Diel and Seasonal Variability in Kelp Forest Soundscapes Off the Southern California Coast

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Soundscape ecology is a relatively new field that can provide insights into the structure and health of marine habitats. Though this field is growing, the acoustics of many marine habitats, including the giant kelp forests off Southern California, remain poorly studied. Here, we examine the diel and seasonal periodicity of kelp forest soundscapes within a protected and unprotected site off San Diego, CA. Singular value decomposition was used to identify frequency bands of interest, enabling tracking of these bands through seasons to examine their variability. Four frequency bands were identified: (1) 60–130 Hz, which encompassed a putative fish chorus, (2) 300–500 Hz, which encompassed a different putative fish chorus, (3) a band that encompassed humming generated by Plainfin Midshipmen Porichthys notatus (fundamental frequency: 85–95 Hz, and two subharmonics 175–185 Hz and 265–275 Hz), and (4) a band that encompassed the snaps of snapping shrimps from 2.5 to 7.5 kHz. Overall, kelp forest soundscapes exhibited diel and seasonal variability. In particular, the two putative fish choruses dominated the dusk soundscapes during late spring and summer, and the Midshipmen hums persisted throughout nights in summer. Snapping shrimp sounds exhibited stereotypic crepuscular activity, with peaks in acoustic energy in the 2.5–7.5 kHz band occurring at dusk and dawn. In addition, vessel noise was identified and found to exhibit strong seasonal and spatial variation. Vessel noise was greatest during August and September at the protected site and was generally lower during the winter and spring months. These findings help establish reference acoustic indices for the kelp forests off Southern California, within and outside of a protected area, and can provide resource managers with information on how well a marine reserve protects a species of interest, as well as the putative human visitation of these protected areas.

Keywords: marine soundscapes, kelp forests, boat noise, fish chorus, soundscape metrics, marine bio-acoustics

INTRODUCTION

Sound sources of biological, geophysical, and anthropogenic origin all contribute to an underwater soundscape (Pijanowski et al., 2011). Marine soundscapes, like the habitats from which they emanate, are dynamic and vary over time and space (Radford et al., 2008; McWilliam and Hawkins, 2013; Monczak et al., 2019), and even adjacent habitats can exhibit distinct soundscapes
Many fish produce sounds during courtship, spawning, and feeding activities (Connaughton and Taylor, 1996; Gilmore, 2002; Mann et al., 2009; Wilson et al., 2020). Fish have evolved the ability to produce sounds. White Seabass, Atractoscion nobilis, form spawning aggregations in Southern California and produce pulse trains, drumrolls and thuds at low frequencies (around 70–80 Hz) during courtship (Aalbers and Drawbridge, 2008). Giant Sea Bass, Stereolepis gigas, produce "boom" calls in association with aggressive behavior toward other Giant Sea Bass (Clark and Allen, 2018). Multiple different choruses of unknown fish species have recently been reported offshore of kelp forests in Southern California (Pagniello et al., 2019). Many of the fish in the kelp forests, such as the iconic Garibaldi, Hypsypops rubicundus, and Barred Sand Bass, Paralabrax nebulifer, are members of the Pomacentridae and Serranidae—families with many sound producing members (e.g., Rice and Lobel, 2003; Parmentier and Diogo, 2006; Mann et al., 2009; Nelson et al., 2011).

Anecdotal evidence from SCUBA divers suggests that these species may produce calls supporting further research efforts into the identification of biological sources of kelp forest sounds.

Unfortunately, the proximity of kelp forests to the coast exposes them to direct human impacts, including pollution, increased sedimentation, and overfishing (Konar and Roberts, 1996; Jackson et al., 2001; Foster and Schiel, 2010). For example, two kelp forests, one offshore of Los Angeles, California and the other offshore of San Diego, California, collapsed likely due to the confluence of extreme environmental factors and rapid coastal urbanization (Foster and Schiel, 2010). Increased regulations on sewage outflows and better treatment techniques have facilitated the recovery of some forests, yet few have returned to their full extent (Shiel and Foster, 2015). In addition to land-based pollution, historical overfishing of coastal ecosystems can lead to trophic imbalances that allow herbivores to overgraze primary producers (e.g., the near-extirpation of sea otters from the Aleutian Islands allowed sea urchins to graze away kelp forests; Estes and Palmasano, 1974). These and other anthropogenic effects on kelp forests will likely be exacerbated as global oceanic temperatures rise and the concomitant decrease in nutrients available for growth. These changes are of particular concern along the coast of Southern California where kelp forests are stressed by decadal and secular changes in temperature and associated nutrient conditions (Parnell et al., 2010).

