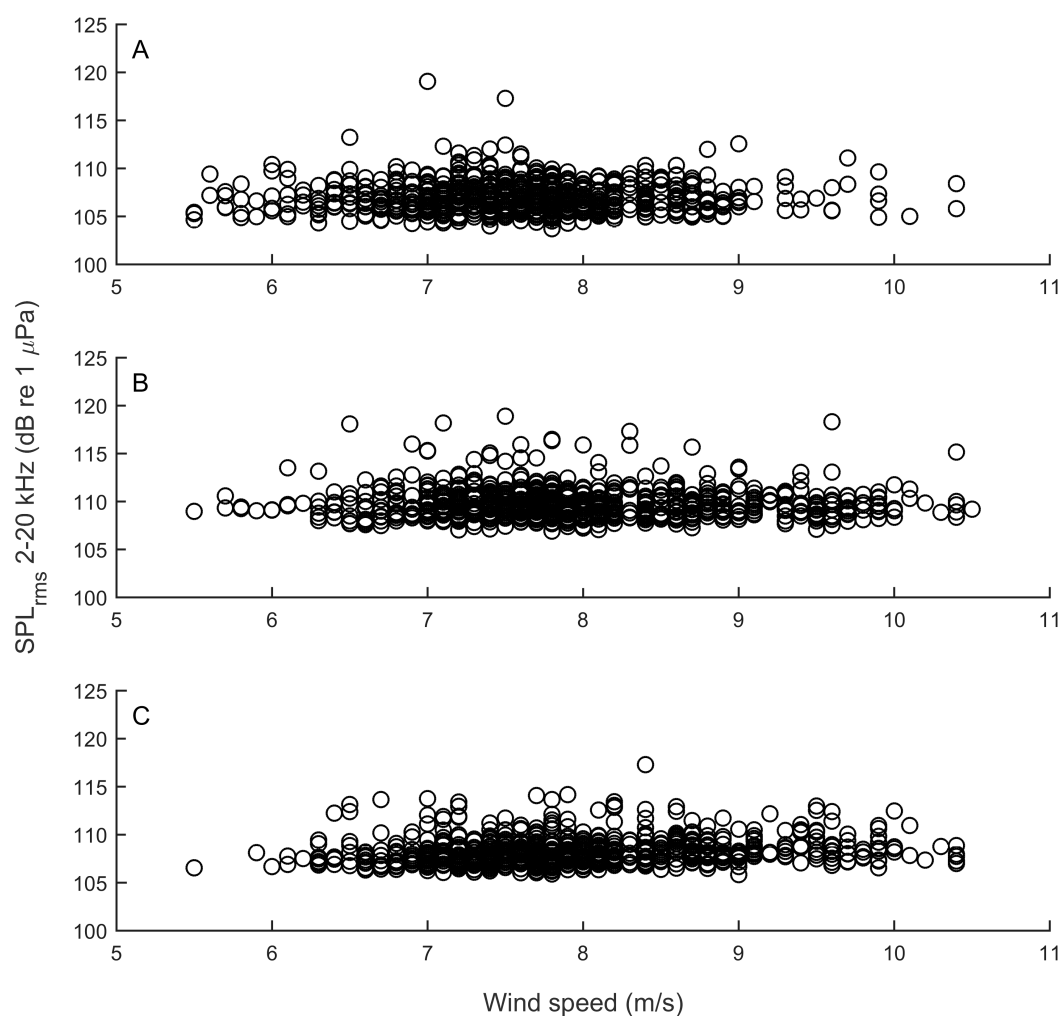
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Supplement.

Fig. S1. Relationship between sound pressure level in the high frequency shrimp band (2-20 kHz) and wind speed for three US Virgin Islands reefs, Tektite (A), Yawzi (B), and Ram Head (C). Only Ram Head demonstrated a significant relationship between wind speed and sound pressure level, albeit weakly.



Here, we outline periodogram-based likelihood ratio (LR) tests for temporal and spatial non-stationarity of the spectral density function (sdf) of reef sound. The statistical results underlying these tests are available in Dzhaparidze (1986).

