

1 Accounting for biological and physical
2 sources of acoustic backscatter improves
3 estimates of zooplankton biomass

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12 **Abstract:** In order to convert measurements of backscattered acoustic en-
13 ergy to estimates of abundance and taxonomic information about the zoo-
14 plankton community, all of the scattering processes in the water column need
15 to be identified and their scattering contributions quantified. Zooplankton
16 populations in the eastern edge of Wilkinson Basin in the Gulf of Maine in
17 the Northwest Atlantic were surveyed in October 1997. Net tow samples at
18 different depths, temperature and salinity profiles, and multiple frequency
19 acoustic backscatter measurements from the upper 200 meters of the water
20 column were collected. Zooplankton samples were identified, enumerated, and
21 measured. Temperature and salinity profiles were used to estimate the amount
22 of turbulent microstructure in the water column. These data sets were used
23 with theoretical acoustic scattering models to calculate the contributions of
24 both biological and physical scatterers to the overall measured scattering level.
25 The output of these predictions shows that the dominant source of acoustic
26 backscatter varies with depth and acoustic frequency in this region. By quanti-
27 fying the contributions from multiple scattering sources, acoustic backscatter
28 becomes a better measure of net-collected zooplankton biomass.

29 Keywords: acoustic backscatter, zooplankton, Gulf of Maine

30 **Introduction**

31 Acoustic surveys of zooplankton and fish offer many advantages over
32 other sampling techniques (Holliday and Pieper 1995; Foote and Stanton
33 2000). Measurements of acoustic backscatter using a scientific echosounder
34 can be made from ships, buoys, or moorings and thus provide greater spatial
35 coverage or longer time series than conventional sampling techniques such as
36 net tows or diver observations. While video sampling methods (Davis et al.
37 1996; Benfield et al. 2001, 2003) provide some of the same advantages as acous-
38 tics, they are not used as extensively as acoustic surveys and typically have
39 much smaller sampling volumes.

40 One of the difficulties in using acoustic backscatter to measure marine
41 life in the ocean is that the data collected are indirect measures of biota.
42 If aggregations of animals in a region are mono-specific and of similar size,
43 theoretical backscatter models can be used to estimate their distribution and
44 abundance (Hewitt and Demer 2000). If more than one frequency of sound is
45 used, then more categories and size classes of animals may be distinguished and
46 their abundance estimated (Martin et al. 1996; Brierley et al. 1998). However,
47 these approaches require that the type of the scatterers present in the water
48 column is known. Because of this, zooplankton samples are typically collected
49 by net tows during an acoustic survey. These samples provide taxonomic and
50 size information that can be used to predict the level of acoustic backscatter
51 in the water column.

52 This prediction is often referred to as the Forward Problem (FP). For
53 a particular type of backscatterer (usually delineated by either size or tax-
54 onomy), the scattering contribution can be found by multiplying the numer-
55 ical density of scatterers (N_i , in units of m^{-3}) present and the backscatter
56 cross section for an individual scatterer of this type (σ_i , in units of m^2). The
57 backscatter cross section is a function of several parameters including: the size,
58 shape, composition, and orientation of the scatterer and acoustic frequency.
59 The output of the FP is the volume backscatter coefficient (s_V , with units of
60 m^{-1}) which is found by summing the contributions from the different types of
61 scatterers present

$$62 \quad s_V = \sum_i (N_i \sigma_i) \quad (1)$$

63 These calculated levels of backscatter are then compared with measured values
64 from field surveys. Often echosounders record the volume backscatter strength
65 (S_V , with units of dB), which is related to s_V by

$$66 \quad S_V = 10 \log_{10} (s_V) \quad (2)$$

67 If the measured and predicted values agree, the Inverse Problem (IP)
68 (using measured scattering values and theoretical scattering models to deter-
69 mine the number, size, or type of animals present) is more likely to be solved
70 correctly. However, occasionally the predicted and measured volume backscat-
71 ter strengths differ by an order of magnitude (or more), which can lead to large

72 errors or uncertainty in solving the IP. For example, a 3 dB difference in S_V
73 corresponds to a factor of two difference in the number of animals or their
74 biomass. In practice, solutions to the FP are used in a diagnostic sense to
75 determine how well the theoretical scattering models predict observed levels
76 of backscatter in the ocean (Wiebe et al. 1996, 1997). These results often in-
77 dicate that only a small subset of the animals present in the water column are
78 acoustically important or, in some cases, that backscatter predictions based
79 on the sampled zooplankton are unable to account for all of the observed
80 backscatter and that another scattering source is unaccounted for in the FP
81 analysis (Mair et al. 2005).

