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Original Article

Wideband (15–260 kHz) acoustic volume backscattering spectra of Northern krill (*Meganyctiphanes norvegica*) and butterfish (*Peprilus triacanthus*)

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Measurements of acoustic backscatter made over a wide frequency band have the potential for improved classification relative to traditional narrowband methods, by characterizing more fully the frequency response of scatterers. In January 2014, five wideband transceivers [Simrad EK80 Wideband Transceivers (WBTs)] and split-beam transducers with nominal centre frequencies of 18, 38, 70, 120, and 200 kHz were used to collect acoustic data spanning a nearly continuous 15–260 kHz bandwidth. The acoustic samples were from ca. 2 m below the surface to the seabed in an area along the US continental shelf break. Bottom trawls and zooplankton nets were also used to sample scatterers contributing to selected features of the acoustic backscatter. Measurements of frequency-dependent volume backscattering strength (i.e. volume backscattering spectra) from aggregations of euphausiids (mostly Northern krill, *Meganyctiphanes norvegica*) clearly resolved the transition from Rayleigh to geometric scattering, consistent with modelled backscatter from the type and length of animals sampled with bongo nets. Volume backscattering spectra from aggregations dominated by butterfish (*Peprilus triacanthus*) revealed a frequency response that was suggestive of superimposed scattering by soft tissue and bone. Backscatter predicted by Kirchhoff ray mode models of butterfish corresponded to trends in the measured spectra, supporting the assumption that acoustic scattering by butterfish is dominated by soft tissue and vertebrae.

Keywords: acoustic backscatter, broadband, fish, zooplankton.

Introduction

Measurements of acoustic backscatter made over "wide" bandwidth provide more data on the frequency response of scatterers than over "narrow" bandwidth, offering the potential for improvements in species categorization, signal-to-noise ratio (SNR), and range resolution (Ehrenberg and Torkelson, 2000). We define wide bandwidth (i.e. wideband) acoustic data as that resulting from the transmission and reception (i.e. backscatter) of signals consisting of a single continuous band (i.e. broadband), multiple broadband signals, multiple narrow bandwidth (i.e. narrowband) signals (e.g. Holliday *et al.*, 1989), or a combination of broadband and narrowband signals. The use of wideband data in underwater acoustics dates back to at least World War II where explosive sources (e.g. Love, 1975) provided broadband sound at lower (10–100s of Hz) frequencies. Military and oil exploration led the development of safer broadband systems, with electronics that were more reliable and consistent, but applications to fisheries research and management have been slow to adopt the technology. Nevertheless, wideband systems have been custom-built for specific *in situ* (e.g. Zakharia, 1996; Nero *et al.*, 2004; Lavery *et al.*, 2010) or *ex situ* (Simmonds and Armstrong, 1990; Kjægaard *et al.*, 1990; Conti and Demer, 2003; Stanton *et al.*, 2003; Reeder *et al.*, 2004; Conti *et al.*, 2005; Lundgren and Nielsen, 2008) applications.

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Although custom systems are favoured for prototype development, commercial-off-the-shelf (COTS) systems are advantageous because maintenance, parts, and components are readily available, and software can be developed through commercial or open-source avenues. Wideband COTS systems for fisheries applications have been available for more than a decade (e.g. Ehrenberg and Torkelson, 2000) but their signal bandwidths were only sufficient to improve SNR and range resolution, not characterize volume backscattering spectra. Commercially available systems were modified for wideband target classification using mid-frequencies (1.7-100 kHz) by Stanton and colleagues (Stanton, 2009; Stanton et al., 2010, 2012), and high frequencies (150-600 kHz) by Lavery et al. (2010). Simrad-Kongsberg's June 2015 release of a commercial WBT and control and data logging software (EK80) may stimulate application of wideband echosounding to fisheries and ecosystem surveys.

We present wideband (15–260 kHz) acoustic volume backscattering spectra from aggregations of euphausiids (specifically Northern krill, *Meganyctiphanes norvegica*) and butterfish (*Peprilus triacanthus*) located along the edge of the continental shelf in the US mid-Atlantic Bight. These measurements are compared with net samples and theoretical models of acoustic backscatter to confirm to our interpretation of the data. In addition, we suggest notation for broadband volume backscatter across a frequency band and the approximation of volume backscattering at the nominal centre frequencies, which are useful for visualizing broadband data.

Material and methods

Data were collected during 8–12 January 2014 on the Fishing Vessel *Karen Elizabeth* (24-m western-rigged stern trawler) near the shelf break south of Cape Cod and east of Long Island, USA (Figure 1). The sampling equipment included a 4-seam bottom trawl (4SBT; Politis *et al.*, 2014), bongo nets (Posgay and Marak, 1980), narrowband Simrad ES60 and EK60, and wideband Simrad EK80 echosounders.

Echosounders with hull-mounted transducers

The ES60 transducers (Simrad Combi-B, single-beam 50- and 200- kHz with $10 \times 16^{\circ}$ and $7 \times 7^{\circ}$ beam widths, respectively; and a Simrad ES38-B, split-beam 38-kHz with $7 \times 7^{\circ}$ beam width) were mounted on the hull of *Karen Elizabeth*. These systems were calibrated with a 38.1-mm diameter sphere made from tungsten carbide with 6% cobalt binder (WC) (Foote *et al.*, 1987), immediately following the wideband calibrations (described below), in Providence, RI, USA. The "triangle-wave" error, inherent in ES60 data, was removed (Demer *et al.*, 2015) before further processing using commercial software (Echoview v6+, Echoview Software Pty Ltd, Hobart, Tasmania).