As continued research reveals the interplay between marine soundscapes and ecological processes (e.g., larval recruitment, Lillis et al., 2013 and community dynamics, Nedelec et al., 2015), it is becoming increasingly important to monitor and conserve coastal soundscapes—particularly the soundscapes of ecologically and economically significant communities such as kelp forests. Thus, the goals of the present study were to examine the spatial and temporal variability in soundscapes of two kelp forest areas, one inside and the other outside of a Marine Protected Area (MPA), off the coast of La Jolla, CA to address the following questions:

1. What sounds dominate the kelp forest soundscape?
2. How do kelp forest soundscapes vary temporally and spatially?
Can these soundscapes be used to estimate anthropogenic use in the kelp forest?

MATERIALS AND METHODS

Study Location and Passive Acoustic Data Collection

Our study was in the La Jolla kelp forest, off San Diego, California, United States. The nearshore habitat is dominated by a hard-bottom substratum scattered with sand and cobble patches (Parnell, 2015). The kelp forest varies in size over time, but its fullest extent is ∼8 km long and ∼1.5 km wide (Parnell et al., 2006). Many species of algae compose the kelp forest community; however, the dominant biogenic engineer of vertical structure is the giant kelp Macrocystis pyrifera, which occurs from ∼8 to 24 m depth in the La Jolla kelp forest (Dayton et al., 1992). The kelp forest is bounded to the north, south, and offshore edges by sand (Parnell et al., 2006). The extreme northern portion of the forest (∼0.08 km²) is protected by the San Diego-La Jolla Ecological Reserve (Parnell et al., 2006), and the southern half of the kelp forest is protected by the South La Jolla State Marine Reserve (SLJSMR).

Two sites were selected along the offshore edge of the kelp forest at ∼25 m depth. The northern site (Site A) lies north of the SLJSMR on a flat pavement reef that rises out of sand from ∼28 m depth, whereas the southern site (Site B) lies approximately 2 km south of Site A within the SLJSMR in an area of rocky rubble reef (Figure 1). At each site, a Wildlife Acoustics SM3M passive acoustic recorder (Wildlife Acoustics, Inc., Maynard, MA) was deployed to record the ambient soundscape. Each unit had an HTI-96-MIN omnidirectional hydrophone (High Tech, Inc., Long Beach, MS; sensitivity: −165 dB re: 1 V/μPa, frequency response: 2 Hz–30 kHz) and recorded uncompressed WAV files at 48 kHz, 16-bit depth. Recorders were calibrated using pure sine wave signals from a signal generator across the range of recording frequencies. The response was flat (67.15 ± 1.02 dB re: 1 μPa) over the 50 Hz–20 kHz frequency range. SCUBA divers deployed acoustic recorders from May 2015 to December 2017 at each site. Hydrophone recorders were programmed to record continuously during deployments. Recorders were deployed six times at Site A (totaling 183 days of recording) and eight times at Site B (totaling 198 days of recording).

Acoustical and Statistical Analyses

Digital acoustic recordings were analyzed in MATLAB 2016b (Mathworks, Inc., Natick, MA) using the Triton software package (Wiggins et al., 2010) and other custom scripts. We created long-term spectral averages (LTSA) from each deployment’s data by calculating 1 min spectral averages with a 1 Hz frequency resolution over the duration of the deployment (Figure 2). These LTSA were used to visualize the multi-terabyte acoustic dataset, as well as to identify frequency bands of interest and to identify vessel noise that occurred within the recordings.

To identify frequency bands of interest from the LTSA, each LTSA was treated as a large matrix, where the columns of the matrix represent each minute of a recorder deployment, each row represents a specific frequency, and each cell value (e.g., the value found in the first row and first column) represents the spectral average at that minute and frequency. These matrices (one matrix for each deployment) were arranged together to create a single, large data matrix to allow for comparison of seasonal variability across both sites. Singular value decomposition (SVD) was used to decompose this data matrix into three distinct matrices: its left singular matrix (U), its singular value matrix (S), and its right singular matrix (V). After decomposition, the original data can be recovered by multiplying $U^T S V^T$. Lower rank approximations (SVD components) that highlight important elements of the original data can be generated by multiplying a singular column in $U (u_i)$ by the singular value in $S (s_i)$ by the transpose of the singular column in $V (v_i)$. For these analyses, the data reconstruction from the first four SVD components explained > 97% of the variation and were used to identify elements of interest within the LTSA data (Figure 3).