Let X_{th} , $t = 1, 2, \dots, T$, be a collection of T independent time series each of length n recorded at times of day $h = 1, 2, \dots, H$ on days $t = 1, 2, \dots, T$ and let $f_{th}(\omega)$ be the unknown sdf at time of day h on day t . Interest centers on testing the null hypothesis $H_0 : f_{th}(\omega) = f_h(\omega)$ for all h and t that the sdf at each time of day is stationary over time against the general alternative hypothesis H_1 that it is not.

Let $I_{th}(\omega_j)$ be the periodogram ordinate for X_{th} at Fourier frequency $\omega_j, j = 1, 2, \dots, J$. It is a standard result that $I_{th}(\omega_j)$ is approximately independent of $I_{th}(\omega_k)$ and has an approximate exponential distribution with mean $f_{th}(\omega_j)$ and probability density function:

$$g(I_{th}(\omega_j)) = \frac{1}{f_{th}(\omega_j)} \exp\left(-\frac{I_{th}(\omega_j)}{f_{th}(\omega_j)}\right)$$

The log likelihood is given by:

$$\log L = -\sum_{t=1}^T \sum_{h=1}^H \sum_{j=1}^J \left(\log f_{th}(\omega_j) + \frac{I_{th}(\omega_j)}{f_{th}(\omega_j)} \right)$$

The maximum likelihood (ML) estimate of $f_{th}(\omega_j)$ under H_1 is simply $I_{th}(\omega_j)$ and the corresponding maximized value of the log likelihood is:

$$\log L_1 = -THJ - \sum_{t=1}^T \sum_{h=1}^H \sum_{j=1}^J \log I_{th}(\omega_j)$$

The ML estimate of the common sdf $f_h(\omega_j)$ under H_o is the periodogram average:

$$\hat{f}_h(\omega_j) = \frac{1}{T} \sum_{t=1}^T I_{th}(\omega_j)$$

and the corresponding maximized value of the log likelihood is:

$$\log L_o = -THJ - T \sum_{h=1}^H \sum_{j=1}^J \log \hat{f}_h(\omega_j)$$

Finally, the LR statistic for testing H_o against H_1 is:

$$LR = 2(\log L_1 - \log L_o)$$

which, under H_o , has an approximate chi squared distribution with degrees of freedom given by $(T-1)HJ$.

The same general approach can be used to test for spatial non-stationarity. Let X_{tk} , $t = 1, 2, \dots, T$ be a collection of T independent time series each of length n recorded at locations $k = 1, 2, \dots, K$ at times $t = 1, 2, \dots, T$ and let $f_{tk}(\omega_j)$ be the unknown sdf at location k at time t . Interest centers on testing the null hypothesis $H_o : f_{tk}(\omega_j) = c_{tk} f_t(\omega_j)$ that at each time the sdf's at the different locations are the same up to a multiplicative scaling against the general alternative hypothesis H_1 that they are not. For definiteness, under H_o , take $c_{t1} = 1$ for all t .

As before, the maximized value of the log likelihood under H_1 is:

$$\log L_1 = -TKJ - \sum_{t=1}^T \sum_{k=1}^K \sum_{j=1}^J \log I_{tk}(\omega_j)$$

where $I_{tk}(\omega_j)$ is the value of the periodogram at time t and location k for Fourier frequency ω_j . Maximizing the log likelihood under H_o must be done numerically. In doing so, it is helpful to note that for fixed values of the scaling parameters $c_1 = 1, c_2, \dots, c_K$, the ML estimate of $f_t(\omega_j)$ is the weighted average:

$$\hat{f}_t(\omega_j) = \frac{1}{K} \sum_{k=1}^K (I_{tk}(\omega_j) / c_k)$$

As before, the LR statistic for testing H_o against H_1 is:

$$LR = 2 (\log L_1 - \log L_o)$$

where $\log L_o$ is the numerically maximized log likelihood under H_o . Under H_o , LR has an approximate chi squared distribution with degrees of freedom given by $T(KJ - (K + J - 1))$.

REFERENCES

Dzhaparidze KO (1986) Parameter estimation and hypothesis testing in spectral analysis of stationary time series. Springer-Verlag, New York