Echosounders with towed transducers

The wideband echosounder system was composed of six WBTs (hardware Rev. 0.13), EK80 software (version 1.2.5087.17756), and six split-beam transducers with nominal centre frequencies (f_{nom}) of 18, 38, 70 120, 200, and 333 kHz (Simrad ES18, ES38-B, ES70-7C, ES210-7C, ES200-7C, and ES333-7C, respectively). Data from the 333-kHz WBT were too noisy to use and are not presented here. Each WBT can transmit and receive signals from10 to 500 kHz, but the functional bandwidth is limited, typically to an octave or less, by the operational bandwidth of the

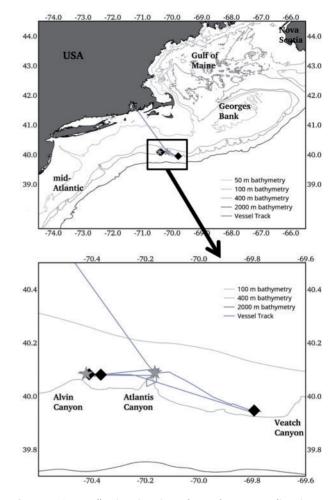


Figure 1. Data collection sites. Stars denote bongo sampling sites and diamonds denote 4SBT hauls.

attached transducer (Table 1). The transducers were mounted on an aluminum deadweight frame (Figure 2) that was deployed near the sea surface from the port quarter of Karen Elizabeth, while drifting. Tow speeds up to 5 kts resulted in transducer instability and increased noise. The broadband signals consisted of 1.024 ms linear chirps with a 10% slope (i.e. "ramp" or "taper") for all bands. At the time of data collection, the transmit pulse slope parameter choices were 5 and 10%, which correspond to the current Simrad terminology of "fast" and "slow", respectively, in the EK80 acquisition software. Generally, the pulses from each WBT were transmitted sequentially, from lowest to highest frequency, with an interval of 500 ms. In addition, data were collected with simultaneous transmissions at all frequencies. Data were collected from 15 to 260 kHz, although post-processing led to gaps in the spectrum (Table 1). Each WBT and transducer pair was calibrated twice. The first calibration was performed prior to the shipboard measurements, using 15-, 21.2-, and 38.1-mm diameter WC spheres, in the tank facility at the University of New Hampshire, USA (www.ccom.unh.edu/facilities/test-tanks/ engineering-tank). The transducers were mounted in the aluminum frame (without tail fin) with the acoustic axes oriented horizontally and the spheres sequentially located at 6- to 8-m range.

Transducer	Transmit/receive frequency band (kHz)	Selected frequency band (kHz)	Transmit power (W)
ES18-11	15-25	15.4-23.3	400
ES38D	25-45	32.0-44.5	400
ES70-7D	45-95	53.0-94.0	500
ES120-7C	95-160	100.0-158.0	125
ES200-7C	160-260	160.0-260.0	75
ES333-7C ^a	260-400	262.0-395.0	25

Table 1. EK80 data acquisition parameters.

The "Transmit/Receive Frequency Band" represents the upper and lower frequency limits of the transmit and receive signal set in the EK80. The "Selected Frequency Band" represents the upper and lower frequency limits that were used for post-processing the broadband data for each channel.

^aLimited amounts of test data were collected from the ES333-7C but are not reported in this article.

These measurements were used to determine the operational bandwidths for each transducer.

At the conclusion of the shipboard measurements, each WBT and transducer pair was calibrated using 21.2- and 38.1-mm diameter WC spheres, dockside in Providence, RI, USA. Calibration procedures followed published methods (Stanton and Chu, 2008; Stanton *et al.*, 2010; Lavery *et al.*, 2010, 2017), and resulted in estimates of frequency-dependent gain, G(f), for each transducer (see Lavery *et al.*, 2017). In contrast, calibrations of narrowband echosounders (e.g. EK60) result in estimates of *G* only at f_{nom} (Lunde *et al.*, 2013).

Wideband data from each WBT were processed (see Lavery *et al.*, 2017), to obtain measurements of frequency-dependent volume backscattering strength ($S_V(f)$, dB re m⁻¹), i.e. volume scattering spectra. Using matched filter (MF) processing (Turin, 1960; Chu and Stanton, 1998; Stanton and Chu, 2008), the data had improved temporal (and hence range) resolution, proportional to the signal bandwidth *B* (Hz), and increased SNR, by $\sim 2BT$, where *T* is the signal duration (s) (Stanton and Chu, 2008, and references therein). Using MF processing, $S_V(f)$ at range *r* is given by (Lavery *et al.*, 2010; Stanton *et al.*, 2010):

$$S_{\rm V}(f) = 10 \log_{10} \frac{\langle |\gamma(f)|^2 \rangle}{P_{\rm T} K_{\rm T}} - 10 \log_{10} (L_{\rm TL}(f))^2 - 10 \log_{10} V(f) - 2G(f)$$
(1)

where $P_{\rm T}$ is the transmitted power (Watts), and $K_{\rm T} = 2Z \frac{\lambda_{\rm nom}^2}{16\pi^2}$ is a Simrad-supplied factor intended to account for the received power in a matched load based on generic transducers, where $Z = 75\Omega$ is the equivalent impedance for this system, and $\lambda_{\rm nom}$ is the wavelength at the nominal frequency (e.g. at the centre of the spectral band). $L_{\rm TL}(f)$ is the frequency-dependent transmission loss on a linear scale attributable to spherical spreading and seawater absorption, and is given by:

$$L_{\rm TL}(f) = \frac{10^{-\alpha(f)r/10}}{r^2}$$
(2)

where $\alpha(f)$ (dB m⁻¹) is the frequency-dependent attenuation factor (Francois and Garrison, 1982a, b). $|\gamma(f)|$ is the absolute value of the fast Fourier transform (FFT) of the compressed pulse

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ES18-11 ES120-7C ES200-7C ES333-7C ES38D

Figure 2. Transducers and aluminum deadweight tow body. Simrad split-beam transducer models are shown for each transducer.

output of the range of interest, and <...> denotes the average over multiple transmissions. V(f) is the frequency-dependent insonified volume at range r, and is given by

$$V(f) = \frac{1}{2} c_{\rm w} \tau_{\rm eff} r^2 \Psi_{\rm D}(f) \tag{3}$$

where $\tau_{\rm eff}$ is the duration of the time window after the MF processing, $c_{\rm w}$ is the speed of sound at range r, and $\Psi_{\rm D}(f)$ is the equivalent beam angle (Urick, 1983). Here we use the theoretical, frequency-dependent beam angle for calculations of frequency-dependent volume backscatter.