From the decomposition of the LTSA data, each SVD component was visually examined to determine which sounds were dominant in that component, and four frequency bands of interest were identified for further analysis. Two bands of interest were associated with putative fish calls and spanned the 60–130 Hz and 300–500 Hz ranges. The third frequency band was associated with long bouts of a tonal sound focused in the 85–95 Hz range, and its harmonics in the 175–185 Hz and 265–275 Hz ranges produced by Plainfin Midshipman, Porichthys notatus (Ibara et al., 1983). The fundamental frequency component of this band was removed from the previous 60–130 Hz band for these analyses. Dissimilar to the previous three frequency bands of interest, the final band of interest was associated with the crackle and pop of snapping shrimp sounds. In our recordings, the primary energy of snapping shrimp snaps occurred between 2.5 and 7.5 kHz. For each frequency band, acoustic energy across the band was root-mean-square averaged for each minute and then z-score standardized to produce a time series for that frequency band for the duration of the deployment of its standardized spectral levels. These time series allow for direct comparison of temporal variation (e.g., diel and seasonal variation) and spatial variation (i.e., differences between Sites A and B) within a frequency band, as well as comparison among the various frequency bands over space and time. To quickly discern patterns within these time series, the time series were rearranged into a matrix, where each row represents a single day and each column represents a 1 min interval within the day (1,440 total 1 min intervals within 1 day). SVD was again used to decompose this matrix and reconstruct lower rank SVD components to explore how these bands of interest vary. For each frequency band, the coefficient of deviation (CD) was calculated to compare the variability of the different frequency bands of interest by dividing the median absolute deviation (MAD) of the first two left singular columns of the decomposition by the absolute value of the median of the first two left singular columns (median) (Marmolejo-Ramos and Ospina, 2019), and the CD was used to determine the variability of the frequency bands. The decompositions of these data were visualized by plotting the first two columns of the left singular matrix as scatterplots (i.e., U1 on the x-axis versus U2 on the y-axis), and k-means clustering
FIGURE 1 | State of California (inset) with study area highlighted by red box. Site locations (indicated by dark blue points) off the coast of Southern California. Marine protected areas are denoted by light gray polygons; blue lines are 10-m isobaths; kelp forest extent is indicated in green.

FIGURE 2 | Example long-term spectral average plots for a 24 h period (midnight August 21st to midnight August 22nd, 2017) at Site A (A) and Site B (B). The two putative fish choruses are highlighted by the yellow and red ovals.
via gap analysis was performed to identify any clusters within the data related to site (A or B) or month.

In addition to these spectral analyses, vessel noise occurrence at each site was manually logged from the LTSAs and acoustic recordings (Supplementary Figure 1). The LTSAs were used to efficiently scan through the acoustic data to find vessel noise; once an occurrence of vessel noise was found, a spectrogram of the acoustic timeseries was created (4,096-point Blackman window; 4,096-point FFT; 90% window overlap) to more accurately estimate the duration of the noise event. All vessel noise logging was performed in MATLAB using the Triton software package with Logger feature (Wiggins et al., 2010). The start and end times of each occurrence were marked within the spectrogram and logged into an Excel database. Counts of vessel noise occurrence for each day during deployments were calculated for each site, and a generalized linear model with a gamma distribution was

FIGURE 3 | Reconstructed time series from acoustic recordings at Site A (A,C,E,G) and Site B (B,D,F,H) using the first four SVD components. Time-of-day is indicated along the x-axis, and calendar day is along the y-axis. Color indicates standardized power spectral density, where cooler colors are quieter and warmer colors are louder. Black lines indicate breaks in deployments. (A,B) Show SVD component 1 that corresponds to the 60–130 Hz frequency bands (choruses are highlighted by arrows). (C,D) Show SVD component 2 that corresponds to the 300–500 Hz frequency band. (E,F) Show SVD component 3 that corresponds to the midshipman frequency band. (G,H) Show SVD component 4 that corresponds to the snapping shrimp frequency band.
used to model the relationship between these counts and the following factors: site (A or B), month, and their interaction in R (R Project for Statistical Computing). Graphs of residuals versus fitted values and quantile-quantile plots were visually inspected to ensure model assumptions were met. Throughout this text, the term “sound” is used to refer to sounds produced by animals, whereas the term “noise” is used to refer to signals emanating from human activity.