82 While nearly every acoustic survey relies on net tow data to ground-
83 truth the acoustic data, many do not take complete advantage of all the
84 available information provided by the net tow. Typically, FP calculations are
85 performed to identify the acoustically dominant taxa in the water column,
86 however the quantification of backscatter contributions from all of the taxa
87 found in the water column is rarely done. While many taxa present in the
88 water column will contribute negligibly to the overall level of backscatter (due
89 to small size, low numerical density, or low scattering efficiency), there are
90 often several scattering sources that contribute substantially to the overall
91 level of measured acoustic backscatter. Furthermore, the vertical distribution
92 of zooplankton in the water column varies, which may cause the taxa that is
93 the largest acoustic scatterer to change as a function of depth. Many of the

94 net tows used to ground-truth acoustic data can not provide this informa-
95 tion, which may further complicate estimates of zooplankton abundance or
96 distribution.

97 Even if the backscatter from marine organisms is accurately measured,
98 there are other processes in the ocean that can contribute measurable amounts
99 of backscattered acoustic energy. Suspended sediments, air bubbles, ocean
100 mixing processes, and other biota have all been observed during acoustic sur-
101 veys of zooplankton (Wiebe et al. 1997; Trevorrow 1998), however backscatter
102 from non-biological processes is rarely quantified during field surveys.

103 Acoustic methods have been used to observe physical mixing processes
104 in the ocean for many years (Thorpe and Brubaker 1983; Orr et al. 2000; Ross
105 and Lueck 2005). It has only been in the last decade that theoretical scattering
106 models for these processes have begun to be tested in the field. These models
107 use the variations in temperature and salinity to calculate changes in the index
108 of refraction and density in the water column that result from turbulence
109 and other mixing processes. The acoustic scattering that occurs from these
110 variations can then be predicted (Goodman 1990; Seim et al. 1995; Lavery
111 et al. 2003).

112 Depending upon the mixing rates present (generally characterized by
113 the dissipation rates of turbulent kinetic energy (ϵ) and temperature variance
114 (χ)), backscatter from turbulent microstructure can be equal to or greater

115 than that from assemblages of zooplankton, particularly at the lower range of
116 frequencies commonly used for acoustic surveys (i.e. < 100 kHz, see Fig. 1 in
117 Warren et al. (2003)). Estimates of S_V from this mechanism range from -110
118 dB in calm waters with small temperature and salinity stratification to -60
119 dB or higher for regions of intense mixing such as the Bosphorus Strait (Seim
120 1999).

121 This study examines the contributions of both biological and physical
122 sources of backscatter to the water column in the Gulf of Maine. Contributions
123 from each scattering source were quantified using theoretical scattering models
124 and either net tow data or hydrographic profiles for multiple depth bins of the
125 water column and multiple acoustic frequencies. The theoretical predictions
126 of backscatter from the different scattering sources were used to correct the
127 amount of measured backscatter in the water column to reflect scattering only
128 from biological sources. The adjusted values of measured backscatter were then
129 compared with measurements of zooplankton biomass.

130 **Materials and methods**

131 As part of a GLOBEC (GLOBal ocean ECosystem dynamics) pro-
132 cess cruise studying the populations of *Calanus* in the basins of the Gulf of
133 Maine, an acoustic survey was conducted in the eastern part of Wilkinson
134 Basin (located between Georges Bank and Stellwagen Bank) in mid-October
135 1997 from the RV *Endeavor* (Table 1). To provide spectral information about

136 the acoustic backscatter processes occurring in the water column, multiple
137 frequency acoustic backscatter data were collected by BIOMAPER-II (Bio-
138 Optical Multi-frequency Acoustical and Physical Environmental Recorder)
139 (Wiebe et al. 2002).