To display the calibrated MF data, which are 3D (distance/ time, range/depth, and frequency), as a 2D "echogram" used for narrowband data, it is necessary to account for: (i) contributions from all frequency components of the transmitted signal for that channel, (ii) changes in sampling volume and transmission loss with range and frequency, and (iii) the frequency-dependent calibration. These are achieved by calculating the absolute value of the MF output as a function of range, and correcting for all other factors at f_{nom} . This scaled compressed-pulse output (Lavery *et al.*, 2010) at the nominal frequency, $\tilde{S}_v(f_{nom})$ (dB re m⁻¹), does not fully account for the frequency dependence of the various factors, but is adequate for illustration purposes because it approximates $S_v(f)$ at the nominal centre frequency.

$$\tilde{S}_{v}(f_{\text{nom}}) = 10 \log_{10} \frac{\langle |y(t)|^{2} \rangle}{P_{\text{T}} K_{\text{T}}} - 10 \log_{10} |L_{\text{TL}}(f_{\text{nom}})|^{2} - 10 \log_{10} V(f_{\text{nom}}) - 2G(f_{\text{nom}})$$
(4)

where |y(t)| is the absolute value of the compressed-pulse output of the received signal.

Biological data

To verify the species and size composition of the echogram features, bongo nets and the 4SBT were towed within 125 m of the sea surface and near the bottom, respectively, at multiple sites (Figure 1). Zooplankton were collected with a 61-cm diameter bongo net equipped with 335-µm-mesh nets towed obliquely at ca. 3–4 kts, the slowest vessel speed achievable. The bongo net wire was deployed at ca. 50 m min⁻¹ and recovered at ca. 20 m min⁻¹. A time-depth recorder was deployed on the initial tows, but was lost partway through the cruise. The 4SBT was towed in accordance with NEFSC bottom-trawl survey protocols (Politis *et al.*, 2014), but with tow durations adjusted to optimize catches.

4SBT catches were sorted to species, each species weighed *en* masse, and lengths were measured for up to 150 randomly sampled individuals. Bongo samples were preserved in 4% buffered formalin and later processed in the laboratory. Samples were examined qualitatively for overall species composition. In tows dominated by euphausiids, all euphausiids were removed, identified to species, and measured for length (Standard Length 3, from the posterior base of the eye stalk to the end of the sixth abdominal segment; Mauchline, 1980).

Acoustic modelling

Target strength (TS, dB re 1 m²) spectra for euphausiids (i.e. an elongated and fluid-like scatterer) was generated over the frequency band of the acoustic measurements using a Distorted-Wave Born Approximation (DWBA)-based bent cylinder model as developed for Antarctic krill (Euphausia superba) in Lawson et al. (2006) and re-parameterized for M. norvegica as in Wiebe et al. (2013). The model was parameterized using a length-towidth ratio of 18.2, a normal distribution of angles of orientation with a mean of 0° and SD of 27° , and density (g = 1.039) and sound speed (h = 1.038) contrasts from Calise (2009). The length of the equivalent cylinder was taken as the distance from the anterior tip of the eye to the end of the sixth abdominal segment (Lawson et al., 2006) and a correction factor of 1.095 was used to scale the lengths measured in bongo net samples. All euphausiid lengths presented here refer to this "acoustic length" (Lawson et al., 2008). Individual acoustic backscattering cross-sections (σ_{bs}, m^2) were calculated from 1 to 275 kHz at 1 kHz intervals for each acoustic length of euphausiid sampled in the bongo net and then averaged over all individuals, thereby accounting for the sampled length distribution.

TS spectra for butterfish were generated with a modified version of the Kirchhoff ray-mode (KRM) model (Clay and Horne, 1994). Lateral and dorsal views of one live (142-mm caudal length (CL) anaesthetized with MS-222) and two previously frozen (165- and 145-mm CL) butterfish were radiographed (Figure 3), and silhouettes of the body and vertebral column were digitized at 1-mm resolution (e.g. Jech and Horne, 1998). The radiographs were taken at 40-cm focal length and 70 kV, with exposures of 0.126 and 1.103 s and 2.7 and 3.3 mA for the lateral and dorsal radiographs, respectively. The digital silhouettes were converted

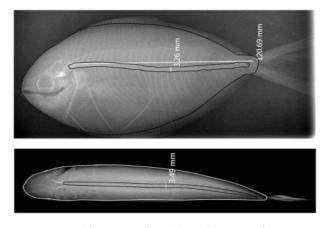


Figure 3. Lateral (upper panel) and dorsal (lower panel) radiographs of a previously frozen 165-mm CL butterfish. The displayed length measurements were applied manually by the radiographer using the radiograph viewing software. Silhouettes (i.e. outlines of the fish body and vertebral column) were those used in the KRM model for this individual.

to equivalent cylinders (Horne and Jech, 1999) and reduced scattering length (RSL) was calculated from 10 to 400 kHz at 1-kHz intervals.

The KRM algorithm of Clay and Horne (1994) was modified so that backscatter (90° dorsal incidence) by the vertebral column was computed as a fluid-filled object embedded in the fish body (rather than the typical implementation where the swimbladder is represented as a gas-filled soft target) and the sound speed and density ratios (h and g) for the vertebral column were calculated as the ratio of the sound speed and density of the vertebral column relative to those of the fish body, not to the surrounding water. Seawater sound speed ($c_{water} = 1477.4 \text{ m s}^{-1}$) and density $(\rho_{\text{water}} = 1026.8 \text{ kg m}^{-3})$ from Jech *et al.* (2016) were used. Material properties (sound speed and density) of butterfish flesh $(c_{\text{flesh}} \text{ and } \rho_{\text{flesh}})$ and vertebrae $(c_{\text{vertebrae}} \text{ and } \rho_{\text{vertebrae}})$ were not directly available. Chosen material property values were constrained such that the reflectivity coefficient (Clay and Medwin, 1977) of fish flesh ranged between 1.2 and 6% (Cushing, 1973), and was $\sim 24\%$ for fish bone (Wright and Colling, 1995); the density ratio, (i.e. $g = \frac{\rho_{flesh}}{\rho_{water}}$) for fish flesh ranged between 1.002 and 1.025; and the sound speed ratio (i.e. $h = \frac{c_{flesh}}{c_{water}}$) was about 1.025 for Atlantic mackerel (Gorska et al., 2005). Given these limits, sound speeds of 1514.2 and 1700.0 m s⁻¹ and densities of 1052.5 and 1400.0 kg m⁻³ were used for the fish flesh and vertebrae, respectively.