RESULTS

Long-Term Spectral Averages and Time Series Comparisons

Singular value decomposition analysis of the LTSAVs generated for each deployment revealed a complex kelp forest soundscape that varied spatially and temporally, containing acoustic components of both biotic and anthropogenic origin. The 60–130 Hz band described the occurrence of short “drum beat” calls consisting of either a single beat or a few repeated beats. These calls became so frequent that they overlapped, creating a biological chorus in the spring and summer months that dominated the 60–130 Hz frequency range (red oval, Figure 2), and on days when this chorus occurred it persisted 6–12 h beginning in the afternoon hours and extending well into night. The time series of this frequency band from both sites exhibited broad dusk peaks (Figure 4) during the summer months. The dusk spectral peaks for this frequency band were absent from the late fall to early spring acoustic recordings. There were higher standardized spectral levels and increased chorus duration of the 60–130 Hz band at Site B, indicating that the chorusing fish are likely closer to the recorder at Site B (and thus possibly within the SLJSMR no-take area).

The 300–500 Hz band measured occurrence of fish calls that exhibited pulsed calls with peak energy between 300 and 400 Hz, with less energy extending up to 500 Hz. Similar to the 60–130 Hz band, these calls formed a biological chorus that dominated the dusk soundscape in this frequency band (yellow circle, Figure 2). Though these choruses occurred almost every evening for up to 2 weeks during the late spring and summer, they were relatively short in duration, lasting only 2–3 h and peaking at sunset. Time series from both Site A and Site B exhibited quick-onset dusk peaks (Figure 4), and a short-lived energy increase was easily discernible in the LTSAVs on nights when chorusing occurred (Figure 2). These dusk choruses were absent from acoustic recordings from the late fall to early spring, and when present in the late spring and summer they did not appear to vary with lunar phase.

In addition to the two putative fish choruses above, the long tonal bouts, likely calls produced by the Plainfin Midshipman (Ibara et al., 1983), occurred primarily from May through August and typically began around sunset, persisted throughout the night, and ended around sunrise (Figure 4). The time series of spectral levels from snapping shrimp snaps displayed strong crepuscular periodicity, rising steeply to a dusk peak, then decreasing slightly to a level above that of daytime, and often rising again to a smaller peak at dawn (Figure 4). The diel periodicity of snapping shrimp was consistent across months.

The singular value decomposition of the timeseries of these frequency bands revealed that the 300–500 Hz frequency band exhibited the highest coefficient of deviation (CD = 17.94), indicating a high degree of variability in the acoustic energy within this band. The 60–130 Hz frequency band and the frequency band likely associated with midshipman calling exhibited lower CDs (3.46 and 4.10, respectively), whereas the frequency band associated with snapping shrimp snaps exhibited much tighter grouping with less variability (CD = 1.85; Figure 5), which indicates a consistent pattern in the acoustic
energy recorded within this band and is represented by a
tighter cluster of points with less spread. *K*-means clustering
did not reveal any groups within the data related to site or
month, which is supported by the scatterplots of the first and
second columns of the left singular matrix; there are no clear
groupings related to season (indicated by color of the points on
the plots).

**Vessel Noise Occurrence**

Overall, 13,424 occurrences of vessel noise were counted over
465 days of hydrophone deployments; of those 6,007 vessel noise
occurrences were at Site A, and 7,417 at Site B. The maximum
count of vessel noise occurrence in 1 day was 138 at Site B on
August 25th 2017, and the minimum count of a single vessel
per day happened three times, all at Site B: June 12th 2017,
October 29th 2017, and November 30th 2017. Only 3% (181
instances) of vessel noise occurred during night at Site A, and
only 1.5% (114 instances) of vessel noise occurred at night at
Site B. The generalized linear regression of daily vessel noise
occurrence showed a significant interaction between site and
month \( F(9, 445) = 5.04, p < 0.001 \), as well as a significant effect
of month \( F(9, 445) = 15.89, p < 0.001 \); however, site did not
have a significant effect \( F(1, 445) = 0.38, p = 0.54 \) (Table 1).
Counts of vessel noise occurrence at both sites followed the same
trend; counts were lower during fall and winter months, they
gradually rose throughout the spring, and peaked during the
summer months (Figure 6). Counts of vessel noise occurrence at
Site B during August, September, and October were significantly
higher than counts of vessel noise occurrence at Site A during
those months (Figure 6).

Though site did not significantly affect the daily counts of
vessel noise occurrence, the duration of noise events (i.e., how
long a single vessel noise occurrence could be heard within the
recordings) was greater at Site A \((8.11 \pm 0.13 \text{ min}; \text{ mean } \pm \text{ st.}
err)\) than at Site B \((7.77 \pm 0.19 \text{ min})\). This was particularly evident
during August when the average vessel noise duration at Site A
was \(13.15 \pm 0.28 \text{ min}\), whereas vessel noise duration at Site B was
\(8.15 \pm 0.08 \text{ min}\).