140 Multiple Opening and Closing Net and Environmental Sensing System
141 (MOCNESS) tows (Wiebe et al. 1985) were conducted to collect zooplankton
142 samples while BIOMAPER-II was concurrently recording acoustic backscatter
143 data from the water column. Profiles of the temperature and salinity of the wa-
144 ter column were made with CTD sensors onboard the MOCNESS and nearby
145 higher-resolution CTD casts (Sea-Bird 9/11) from the vessel. Data from two
146 sampling periods are presented herein, with samples from yearday 287 (CTD
147 #08, MOC #07) and yearday 289 (CTD #10, MOC #09). The CTD and
148 MOCNESS stations took place in the same general area (Table 1). These data
149 sets (acoustic backscatter, zooplankton taxa and size, and temperature and
150 salinity profiles) provided enough information to estimate the contributions
151 from biological and physical sources of acoustic backscatter.

152 **Acoustic backscatter measurements**

153 BIOMAPER-II (Wiebe et al. 2002) is a towed body with numerous
154 acoustic, environmental, video, and bio-optical sensors. The acoustic system
155 consists of five pairs of transducers (operating at 43, 120, 200, 420, and 1000
156 kHz), with one of each frequency looking upward and the other downward.

157 The transducers have depth ranges of 200, 200, 150, 100, and 35 meters re-
158 spectively with a vertical resolution of 1 m depth bins. Backscattered energy
159 from each transducer and for each depth bin was recorded as echo-integrated
160 volume backscattering strength every 12 s. At typical survey speeds this ping
161 rate corresponds to a horizontal range between pings of between 20 and 50 m.
162 The instrument was normally towed obliquely through the water column, how-
163 ever since additional equipment was in the water during these measurements,
164 the tow-body was kept at a constant depth of approximately 5 m below the
165 surface and slower tow-speeds resulted in a horizontal resolution between pings
166 of approximately 10 to 15 m. Because of this configuration and the limited
167 depth range of the 1 MHz transducer, only data from the downward-looking
168 transducers at 43, 120, 200, and 420 kHz were analyzed.

169 The acoustic data were processed and combined with data from the ESS
170 (Environmental Sensing System) sensors that are also on board BIOMAPER-
171 II. The acoustic system was calibrated using standard target spheres before
172 the cruise. The final data file provided echo integrated volume backscatter
173 coefficients (s_V) for the water column along with date, time, position (latitude,
174 longitude, instrument depth), temperature, salinity, fluorescence, turbidity,
175 and other sensor data.

176 **Zooplankton net sampling**

177 Two 1 – m² MOCNESS (Wiebe et al. 1985) tows were analyzed to

178 identify, enumerate, and measure the zooplankton present in the waters of
179 Wilkinson basin. A MOCNESS system consists of a series of nine nets, which
180 enables specific depth strata to be sampled. Generally, net #0 was open from
181 the surface to the deepest point of the tow (ten to twenty meters above the
182 bottom), the remaining nets (#1-8) were opened and closed in succession every
183 25 to 50 meters during the return to the surface. The MOCNESS system also
184 recorded the volume of water filtered by each net, the time that each net was
185 opened and closed, depth, salinity, temperature, density, and fluorescence.

186 The nets were equipped with 333 μm mesh and cod end buckets for
187 collection of zooplankton and larval fish. Each cod end sample was split and
188 stored in a buffered formalin solution. Post-processing of the samples consisted
189 of silhouette photography of the animals (Davis and Wiebe 1985). These im-
190 ages were then examined under a microscope and the organisms were measured
191 and identified by taxonomic group. Numerical density and biomass (mg m^{-3})
192 were then calculated for each net for each taxonomic group (Davis and Wiebe
193 1985; Wiebe 1988).

194 **Scattering models**

195 Mathematical models that combine scattering physics and the geome-
196 try of the animal shape for several types of zooplankton were used to provide
197 backscatter information for single animals for use in the FP analysis (Table 2).
198 These models have been developed previously and only slight modifications

199 to some input parameters have been made in this work. These modifications
200 were limited to body length-to-width relationships and the use of a simple
201 fluid-like tissue model for certain gelatinous animals. The models represent
202 the three main taxonomic types of zooplankton: fluid-like, elastic-shelled, and
203 gas-bearing animals (Stanton et al. 1998).