To compare theoretical TS(f) to $S_v(f)$ of *in situ* scatterers, the length distribution was used to weight the TS(f). The RSL at each 1-kHz frequency interval and 1-mm length interval was scaled by the proportion of that length class in the length distribution, and then the scaled RSLs were summed over the length intervals to calculate a length weighted backscattering cross-sectional area $(\overline{\sigma}_{bs})$ at each 1-kHz frequency interval:

$$\overline{\sigma_{\rm bs}}(f) = \sum_{i=1}^{N_\ell} \left[\text{RSL}(f, \ell_i) * \ell_i * p(\ell_i) \right]$$
(5)

where *N* is the number of length classes (ℓ) , *p* is the proportion of the *i*th length class, and the proportions sum to 1.

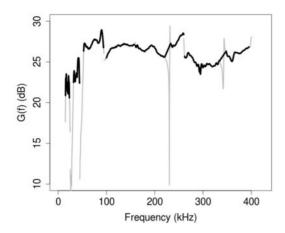


Figure 4. Calibration curves [G(f)] for each broadband channel. The grey lines represent the G(f) for the full transmit/receive band and the thick black lines represent the selected portion of the G(f) curve (Table 1) used to post-process each channel. The spikes in the 200 (between 221.0 and 233.0 kHz) and 333 (between 334.0 and 347.7 kHz) kHz G(f) curves were removed by linearly interpolating over each section of its G(f) curve.

Results

Calibration

Calibrated G(f) was generated for each WBT/transducer pair (Figure 4) using the calibrations conducted in the large tank. The dockside calibrations confirmed the precision of these G(f) measurements. The selected bandwidth used in post-processing for each WBT/transducer pair was usually less than the transmit/receive band (Table 1; Figure 4). The useful bandwidths were selected based on visual inspection of G(f) and *ad hoc* decisions to maintain a ca. <3 dB dynamic range between the minimum and maximum *G* with each G(f). Outside of these bands, the resulting $S_v(f)$ increased implausibly, suggestive of poorly constrained G(f). In addition, the peaks and nulls in G(f) were replaced by linear interpolations (Figure 4). These artifacts, resulting from constructive and destructive interference patterns inherent in the sphere TS(f), would otherwise degrade data analysis and interpretation (Lavery *et al.*, 2017).

Volume backscattering measurements and comparison to theoretical models

During the night of 10 January 2015 at the head of Atlantis Canyon (Figure 1), a scattering feature was observed within 60 m of the sea surface (Figure 5). The $\tilde{S}_v(f_{nom})$ was lowest in the 18kHz band and strongest in the 120- and 200-kHz bands, suggesting a fluid-like scatterer. Bongo-net samples from the vicinity of this feature were dominated by euphausiids. Smaller fluid-like scatterers (copepods) that would contribute negligibly to overall scattering were also present at lower abundances while smaller, strongly scattering elastic-shelled taxa (i.e. the cosomatous pteropods) that can at times dominate scattering even at low abundances (Stanton et al., 1994; Lavery et al., 2007) were absent. The scattering layer thus appeared attributable to euphausiids, specifically M. norvegica, which comprised 99% of the sampled euphausiids by weight. Their lengths ranged from 14 to 33 mm, with a mean and standard deviation of 22.4 and 3.2 mm, respectively (Figure 6). To characterize $S_v(f)$ for this scattering feature,

regions within and outside of the euphausiid patch were visually selected from the echograms (Figure 5), and $S_v(f)$ for each band were combined into a single wideband spectrum (Figure 7). The spectra were combined by applying an FFT over 1-m depth bins for individual pings, and then frequency-dependent spectra were averaged in linear space over range and ping number. Single pings with 1-m depth bins produced highly variable spectra (Figure 7a and b), while averaging spectra over a larger region (i.e. multiple pings and depth bins) produced a smoother spectrum (Figure 7b), which highlighted the overall scattering response for the euphausiids in that feature. The SNR, based on comparison of spectra from within the feature to that from outside of the feature, was about 10 dB in the 15-25 kHz band and increased to nearly 30 dB at the higher-frequency bands for this patch (Figure 7c). In lower-amplitude backscattering regions, spikes were evident in the spectra near 131 and 219 kHz and to a lesser extent near 199-212 and 234 kHz (Figure 7c). These spikes were also evident when data were collected in passive mode (not shown) and hence appear to be due to noise. Spectra averaged for different depths and pings within the main patch and one region of lower backscatter above the main patch showed overall very similar shapes, with only slight variations in the location of the transition from Rayleigh (acoustic wavelength much larger than the scatterer size) to geometric (acoustic wavelength much smaller than the scatterer size) regions, consistent with generally constant length distributions. Magnitudes of the spectra differed between regions, however, consistent with variations in abundance (Figure 7c).

The M. norvegica length distribution was used to generate a predicted TS(f) over the bandwidth of the measured $S_v(f)$ data (Figure 7d). Backscattering levels observed in the euphausiid patch were high, and the predicted TS(f) was scaled by a density of 700 ind m⁻³ to generate a predicted $S_v(f)$ curve for comparison to measured data. In situ euphausiid densities estimated from the bongo samples were ca. 0.5 ind m⁻³, much lower than those assumed in calculating the predicted $S_v(f)$ response curve and likely suggesting that the net system did not pass directly through the patch, avoidance by euphausiids of the bongo net is quite high (as reported in Wiebe et al., 2013), or both. Unfortunately, time-depth recorder information was not available for this tow and hence the exact depth range sampled and proximity to the acoustic patch are uncertain. It is important to note though that while the magnitude of the predicted spectrum is a function of the assumed density (which was adjusted to allow the predicted and measured curves to overlay one another and hence is not independent of the acoustic data), its shape is determined by independent information on the size of euphausiids from the net samples combined with the fluid-like scattering model. The predicted and measured $S_v(f)$ differ by 2–5 dB (predicted less than measured) at the lower bands and <3 dB at the higher bands (Figure 7d).