**TABLE 1** Analysis of deviance of the generalized linear regression model
describing the relationship between month and site on vessel noise occurrence
within acoustic recordings.

|       | DF | Deviance | Resid. DF | Resid. Dev. | F     | p     |
|-------|----|----------|-----------|-------------|-------|-------|
| Site  | 1  | 0.09     | 454       | 153.27      | 0.38  | 0.54  |
| Month | 9  | 33.94    | 455       | 153.36      | 15.89 | 2.48e-26 |
| Month * site | 9 | 10.77 | 445 | 142.49 | 5.04 | 7.79e-7 |

*Indicates the interaction between these two predictors.
DISCUSSION

This study represents the first multi-season study of kelp forest soundscape dynamics to our knowledge. The results showcase the complex and dynamic nature of kelp forest soundscapes. During the late spring and summer, choruses of two putative fish calls dominated the dusk soundscapes. One chorus, with much of its energy between 300 and 500 Hz, was short-lived each night, lasting only about 2 h around dusk. This chorus matches descriptions reported in Johnson (1948) that described a chorus of harsh “croaks” that built to peak just after sunset and lasted for 2–3 h just off the Scripps Institution of Oceanography pier and reported in Gottesman et al. (2020) from kelp soundscapes off the Channel Islands, CA. This chorus was likely produced by a croaker species found off the coasts of Southern California (Black Croaker, *Sciaena saturna*; Spotfin Croaker, *Roncadro stearnsi*; or, Yellowfin Croaker, *Umbrina roncador*). The lower frequency chorus, with most of its energy between 60 and 130 Hz, persisted much longer throughout days it was present, beginning in the afternoon and lasting well into night. Anecdotal evidence by divers in the kelp forest near our sites suggest that the Lingcod, *Ophiodon elongatus*, may produce these calls. These two choruses appear to be the same as Chorus I and Chorus IV reported by Pagniello et al. (2019) in the same area.

In addition to these two putative fish choruses, long bouts of calling likely produced by the Plainfin Midshipman were present in our recordings. The timing of these bouts was consistent with their mating season. During the late spring and summer, the times when we recorded their long nighttime calling bouts, male Plainfin Midshipmen prepare nests in burrows under rocks, where they shelter and produce long “hums” to attract females; once a mate is found, female Plainfin Midshipman attaches eggs inside the burrow and leaves the male to care for and guard the brood (Ibara et al., 1983).

Snapping shrimp in the family Alpheidae are common throughout the world’s tropical and temperate waters and are perhaps the most ubiquitous sources of underwater sound in shallow waters (Au and Banks, 1998; Radford et al., 2008; Bohnenstiehl et al., 2016). These shrimps possess enlarged chelae, which, when rapidly closed, create a cavitation bubble that produces a loud “pop” upon its collapse (Versluis et al., 2000). In our acoustic data, snapping shrimp sound displayed the stereotypic crepuscular periodicity, peaking at dawn and dusk, as seen in other regions worldwide (Lammers et al., 2008; Radford et al., 2008; Bohnenstiehl et al., 2016; Butler et al., 2016). However, unlike some previous studies (e.g., Bohnenstiehl et al., 2016), we did not observe much seasonal variation in acoustic energy in the snapping shrimp frequency band (i.e., 2.5–7.5 kHz). Previous studies on snapping shrimp snap rate (i.e., counts of snapping shrimp snap over a given time period) and acoustic energy from snapping shrimp have correlated those metrics with water temperature, length of day, and dissolved oxygen (Watanabe et al., 2002; Bohnenstiehl et al., 2016), so our results could indicate more constancy in the environmental conditions during our deployments. For example, divers who deployed the acoustic recorders at our sites reported water temperatures near the sea floor varied between 8 and 15°C, whereas Bohnenstiehl et al. (2016) reported water temperatures between 10 and 30°C that...
influenced snapping shrimp snap rate and water pressure levels. Thus, the lower variability in water temperatures near the sea floor around our sites might reduce the variability in snapping shrimp sounds as well.