204 *Biological scatterers*

205 Fluid-like models were used for copepods, euphausiids, amphipods, and
206 other animals that have a thin shell (which does not support a shear wave)
207 and a body composition that has similar density and sound speed to that of
208 sea water. Fluid-like animals, which constituted the majority of zooplankton
209 taxa that were encountered in the Gulf of Maine and Georges Bank region,
210 were modeled as weakly-scattering, bent, fluid cylinders (Stanton et al. 1993b;
211 Chu et al. 1993). The model has input parameters of: animal size (typically a ,
212 the radius), the acoustic frequency (f) or wavenumber ($k = \frac{2\pi f}{c}$ where c is the
213 speed of sound in seawater), the ratio of sound speed and density between the
214 scatterer and the surrounding fluid (g and h), the length to width ratio of the
215 animal (β_D), and the orientation of the cylinder relative to the acoustic wave
216 front. An assumed range of orientation angles based upon previous studies
217 was used in modeling the euphausiids, as acoustic backscatter strength is a
218 function of animal orientation (Sameoto 1980; McGehee et al. 1998; Warren
219 et al. 2002). These animals often orient in a slight head-upward posture and
220 were modeled with a $20^\circ \pm 20^\circ$ orientation distribution where 0° is broadside

221 orientation.

222 Modeling of other fluid-like animals was similar to that for the eu-
223 phausiids. The value of β_D was changed slightly to better reflect the body
224 shapes of the other fluid-like animals (Table 2). Although Benfield et al. (2001)
225 indicated that copepods may tend to orient themselves vertically in the water
226 column, it was not known under what conditions this occurs, so an average
227 over all orientations (uniform distribution) was used for all fluid-like animals
228 except for euphausiids. Small changes in the values of g and h can cause large
229 variations in the level of scattering from an animal (Chu et al. 2000; Chu
230 and Wiebe 2005), so to minimize variability in this analysis, constant values
231 of $g = 1.0357$ and $h = 1.0279$ (Foote 1990) were used for all fluid-like ani-
232 mals. It is not known whether the fluid-like animals found in this region have
233 similar material properties as few data are available for animals other than
234 copepods and euphausiids. If differences in the material properties exist for
235 the fluid-like animals, that would cause larger variations in the predicted level
236 of biologically-caused backscatter.

237 Elastic-shelled models were used for animals with a hard, elastic shell
238 such as pteropods. Pelagic pteropods are typically very small (< 1 mm in diam-
239 eter), but scatter a large amount of sound (per unit biomass) due to their dense
240 shell. Other strong scatterers are gas-bearing animals such as siphonophores,
241 where the scattering is caused by small gas bubbles used for buoyancy. Gelati-
242 nous animals (e.g. salps or medusae) or parts of animals (e.g. siphonophore

243 nectophores) have not been modeled as thoroughly as the fluid-like or elastic-
244 shelled animals, however due to their body composition, it is believed that
245 they scatter smaller amounts of sound by a mechanism similar to that of the
246 fluid-like animals (Monger et al. 1998).

247 Several of these zooplankton models have been compared with mea-
248 sured scattering from individual animals (copepods, euphausiids, pteropods,
249 and siphonophore nectophores and pneumatophores) (Stanton and Chu 2000;
250 Stanton et al. 2000; Warren et al. 2001). The remaining zooplankton models
251 have not specifically been tested against measurements from individual an-
252 imals (amphipods, salps, polychaetes, chaetognaths, larval crustaceans, and
253 cyphanautes), however these groups contain animals that are believed to be
254 less important acoustically in this study due to either low numerical densities
255 or very weak scattering characteristics.

256 The zooplankton backscatter models were combined with the abun-
257 dance data from the MOCNESS tows to estimate the level of biologically-
258 caused scattering in the water column. For each animal type collected in a net
259 tow, the backscatter contribution for an individual animal was determined us-
260 ing the appropriate scattering model. These contributions were summed over
261 all animals collected and then divided by the volume of water sampled by
262 the net to arrive at a volume backscatter coefficient for the depth stratum of
263 the net. The contributions from all of the zooplankton were summed and the
264 result was a predicted volume backscatter coefficient for biological sources.

266 Additional sources of backscatter that have been observed in the vicin-
267 ity of Georges Bank include suspended sediments and bubbles. These scat-
268 terers are not believed to be important in this study due to the absence of
269 sediment in net tows and the relatively calm sea state during the survey pe-
270 riod. However, internal waves were seen in the acoustic record during the
271 survey and thus the importance of backscatter from the resultant turbulent
272 microstructure was examined.