The TS(f) predicted by the KRM model for the live 142-mm butterfish shows numerous peaks and sharp nulls for the fish body, with relatively flat amplitudes from 10 to 400 kHz. In contrast, TS(f) for the vertebral column has fewer peaks and nulls, the amplitude is comparable to that of the body from 10 to about 100 kHz and above about 375 kHz, and exceeds that of the body between about 100 and 375 kHz (Figure 8a). These patterns are indicative of weak scatterers, where the larger (i.e. thicker) body has nulls that are closer together in frequency space than the nulls from the vertebrae, which is a smaller anatomical feature

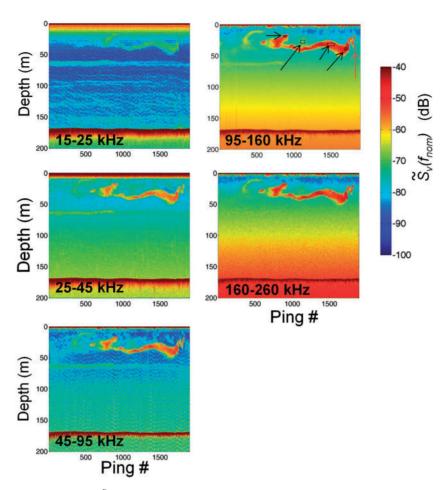


Figure 5. Scaled compressed-pulse output $[\tilde{S}_v(f_{nom})]$ echograms of broadband data collected 2034–2108 (Eastern Standard Time, EST) on 10 January. The large scattering feature between 20 and 50 m depth is a euphausiid patch. Pings 1-1688 were collected with all five channels transmitting simultaneously, after which data were collected in sequential mode. Black boxes in the 95–160 kHz channel echogram represent regions where spectra were calculated within and near the euphausiid patch and the far-right red box represents "open" water where the spectrum was calculated as an approximation for background noise levels.

(Figure 3). The peaks in the vertebra and whole fish scattering curves have a larger-scale trend which increases in amplitude from 10 to about 200 kHz, and then decreases in amplitude to 400 kHz. This decrease at our "higher" frequencies is part of an overall pattern (not shown) where TS(f) eventually increases with increasing frequency.

Around sunset on 10 January 2015, in water depths of ca. 160 m along the continental-shelf edge, between Alvin and Atlantis Canyons (Figure 1), a scattering feature was observed, initially on the seabed and then rising about 20-30 m (Figure 9). The 18 and 38 kHz $\tilde{S}_v(f_{nom})$ was lower than that for the higherfrequencies, again suggesting a fluid-like scatterer. Spectra from single pings and individual 1-m depth bins again showed strong variability (Figure 10a and b). Spatially-averaged spectra for multiple region and depths within the scattering feature were flat from 15 to about 125 kHz, then increased from ~ 125 to 160 kHz, and leveled again about 160 kHz (Figure 10b and c). The spectra from "open" water at different depths surrounding the scattering feature were similar to those from within the layer, but were 10 dB lower in the lower-frequency bands and 2-5 dB lower in the higher-frequency bands (Figure 10c). This result suggested that the SNR for $\tilde{S}_v(f_{nom})$ from the feature was

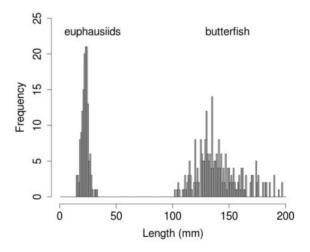


Figure 6. Length-frequency distributions of Northern krill (shorter lengths labelled "euphausiids") captured in a bongo net tow conducted in the vicinity of the shallow scattering feature (Figure 4) and of butterfish FLs (longer lengths labelled "butterfish") subsampled from a 4SBT tow conducted in the vicinity of the deep scattering feature (Figure 8).

acceptable. Again, there were noise spikes near 131, 219, and 234 kHz, and smaller spikes in the 180–212 kHz band (Figure 10b and c).

A bottom trawl haul in the location of this scattering feature, albeit 5 hours earlier, caught 4096.1 kg, dominated by butterfish (97.6% by weight), with 0.5% squid (Loligo and Illex sp.), 0.9% spiny dogfish (Squalus acanthius), 0.4% four-spot flounder (Hippoglossina oblonga), and 0.5% other by weight. The sampling volume for this tow was ca. 41 000 m³ (10-min tow duration at 3 kts with a mouth opening of \sim 45.5 m² when fishing). The numeric density of butterfish was estimated at 1.9 ind m^{-3} for this tow by scaling the catch weight to numbers of fish using the length-weight regression for winter- or spring-caught butterfish $(\ln(W) = -11.6824 + 3.293*\ln(FL);$ Wigley et al., 2003). Butterfish fork lengths (FLs) were on average 14.0 cm with standard deviation 2.1 cm (Figure 6). The butterfish length distribution (Figure 6) was used to weight the KRM TS(f) (Figure 8a) using Equation (5). The length-weighted $\overline{\text{TS}}(f) = 10\log_{10}(\overline{\sigma_{\text{bs}}}(f))$ was scaled by -8 dB (ca. 0.16 ind m⁻³) to match the $S_v(f)$ amplitude (Figure 10d).