We also found extensive anthropogenic noise within our acoustic data. Noise from small vessels pervaded the soundscapes of both sites. The kelp forests off Southern California support thriving recreational and commercial fisheries (e.g., Spiny Lobster Panulirus interruptus fishing from late September to late March), and their proximity to a dense human population along the coast makes them vulnerable to anthropogenic stress (Foster and Schiel, 2010). The infusion of anthropogenic sounds within natural soundscapes could have deleterious consequences on the ecological community. For example, Nichols et al. (2015) found that Giant Kelpfish, Heterostichus rostratus, exhibited an acute stress to the playback of small vessel noise, and Bruintjes and Radford (2013) found that playback of small vessel noise altered the social behaviors of the cooperatively breeding Cichlid Neolamprologus pulcher. Furthermore, vessel noise has been found to alter orientation and settlement behaviors in coral reef fish larvae (Holles et al., 2013), as well as prevent coral planulae from selecting preferred settlement habitat (Lecchini et al., 2018). Though one of our sites was within a protected area, the occurrence of vessel noise was greater at this site during late summer months—likely due to its proximity to the inlet to Mission Bay, a busy harbor in San Diego County. Thus, the noise from small vessels still has the potential to affect the ecology of the kelp forest in subtle ways, even within protected areas.

**Marine Soundscapes as a Monitoring Tool**

Ecoacoustics, and its many subdisciplines like soundscape ecology, is a developing field of marine ecological research whose goal is to explore how underwater soundscapes vary over space and through time (Radford et al., 2008, 2010; Staaterman et al., 2014; Butler et al., 2016), how animals use underwater sounds for orientation and navigation (Stanley et al., 2012; Lillis et al., 2013; Chapuis et al., 2019), and how anthropogenic contributions to marine soundscapes alter the soundscapes themselves, as well as other ecological processes relying on soundscapes (Holles et al., 2013; Marley et al., 2017; Lecchini et al., 2018). Recently, “ocean sound” has been designated as an essential ocean variable by the Global Ocean Observing System committee (Tyack, 2017), and passive acoustic monitoring stations are collecting baseline measurements for and monitoring changes in marine soundscapes (e.g., Haver et al., 2018). Our study employed passive acoustic monitoring and multivariate data analysis techniques (namely, singular value decomposition) to explore the soundcape of the kelp forests off La Jolla, CA. These data and techniques can be used as baseline measures for these kelp habitats for comparison to future studies on how coastal ecosystems are changing.

As the need to protect marine resources and ecosystems increases with human exploitation and climate change, more governing entities such as local or state governments and fisheries management councils have begun implementing area closures and usage regulations to conserve species or habitats of concern. Recognizing the imperative to protect its economically and ecologically valuable marine resources, the state of California passed the Marine Life Protection Act in 1999 and established a network of marine protected areas (MPAs) in 2012, which protects roughly 1,920 km² (16%) of California’s jurisdictional waters (Marine Protected Area Monitoring Action Plan, Oct. 2018). Long-term passive acoustic recorders used in this study helped us catalog the soundscapes within the South La Jolla State Marine Reserve (part of California’s MPA network), as well as the soundscape just outside the reserve. This reserve was established in 2012 to protect a portion of the most diverse habitat in the southern coastal region of California [California Dept. of Fish & Wildlife (CDFW 2016)], and these recordings can serve as reference indicators of biophony and vessel activity to which future acoustic efforts can be compared.

The two putative fish choruses that occurred in our recordings during the late spring and summer months are likely related to spawning. Future acoustic monitoring efforts can explore whether the protection provided by the marine reserve has led to increased chorusing (and, thus, possibly increased spawning), and also how these choruses respond to long-term environmental changes or physical processes (e.g., the timing of the choruses might move earlier in the year as coastal waters warm). The anecdotal evidence of possible species producing the 60–130 Hz chorus—Lingcod—was identified as species likely to benefit from the implementation of the marine reserve. Lingcod have traditionally supported recreational and for-hire commercial fisheries, however, fishery-dependent and fishery-independent data analysis indicated significant population declines (CDFW Status of the Fisheries Report 2008). These declines led to the enactment of new size and bag limits for Lingcod in 2019. In addition, though the Croaker species typically inhabit nearshore sandy bottom along the Southern California coast, evidence suggests that these species move to deeper water to form spawning aggregations in the summer months (Pondella et al., 2008; Williams et al., 2011).

Studies of the life histories of these fish have provided fisheries managers useful information for their future management, as well as suggested that marine protected areas might work well to bolster their stocks. Lingcod exhibit spawning behavior that lends itself to protection via spatial closures. Male lingcod defend territories in rocky reef areas, where a female will lay a nest of up to 500,000 eggs in rock crevices or under rocks (King and Withler, 2005). Females leave the spawning area, but males remain and actively defend their nests. Additionally, male Lingcod exhibit extreme nest site fidelity, reusing the exact same territory and nest site through multiple years. Therefore, if protecting Lingcod spawning locations is a management priority, then establishing an MPA in an area where spawning has previously occurred is crucial.