273 Scattering from turbulent microstructure in the water column was ana-
274 lyzed in a parallel manner to that from zooplankton except that hydrographic
275 data is used as the scattering model input instead of net tow data. Predic-
276 tions from the theoretical backscatter model were made based on inputs of
277 temperature, salinity, and the dissipation rates of turbulent kinetic energy
278 and temperature variance (Seim 1999). The latter two values were estimated
279 using temperature and salinity profiles from CTD casts taken either before
280 or after the MOCNESS tow (Table 1) (Thorpe and Brubaker 1983; Warren
281 et al. 2003). Although the ESS system on the MOCNESS provided temper-
282 ature and salinity profiles, the sampling rate was limited to 0.25 Hz, thus in
283 order to resolve temperature and salinity variations at vertical scales less than
284 a meter, the higher resolution CTD cast data (sampled at 24 Hz) were used.

285 The CTD data were not collected concurrently with the acoustic and

286 net tow data so there are potential errors in using the CTD data to describe the
287 structure of the water column when the net and acoustic data were recorded.
288 However, hydrographic profiles for each CTD cast were consistent with the
289 profile recorded by the corresponding MOCNESS tow. While this method is
290 far from ideal for measuring values of turbulent kinetic energy and tempera-
291 ture variance, this method has been used previously to make realistic estimates
292 that compare favorably to measurements of turbulent kinetic energy and tem-
293 perature variance made in a similar region (Warren et al. 2003; Seim 1999).
294 Sea state was relatively calm during these tows so we believe that errors due
295 to vertical ship and CTD sensor movement are minimal. The estimated level
296 of backscatter from microstructure was then averaged over the depth ranges
297 sampled by each MOCNESS net so that it could be compared with the FP
298 estimates from the zooplankton.

299 **Results**

300 Data are presented for two MOCNESS tows that occurred in nearby
301 regions but differed in the types of zooplankton present, levels of acoustic
302 backscatter, and water column structure. MOCNESS #7 was lowered to 191
303 meters depth and was brought to the surface with a net closed and new net
304 opened at 175, 150, 125, 101, 74, 50, 26, and 0 m. The lower nets contained
305 large amounts of biomass ($150 - 200 \text{ mg}\cdot\text{m}^{-3}$) and were dominated by copepods
306 and euphausiids (Figure 1a). The surface layer (0 - 26 m) had a higher level of
307 biomass (over $100 \text{ mg}\cdot\text{m}^{-3}$) than the other upper water column samples and

308 was composed of copepods, polychaetes, chaetognaths, and amphipods.

309 [Figure 1 here]

310 Data from MOCNESS # 9 show a different depth distribution of biomass,
311 as well as a slightly more diverse taxonomic composition (Figure 1b). Nets
312 were opened and closed at depths of 180, 153, 124, 99, 80, 60, 39, 20, and 0 m.
313 The zooplankton collected were dominated by an enormous number of salps
314 (2,500 animals m^{-3}) near the surface (from 20 - 40 m depth) resulting in a
315 large amount of biomass, nearly $1 \text{ g}\cdot\text{m}^{-3}$. There was also a substantial amount
316 of biomass from 80 - 124 m that was composed of copepods and euphausiids,
317 as well as a copepod-dominated bottom layer. The salp surface layer was an
318 unusual occurrence on this cruise and no other net sample from the nine MOC-
319 NESS tows collected during the cruise had such a large amount of biomass.
320 For both net tows, the dominant component of biomass at most depths were
321 calanoid copepods.

322 As each MOCNESS tow was being conducted, BIOMAPER-II collected
323 acoustic data while being towed at a depth of approximately 5 meters beside
324 the ship. The acoustic data were offset horizontally from the MOCNESS sam-
325 ples by the amount of wire out on the net tow (at most a few hundred meters).
326 In order to compare the acoustic regions with the MOCNESS information, the
327 trajectory of the MOCNESS was overlaid on the acoustic plot to determine
328 where each net sampled (Figure 2).

329 [Figure 2 here]

330 The acoustic data for MOCNESS #7 shows strong backscatter at the
331 higher frequencies and weaker backscatter at 43 kHz for much of the water
332 column, although this pattern is reversed for the surface and deepest wa-
333 ters sampled (Figure 2a). Remnants of an internal wave were observed in the
334 echogram and the upper layer of the wave, sampled by nets # 7 and 8, had the
335 strongest backscatter at the lowest acoustic frequency, while backscatter from
336 the thick layer between 50 and 100 m was strongest at the highest frequen-
337 cies. The echogram collected for MOCNESS # 9 show a mid-water scattering
338 layer that was sampled by nets #4 - 6 with backscatter that had a simi-
339 lar relationship between scattering strength and acoustic frequency (Figure
340 2b). This frequency dependence is consistent with the backscatter model used
341 for fluid-like scatterers (Warren et al. 2003). A near-surface scattering layer
342 (sampled by net # 7) shows the opposite effect (strongest backscatter at lower
343 frequencies) that indicates the scattering was dominated by a different type
344 of scatterer, possibly the large amount of salps or physical processes occurring
345 at the thermocline.