Discussion

We present *in-situ* acoustic backscattering measurements on euphausiids and butterfish over more than four octaves of bandwidth. The species composition of the euphausiid aggregation was verified using nets and the $S_v(f)$ was compared with theory. Both lines of evidence support our interpretation that the scattering aggregation was composed of euphausiids. Trawl catch data confirmed that the demersal scatters were butterfish, for which we present a theoretical scattering model. Butterfish have a complicated swimbladder ontogeny where the swimbladder is air filled and then lipid as the fish grows from juvenile to adult states. Therefore, the TS(f) for butterfish will evolve from that for a fish with a gas-filled-swimbladder (e.g. like a clupeid; Reeder et al., 2004; Stanton et al., 2010) to that for a fish without a swimbladder or a lipid-filled swimbladder (e.g. similar to Atlantic mackerel; Simmonds and Armstrong, 1990; Korneliussen, 2010). Our measured and our KRM-based TS(f) for the body, vertebral column, and combined (coherent summation of the body and vertebral column) are consistent with measurements of individual, ex-situ Atlantic mackerel (Scomber scombrus) (Nesse et al.,

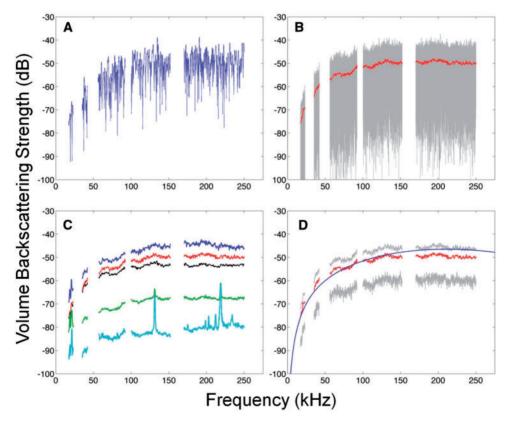


Figure 7. Volume backscattering spectra $[S_v(f)]$ from the areas denoted in and out of the euphausiid patch in Figure 5. Panel (a) shows the spectrum from a single ping (ping number 1689 in Figure 5) averaged over a single 1-m depth bin within the krill patch. Panel (b) shows the spectra (grey lines) from all 1-m depth bins within a 31 ping \times 4 m depth region of the krill patch (pings 1689–1719 and 37–41 m in Figure 5) and the average spectrum for this region (red line). Panel (c) shows spectra from different regions of varying sizes and depths within the main patch (red line as in panel (b), blue and black lines correspond to pings 860–900 and depths 18–20 m and pings 1450–1550 and 30–34 m, respectively), a spectrum from a lower-scattering region just above the main patch (green line, pings 1090–1150 and 25–30 m), and a spectrum from outside the patch but at similar depths in "open" water (cyan line, pings 1818–1833 and 35–41 m). This latter spectrum was used to approximate noise levels. Note that at the time the data shown by the blue, black, and green lines were collected (i.e. for pings 1–1688), the system was transmitting in simultaneous mode, while for pings 1689 onwards it was in sequential mode. Panel (d) shows the average spectrum (red line, panel b) as well as the 10th and 90th percentiles (grey lines) of the distribution of spectra from the region and the theoretical volume backscattering response of fluid-like scatterers modelled over the net-sampled euphausiid acousticlength distribution and assuming a numerical density of 700 individuals m⁻³ (blue line).

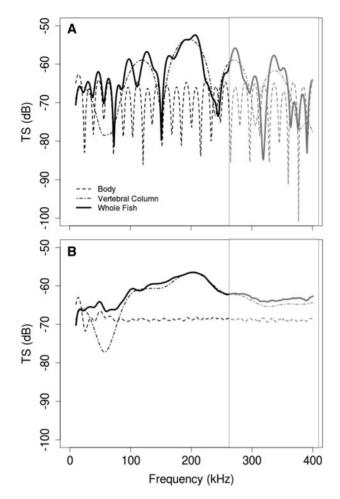


Figure 8. KRM TS response curves for a single, 142-mm CL butterfish at dorsal incidence. The upper panel (a) shows the body only (dashed line), vertebral column only (dash-dot line) and the combined body and vertebrae (solid line). The lower panel (b) shows the TS response curves for the same anatomical parts weighted by the length-frequency distribution of butterfish sampled by the 4SBT haul. The shaded area from 260 to 400 kHz highlights frequencies greater than those used for measurements presented in this study, but are shown to provide context for the KRM predictions.

2009; Forland *et al.*, 2014). They are also consistent with the conceptual and theoretical models by Gorska *et al.* (2007) and Korneliussen (2010), wherein flesh and bone dominate scattering in different portions of our sampled band and the observed spectrum is a complicated superposition of the TS(f) for these two kinds of fluid-like scattering structures. The consistency between mackerel and butterfish backscatter suggests common scattering mechanisms for fish without a gas-filled swimbladder, and provides confidence that these can be extended to other vertebrate (and possibly invertebrate) species with similar anatomical traits (*sensu* Gorska *et al.*, 2005).

The KRM model predicts TS when the target conforms to the Kirchhoff approximations (i.e. g and h values not close to 1, and broadside incidence), but prediction accuracy can deteriorate when those approximations are not met (Jech *et al.*, 2016). We selected the KRM model because it can accommodate digital representations of the fish and anatomical features within the fish,

rather than simplifying the overall morphology to a prolate spheroid or bent cylinder (e.g. Gorska et al., 2005), and without knowledge of the orientation distribution of in-situ butterfish, predictions were constrained to dorsal incidence. The KRM predictions of butterfish scattering were consistent with our mean empirical spectrum (after appropriate scaling), as well as the earlier models of backscatter by Atlantic mackerel (Gorska et al., 2007; Korneliussen, 2010). However, given the uncertainty in material properties of butterfish flesh and vertebrae and that butterfish can be considered weak scatterers (i.e. g and h values close to 1), more work is required to improve confidence in model predictions. For example, empirical studies of the material properties can be used in sensitivity analyses of model predictions by analytical models specifically developed for weakly scattering targets, such as the DWBA (Chu et al., 1993; Stanton et al., 1993) or its variants (phase-tracking DWBA, Jones et al., 2009; Stochastic DWBA, Demer and Conti, 2003), or numerical models such as the boundary element method (e.g. Francis and Foote, 2003) or finite element method (e.g. Macaulay et al., 2013). Selection of appropriate analytical or numerical models depends on a number of factors, such as computing power, availability of 3D digital representations of anatomical features, and material properties (Jech et al., 2016), but scattering models are necessary for understanding the empirical spectral patterns as well as improving predictive capability for a wide variety of environmental and biological conditions that may be encountered throughout resource surveys. With better characterization of the scattering properties of butterfish, algorithms (e.g. De Robertis et al., 2010; Trenkel and Berger, 2013) can be developed to classify butterfish in broadband and narrowband data.