Though passive acoustic methods show promise as a non-invasive method to monitor ecosystem health (e.g., Parks et al., 2014; Mooney et al., 2020), they are not without limitations. Though sound has the potential to convey biologically relevant information long distances through seawater (Urick, 1983), only soniferous species will occur in the acoustic record. Furthermore,
sounds for only a small portion of fish species that occur in
the world’s oceans have been cataloged (Rountree et al., 2006).
Studies that have cataloged fish sounds have generally recorded
fish housed within tanks (e.g., Ibara et al., 1983; Hawkins and
Amorim, 2000; Širović and Demer, 2009; Allen et al., 2020);
however, linking recorded sounds with the behavioral context of
the fish that produced them is difficult to do both in laboratory
settings (due to behavioral changes of fishes within tanks) and
in situ. Recent studies have begun pairing passive acoustic
recording with time-lapse photography or videography (e.g.,
Pagniello et al., 2018; Mouy et al., 2020). Associating specific
underwater sounds, the species that produces those sounds, and
the behavioral context in which those sounds are made will create,
in essence, a database to allow efficient determination of species
that are present within recordings and their behavioral state
(Hastings and Širović, 2015).

In addition to the limited number of marine sounds attributed
to individual fish species, another challenge is application of long-
term acoustic recordings to abundance or density estimation,
rather than solely presence/absence. For example, most ecological
studies that employ passive acoustic recorders use only a single
element (i.e., a single hydrophone). Studies using multi-element
arrays can leverage the physics of acoustic propagation to
determine sound source locations (Spiesberger and Fristrup,
1990; Au and Hastings, 2008) and are becoming increasingly
common in the ecological literature to locate soniferous fishes
(e.g., Locascio and Mann, 2011; Putland et al., 2018; Wilson
et al., 2019) and marine mammals (e.g., Adam et al., 2006,
among many others). Information about location of sources of
underwater sounds can be used to estimate species density (e.g.,
Marques et al., 2013) and spatial use patterns (e.g., Rowell et al.,
2015). Employing multi-element arrays in future studies assessing
the efficacy of marine reserves can help managers determine
whether the reserve was placed in areas used by target species.
For example, anecdotal evidence in this study suggests that
Croaker and Lingcod may produce the two choruses described
here. Lingcod, in particular, exhibit spawning spatial constraint
(i.e., high interannual nest site fidelity; King and Withler,
2005). If choruses were produced by these species, acoustic data
could inform on whether the marine reserve was placed in the
appropriate area to protect sites used by these species.

Singular Value Decomposition and Its
Use to Discriminate Acoustic Features of
Interest
Monitoring biodiversity is essential to tracking the effects of
human-exacerbated global climate change (Tittensor et al.,
2014), and the use of passive acoustics as a cost-effective, non-
invasive way to monitor biodiversity and ecosystem health is
increasing in both the terrestrial and marine realms (e.g., the
U.S. National Parks Service is actively monitoring and managing
soundscapes in terrestrial and marine parks). Though cost-
effective and generally easy-to-use, monitoring of soundscapes
via passive acoustics yields large datasets that preclude manual
analysis. To investigate ecological patterns emerging from
environmental sounds and to create rapid assessments of
biodiversity and ecosystem health, several ecoacoustics indices
have been developed [e.g., Acoustic Complexity Index (ACI);
Pieretti et al., 2011, Acoustic Richness (AR) and Acoustic Entropy
Index (H); Sueur et al., 2008, among others]. However, care
needs to be taken when applying these indices as monitoring
tools, as their link to the actual biodiversity present in habitats
can break down for various reasons (Fuller et al., 2015;
Harris et al., 2016). For example, Gasc et al. (2013) and
Fuller et al. (2015) found that the ACI is not sensitive to
constant, sustained acoustic signals. Therefore, it would not
serve well to detect the nocturnal humming of the Plainfin
Midshipman within our recordings, and also was not well
suited for the forests of New Caledonia and eastern Australia.
Harris et al. (2016) formalized a set of four criteria that would
make an acoustic index successful in marine environments
and tested ACI, AR, and H against visual surveys of reef fish
abundance in New Zealand, finding that ACI and H could be
appropriate for use on temperate reefs to monitor fish diversity.
However, Lyon et al. (2019) found that in tropical back-reef
environments ACI did not correlate with habitat complexity or
fish community structure.