346 The water column profile for MOCNESS #7 showed a well-mixed re-
347 gion from 20 - 60 m with a steep temperature and salinity gradient above
348 this layer and a shallower gradient below (Figure 3a). These mixing processes
349 likely contributed to the backscatter observed between 15 and 100 m in the
350 echogram (Figure 2a). The hydrographic data for MOCNESS # 9 showed a

351 well-mixed region in the upper 20 m of the water column with a large gra-
352 dient in temperature, salinity, and density that occurred in the next 10 m
353 (Figure 3b). There were several regions of potential or recent mixing (shown
354 by unstable or nearly vertical sections of the density profile) occurring between
355 0 - 20 m, 60 - 100 m and 140 - 180 m, although there were smaller instabilities
356 that occurred throughout the profile.

357 [Figure 3 here]

358 When FP predictions of backscatter were examined for the individual
359 contributions for different animals or processes, the dominant scatterers for
360 MOCNESS # 7 were turbulent microstructure, euphausiids, and siphonophore
361 pneumatophores, however the amphipod category (which included other larval
362 crustaceans) and chaetognaths also caused appreciable amounts of backscat-
363 ter (Figure 4). The other animals (particularly the abundant copepods) con-
364 tributed little to the overall predicted scattering except at the highest frequen-
365 cies. Copepods also contributed little to the predictions for MOCNESS # 9.
366 Turbulence, euphausiids, salps, and siphonophore pneumatophores were the
367 largest contributors to the backscatter (Figure 5). It is striking that the cope-
368 pods which were by far the largest contributors to biomass have such a small
369 contribution to the predicted levels of backscatter. This is primarily a function
370 of copepod size (a few mm in length) and acoustic frequency or wavelength.
371 For the frequencies used in this survey, copepod backscatter is primarily a
372 function of animal size and despite their numerical abundance in the net tow

373 data, they are simply too small to contribute much to the predicted level of
374 backscatter except at the higher frequencies.

375 [Figures 4 and 5 here]

376 The MOCNESS data provided information about the contributions
377 that different zooplankton taxa make to the overall amount of biomass. Simi-
378 larly, the relative contributions of different biological and physical sources to
379 the total amount of predicted backscatter in the water column can be made by
380 combining MOCNESS data, CTD data, and backscatter models. The relative
381 contribution of each scattering source (each animal taxa and microstructure)
382 was calculated for each MOCNESS net depth range and BIOMAPER-II fre-
383 quency (Figures 6 and 7). The percentage contribution to the total predicted
384 backscatter strength was found by dividing the predicted volume backscat-
385 ter coefficient for each scatterer type by the overall calculated backscatter
386 prediction. The percentage of total predicted scattering from physical (non-
387 biological) sources was calculated. The measured level of scattering (from the
388 BIOMAPER-II data) was then reduced by this percentage to arrive at a cor-
389 rected amount of measured backscatter that is believed to be from biological
390 scatterers.

391 [Figures 6 and 7 here]

392 For example, the measured level of backscatter for MOCNESS #9
393 from 0 - 20 m depth at 120 kHz is $s_V = 1.58 \times 10^{-6} \text{ m}^{-1}$ ($S_V = -58.0 \text{ dB}$).

394 From Figure 7, only 8% of the predicted backscatter for this sample is from
395 biological sources. By multiplying the percentage of biologically-caused pre-
396 dicted backscatter and the measured level of backscatter, an estimate of the
397 biologically-caused scatter in the water column was made, $s_V = 1.26 \times 10^{-7}$
398 m^{-1} ($S_V = -69.0$ dB).

399 Non-biological backscatter contributions were important for several re-
400 gions sampled by MOCNESS # 7 (Figure 6). The predicted contributions
401 from microstructure were largest for the region between 20 and 100 m (which
402 again corresponded to regions of mixing indicated in the hydrographic data)
403 while euphausiids were the main scatterers for the deepest nets. Siphonophores
404 contributed to the backscatter more for lower acoustic frequencies and were
405 negligible at the highest frequency. The backscatter in the near surface was
406 the most diverse with regard to scatterer type with nearly all taxonomic types
407 contributing.