Similar to Simrad's General Purpose Transceivers (GPTs), the WBTs can be transmitted simultaneously or sequentially. Multiple narrowband (i.e. multifrequency) systems are typically transmitted simultaneously to achieve consistent sampling volumes among systems where cross-talk is usually not an issue (i.e. frequencies are selected so that they are not within the bandwidth of each other and so they are also away from harmonics). Multiple broadband systems can have overlap among bandwidths, especially at the band edges, and can overlap harmonics, which may contribute to system noise. More studies are needed to ascertain the factors that affect bandwidth and whether the useful bandwidths are inherent to each specific transducer, WBT, our experiment, or a combination of acoustical and environmental conditions. During a pass over a euphausiid aggregation, we transmitted simultaneously for the first portion of that pass and sequentially for the remainder of that pass (Figure 5). The spectra from the different portions of the euphausiid aggregation did not show evidence of cross-talk among channels, suggesting configurations and processing methods are available to minimize interference among channels. Noise spikes were evident at consistent frequencies throughout the measurements. These spikes were not at the nominal centre frequencies of the bands, and highlight the need to be mindful of noise that may interfere across a wider band. The SNR for the features were marginally acceptable, especially for the butterfish aggregation. At 160 m range, the SNR at the higher frequencies (e.g. > 200 kHz) was less than 10 dB. Our transmit source levels were set to avoid non-linear effects (Tichy et al., 2003), especially at the higher frequencies, primarily due to uncertainty in the response of each transducer to increased power levels. More study is needed to find optimal power levels for the transducers and applications. The measured spectra were

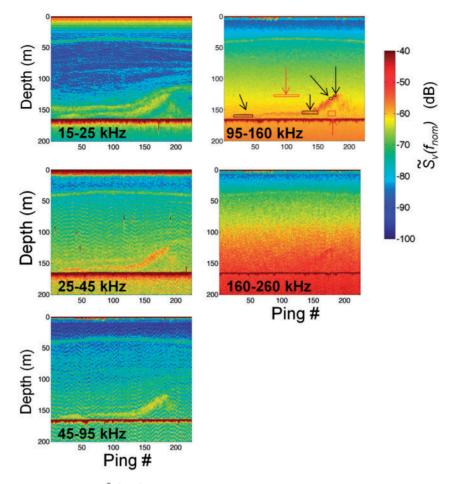


Figure 9. Scaled compressed-pulse output $[\tilde{S}_v(f_{nom})]$ echograms of broadband data collected from 1700 to 1714 (EST) on January 10. All data were collected pinging sequentially. The scattering feature near the bottom (to the left) and rising to about 140 m depth (to the right) is butterfish. Black boxes represent regions where spectra were calculated within this butterfish layer and red boxes indicate regions at two different depths representing "open" water for which spectra were calculated and used as approximations of noise levels.

contiguous across the bands of the five separate channels, confirming the overall validity of their respective calibrations, although some of the variability within band may relate to calibration uncertainty (e.g. the 45–95 kHz band for the euphausiid patch showed a consistent s-shape not expected from the theoretical scattering model).

We observed consistency in the euphausiid and butterfish spectral shapes at various locations within each aggregation, which suggests homogeneity of sizes, behaviour, and likely species within those patches/aggregations. An overarching paradigm of fish and zooplankton groups is that individuals tend to associate with others of the same species and similar sizes (Pitcher, 2001) and although heterospecific fish and zooplankton aggregations and layers are not uncommon (e.g. Lavery et al., 2007; Stanton et al., 2012), fine-scale segregation by species within those heterospecific aggregations has been observed (Parrish, 1989). Acoustic methods have been able to measure and characterize acoustic scattering patterns within aggregations (e.g. Stanton et al., 2010) at the sub-metre scale (e.g. Benoit-Bird, 2009a, b; Lavery et al., 2010). Now with widely available wideband technology, that resolution improves to centimetre scale which facilitates observations of the internal structure of fish and zooplankton aggregations at ever finer vertical resolution.

Although comparisons of the shapes of the scattering response curves based on acoustic scattering theory and organism anatomy to the corresponding backscattering spectra were reasonable, the magnitude of the $S_v(f)$, when scaled by numeric density measures from net catches [i.e. conversion from TS(f) to $S_v(f)$], did not match the mean spectra. There are numerous potential reasons for this mismatch, including within-aggregation patchiness, unknown and/or species-specific net catchability, and differences in sampling volumes (spatially and temporally) between nets and acoustics. In the present application, for instance, the magnitude of scattering spectra was observed to vary within the euphausiid aggregation, indicative of variations in animal packing density. Within one portion of the patch, from comparisons of theoretical to measured spectra, the average density of euphausiids can be estimated as \sim 700 ind m⁻³. These are relatively high densities and much higher than measured by net samples collected in the vicinity of the patch (likely due to a combination of the nets missing the main patch and avoidance), but previous density estimates for M. norvegica as well as the Antarctic krill (E. superba; another large species of euphausiid known to avoid small and slow-moving nets) made from visual, photograph, and acoustic observations are usually much higher than those made with nets (see review in Watkins, 2000; Wiebe et al., 2013).