In the present study, we did not implement ecoacoustic
indices, but rather created LTSAs of the acoustic timeseries,
and then decomposed these LTSAs via singular value
decomposition (SVD). Sattar et al. (2016) had proposed the
use of robust principle components analysis, which uses SVD,
and multiresolution acoustic features (MRAF) to automatically
classify fish sounds within large, marine acoustic datasets.
In their study, Sattar et al. (2016) employed robust principle
components analysis to decompose their multiresolution
acoustic features dataset into lower-rank approximations of the
data to aid in feature selection to train a multiclass support
vector machine classifier based on manually annotated data.
In our study, however, we used it as an approach to rapidly
determine frequency bands of interest to track through time in
an objective manner.

Though decomposition of our acoustic dataset identified
frequencies of interest, k-means clustering of SVD components
was unable to discriminate site and temporal patterns of these
frequencies bands (Figure 5). For example, the putative fish
choruses in the 300-500 Hz frequency band exhibited
higher acoustic energy at dusk during the spring and summer
and lower energy during fall and winter (Figures 3C,D), yet
these groupings did not separate via k-means clustering (top
right panel, Figure 5). This could be an artifact of the root-
mean-square averaging and z-score standardization of our
acoustic energy time series; that is, the variability in the raw
acoustic energy might have allowed the k-means clustering
algorithm to identify distinct groupings of data, whereas the
z-score standardization constrained this variability and possibly
limited the effectiveness of the k-means clustering algorithm.
Singular value decomposition underpins many multivariate
analysis techniques and harnessing some of those techniques
to extend the use of SVD and k-means clustering to the types of
analyses presented in this study could aid the clustering
techniques in discriminating distinct groupings. For example,
future research could explore whether multivariate distance
or dissimilarity transforms (e.g., Bray-Curtis; Ricotta and Podani, 2017) aid discriminating groups among times-of-day, seasons, and sites.

In addition, future studies of kelp forest soundscapes need to further explore the spatial variability of these soundscapes. The sites we selected for our study were relatively close together (only ~2 km apart) and exhibited similar diel and seasonal variability. Our study employed single-hydrophone recorders, yet future studies could use arrays of two or more hydrophones to allow sound source localization (e.g., Locascio and Mann, 2011) and restrict the analysis of spatial differences in soundscapes to sounds that occur within a set distance from the hydrophones. This technique would allow researchers to explore small-scale spatial variability in underwater sounds, while also exploring where those sounds were produced and whether there is a relationship between sounds and occurrence of sound-producers and the habitat.

Still, this technique can aid biodiversity monitoring efforts to identify robust features of interest to monitor over time that relate to the biology of species targeted for conservation or the ecology of imperiled habitats. Though ecoacoustic indices such as ACI or AR can be rapidly applied to large acoustic datasets, care should be taken when applying ecoacoustic indices to large datasets from disparate and distinct marine environments. Many factors can influence ecoacoustic metrics (e.g., sound source levels, spatial and temporal scales), thus the end user must be aware of how these factors might affect the relationship between acoustic and biological diversity (Mooney et al., 2020). Augmenting ecoacoustic indices with objective acoustic features that show clear links to the unique ecology of habitats or areas of concern can help monitoring efforts and ultimately aid resource managers to more accurately assess the health of ecosystems and track their responses to disturbance as the world’s oceans continue to change.

Here, we explored how the soundscapes emanating from two sites off La Jolla, CA, vary spatially and temporally. We applied singular value decomposition and k-means clustering to identify acoustic bands of interest in the data. Two fish choruses dominated the dusk soundscape during the late spring and summer, and the din of snapping shrimp exhibited stereotypic crepuscular peaks. Vessel noise was also often heard in the kelp soundscapes, and our analysis indicated seasonal trends in vessel activity, peaking in summer months and declining to winter. These analyses revealed a complex and highly variable soundscape, comprised of biological and anthropogenic sounds, and provide valuable baseline for researchers and natural resource managers for future comparisons.

DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

AUTHOR CONTRIBUTIONS

JJ, PEP, and AS designed the experiment and secured funding. JB, CP, and PEP conducted field work. JB and CP performed analyses with guidance from JJ, PEP, and AS. JB wrote the manuscript, with significant input from CP, JJ, PEP, and AS. All authors contributed to the article and approved the submitted version.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fmars.2021.629643/full#supplementary-material

Supplementary Figure 1 | Example spectrogram of vessel noise occurring within acoustic recordings of kelp forests off Southern California.

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**Conflict of Interest:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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