408 For MOCNESS # 9, the deeper water column and surface layer were
409 dominated by scattering from microstructure (Figure 7). These large backscat-
410 ter contributions from turbulent microstructure occurred in the same regions
411 that the hydrographic profile data indicated was well-mixed (0 - 20 m and 140
412 - 170 m). Euphausiids were the dominant scatterers in the mid-water depths,
413 with siphonophore pneumatophores and nectophores also contributing. Salps
414 were extremely weak scatterers and while outnumbering the other animals and
415 dominating the biomass in the near-surface, they contributed only 30% - 60%

416 to the total backscatter in that region.

417 If the FP is well-posed and one taxa dominates the measured scattering,
418 then there should be a relationship between biomass and measurements of
419 backscatter strength. Both biomass and volume backscatter cross-section (s_v)
420 are linear functions (for a particular taxa) of the number of animals. Therefore
421 it is likely that a relationship between biomass and volume scattering strength
422 should exist. The relationship between biomass and measured backscatter may
423 not be linear however if more than one scattering process (or taxonomic type
424 or size class) is substantially contributing to the measured backscatter.

425 The biomass and acoustic backscatter data sets from both MOCNESS
426 tows were combined and the regression between the logarithm of biomass
427 and measured acoustic backscatter strength (S_v , a logarithmic measure of
428 acoustic backscatter) was found for each acoustic frequency. The log of both
429 biomass and backscatter was used as some of the acoustic data (specifically
430 predicted backscatter for some scatterer types) ranged over nearly five orders
431 of magnitude. The backscatter model used for salps has not been as well tested,
432 by comparing theoretical backscatter predictions with measured backscatter
433 from individual animals, as the backscatter models used for other animals.
434 Because of this fact and the extremely high biomass of salps caught in net #7
435 of MOCNESS #9, the data from this net were not included in this analysis.

436 There was not a strong relationship (all r^2 values < 0.4) between log-

437 transformed biomass and measured backscatter levels for any of the four fre-
438 quencies used (Figure 8). A poor relationship between zooplankton biomass
439 and acoustic backscatter is likely to occur when non-zooplankton scatterers
440 are contributing to the measured amount of acoustic backscatter. This result
441 was not surprising since turbulent microstructure was predicted to contribute
442 greatly to the measured backscatter for some portions of the water column.

443 [Figures 8 and 9 here]

444 A similar analysis was performed for biomass and biologically-caused
445 backscatter (Figure 9). When backscatter attributed to physical processes
446 was removed, the relationship between log measures of biomass and acoustic
447 backscatter was more linear. Regression coefficients improved for all frequen-
448 cies indicating a better correlation between biomass and backscatter. It must
449 be noted that the regression coefficients for each frequency are still fairly small
450 (r^2 ranged from 0.38 to 0.52), however these values are a factor of two or three
451 larger than if the source of the backscatter is not identified. By accounting for
452 the source of acoustic backscatter using hydrographic and net tow informa-
453 tion, this method can be used to improve the use of acoustic backscatter data
454 as a measure of zooplankton biomass.

455 Discussion

456 One of the goals of acoustic surveys is to estimate zooplankton biomass
457 and this requires that the relationship between biomass and acoustic backscat-

458 ter is well understood. The data presented here indicate that improvements can
459 be made in the interpretation of field collected survey data if the contributions
460 of all scattering sources are quantified. The relative importance of physical and
461 biological sources of acoustic backscatter will vary with location in the ocean
462 and certainly some regions will not have substantial backscatter from physi-
463 cal processes in the water column while other areas (such as the sites in this
464 study which have internal waves present) will have significant contributions
465 to the backscatter from non-biological sources. The modeling efforts outlined
466 in this work provide one way of determining if physical sources of backscatter
467 will need to be accounted for when interpreting acoustic backscatter survey
468 data. However, these improvements are just one step of many that need to
469 be taken in order that acoustic surveys may provide estimates of zooplank-
470 ton abundance that are accurate and ecologically useful. Given that biomass
471 and predictions of biologically-caused backscatter are not perfectly correlated,
472 sources of error in the analysis, such as inaccuracies in the backscatter models
473 used, must be examined.