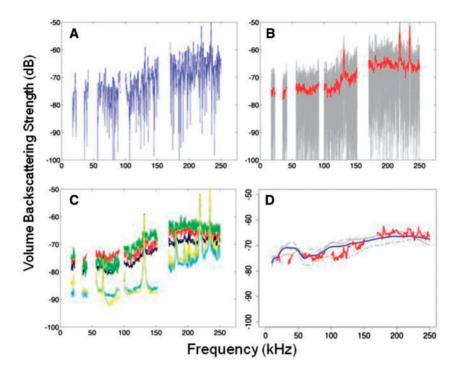


Figure 10. Volume backscattering spectra $[S_v(f)]$ from the highlighted regions in Figure 8 and KRM TS response curves. Panel (a) shows the spectrum from a single ping (ping number 168 in Figure 8) and averaged over a single 1-m depth bin within the butterfish layer. Panel (b) shows the spectra (grey lines) from all 1-m depth bins within a 6 ping \times 4 m in depth region of the butterfish layer (pings 168–173 and 130–134 m in Figure 8) and the mean spectrum for this region (red line). Panel (c) shows spectra from different regions of varying sizes and depths within the butterfish layer (red line as in panel (b), blue, black, and green lines corresponding to pings 16–46 and depths 158–162 m, pings 125–150 and 152–157 m, and pings 175–179 and 126–129 m, respectively); and two spectra from outside the layer in "open" water but at similar depths to where butterfish spectra were calculated (cyan and yellow lines, pings 80–120 and 126–129 m in depth and pings166–178 and 152–160 m, respectively). These latter spectra were used to approximate noise levels. Panel (d) shows the mean butterfish spectrum with noise spikes removed for clarity of presentation and averaged into 1-kHz frequency bins (red line, as in panel b), individual KRM TS response curves weighted by the butterfish length-frequency distribution for the three radiographed butterfish (live 142-mm denoted by the solid grey line, frozen 165-mm denoted by the dashed grey line, and frozen 145-mm denoted by the dash-dot grey line), and the mean length-frequency-weighted KRM TS response curve (blue line). The KRM TS response curves were scaled by -8 dB (ca. 0.16 ind m⁻³) to match the magnitude of the mean Sv(f) spectra.

In the case of butterfish, the trawl catch suggested greater densities than that measured by the EK80. The length-weighted TS(f) needed to be scaled by $-8 dB (0.16 ind m^{-3})$ to match the measured backscatter amplitude. A bottom trawl was conducted in the same location as the observed butterfish scattering feature (Figure 9), but 5h earlier during daylight hours. Using the swept volume (Jech and Sullivan, 2014) from that tow, a butterfish density of 1.9 ind m^{-3} was estimated, giving a scaling factor of 2.8 dB. The order of magnitude difference between these two scaling factors is most likely due to the nocturnal vertical migration behaviour of butterfish, where butterfish are typically on the seabed during daylight hours and rise above the seabed (e.g. Figure 9) at dusk. Examination of the narrowband 200-kHz volume backscatter data collected by Karen Elizabeth immediately prior to the trawl showed a 0.5-1 m thick layer of butterfish on the seabed with a maximum of $-36 \,\mathrm{dB}$. During the ascent, the layer expanded vertically (Figure 9) and narrowband data immediately following wideband data collection on the scattering feature (Figure 9) showed multiple aggregations off the seabed with maximum values ranging from -41 to -47 dB. The 5-10 dB difference in volume backscatter between butterfish on vs. above the seabed is likely due to changes in packing density. Orientation of the butterfish is another factor, but the observed difference is commensurate with the difference in scaling factors needed to match model-based TS(f) to the volume backscattering measurements, so a change in packing density is the more likely cause.

Over the decades since COTS multifrequency systems became available, multifrequency narrowband data have been collected during marine resource surveys and studies of marine ecology, and methods have been developed to classify these data to the lowest taxonomic level possible. In most cases, species identification is not possible, but classification to family or genus or scattering type (e.g. gas-bearing, fluid-like) has been useful for fisheries management, research, and addressing ecological questions (Trenkel et al., 2011 and references therein). Frequency dependent backscatter by fluid-like scatterers has been exploited to separate fluid-like scatterers from gas-bearing scatterers (e.g. Madureira et al., 1993), discriminate multiple scattering types (e.g. Korneliussen and Ona, 2003; De Robertis et al., 2010), or generate acoustic indicators (e.g. Trenkel and Berger, 2013). The progression from single to multiple narrowband to wideband data certainly provides more signal and data, but whether more

data translates to more useful information and better classification remains an open question. For example, simply adding more narrowband frequency information does not necessarily improve length-based abundance estimates (Horne and Jech, 1999), and broadband data solely within the geometric scattering region of a fish does not necessarily improve species identification (e.g. Simmonds *et al.*, 1996). This is because to maximize the information content, the ratio of organism size to acoustic frequency must span the transition region (where the acoustic wavelength is similar to the size of the object) from Rayleigh to geometric scattering (Holliday and Pieper, 1995), or at a minimum include scattering features such as resonance peaks (Stanton *et al.*, 2010), that are useful for inversions (Horne and Jech, 1999).

Judicious choice of narrowband echosounders can lead to successful classification, so why make the switch to wideband? Wideband data can be used to fill in the gaps between multiple narrow bands, and conversely, volume backscattering levels at the bands that match historical narrowband data can be selected and used to estimate parameters for classification algorithms. For the euphausiids encountered during our measurements, selecting narrowband data at centre frequencies commonly used for fisheries surveys (e.g. 18, 38, 120, 200 kHz) map the Rayleigh to geometric scattering regions quite well, so one can contend that wideband data are not advantageous and are more costly in terms of data storage, electronic requirements, and processing than multiple narrowband systems. For example, when the species of interest is in monospecific aggregations, estimates of scatterer density and abundance are quite satisfactory and robust using narrowband data, whereas broadband information may be unnecessarily complex for these estimates and it actually may be more effective to reduce broadband to narrowband when estimating abundance for fisheries assessments or other needs. However, when species are mixed, broadband information may be useful for classification and selecting data to be used in abundance estimates (Stanton et al., 2012). For our data, by filling in the gaps between narrow bands, we know that the acoustic response is monotonically increasing from 18 to 38 kHz (i.e. neither is a peak or valley in the spectrum which could indicate a resonance by small bubbles or gas-filled swimbladders) and the transition from Rayleigh to geometric scattering can be more accurately located, which provides a better estimate of euphausiid size than with multiple narrow bands. The wideband spectrum arguably contains more information for the butterfish than multiple narrow bands. The butterfish spectrum would be difficult to decipher and especially the inflection between 120 and 160 kHz would be difficult to detect without the wideband data. The frequency of the transition may be dependent on the material properties and length of the fish (e.g. Gorska et al., 2007; Korneliussen, 2010) and may be useful for sizing the butterfish and discriminating butterfish from other fish without gas-filled swimbladders.

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