474 The zooplankton backscatter models for fluid-like, elastic-shelled, and
475 gas-bearing zooplankton have been used previously in the analysis of field-
476 collected data (Wiebe et al. 1996, 1997; Greene et al. 1998), however there
477 are many variables used in these models that are inadequately understood
478 such as animal behavior and orientation or the material properties (g and
479 h) of the zooplankton. A better understanding of the scattering model inputs

480 would reduce errors associated with these types of animals. Furthermore, there
481 are numerous types of animals (salps, polychaetes, chaetognaths, gelatinous
482 zooplankton) whose backscatter characteristics have neither been modeled or
483 measured in a laboratory environment.

484 Uncertainty about the inputs to the backscatter models is a concern
485 for the microstructure model as well. Proper instrumentation was not present
486 to measure the dissipation rates of turbulent kinetic energy and temperature
487 variance, which are vital inputs into the theoretical microstructure backscatter
488 models, so the method used to estimate these inputs was not ideal. While this
489 method provided reasonable estimates of ϵ and χ , it likely overestimated the
490 scattering contributions from microstructure. For example, some regions of
491 MOCNESS # 9 have microstructure-caused s_V values that were larger than
492 the backscatter measured by BIOMAPER-II (Figure 5). Further complicating
493 this issue is the possibility that the vertical migration of animals may be
494 creating significant amounts of turbulence and mixing (Huntley and Zhou
495 2004; Kunze et al. 2006).

496 Other possible sources of error in the FP analysis include erroneous
497 zooplankton abundance and composition data and inaccurate measurements
498 of the acoustic backscatter. Net tow information from MOCNESS systems has
499 been used for several decades and sampling errors from it are likely limited to
500 net avoidance by large zooplankton (Wiebe et al. 2004) and gelatinous animals
501 being destroyed by the net mesh. Finally, the under-sampling of animals either

502 by nets (e.g. large euphausiids or small fish) or lower acoustic frequencies (e.g.
503 copepods) will cause errors in the FP analysis. One approach that has been
504 used in the analysis of acoustic scattering data (Warren et al. 2003) is to
505 use the net tow estimates of numerical density as a lower bound on the true
506 value (since you can not have more animals in a net than are present in
507 the water column) and use the acoustically-inferred estimates of biomass as
508 an upper bound (since these rely on measures of backscatter strength that
509 likely contain contributions from other scattering sources). In this manner,
510 combining acoustic and net tow data can provide an upper and lower estimate
511 of the abundance of zooplankton in the water column.

512 This study also demonstrates the importance of resolving the changes
513 in biological and physical backscatter sources within the water column. A mul-
514 tiple net system, or other method such as video or optical ground-truthing,
515 may be a necessary piece of equipment to accurately assess acoustic surveys
516 of zooplankton biomass, particularly where the taxonomic components of the
517 zooplankton community are diverse. Providing this vertical resolution and
518 ground-truthing of the acoustic data also allows us to observe partitioning
519 of the water column into different habitats that would not be apparent from
520 either the acoustic data alone or a vertically integrating net tow. While some
521 regions of the ocean do have patches with a single dominant taxa (e.g. Antarc-
522 tic krill), the variation in abundance and distribution of zooplankton taxa ob-
523 served over a 200 m vertical span in this study demonstrates the importance

524 of measuring and quantifying these changes.

525 The difference between predictions of S_V and those values that would
526 perfectly correlate with the biomass data are on the order of 5 - 10 dB (assum-
527 ing that the biomass data are accurate) (Figure 9). These differences become
528 very large when backscatter strengths are converted to estimates of biomass,
529 therefore these predictions result in estimates of zooplankton biomass that
530 are correct to roughly an order of magnitude. In certain cases this level of er-
531 ror may be acceptable, but further work is needed to reduce this uncertainty.
532 Without accounting for the source of acoustic scattering in the water column,
533 estimates of biomass from acoustics are likely to have even larger errors.

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Table 1

Location and time of the CTD vertical profiles and MOCNESS tows used in this study collected in October 1997 from the RV *Endeavor*.

Event	Julian Yearday	Latitude (N)	Longitude (W)	Begin/End
CTD 08	287.451	42° 14.97'	68° 44.77'	Begin
	287.467	42° 14.97'	68° 44.77'	End
MOC 07	287.620	42° 24.04'	68° 49.03'	Begin
	287.686	42° 24.93'	68° 44.22'	End
CTD 10	289.535	42° 25.08'	68° 44.49'	Begin
	289.562	42° 25.08'	68° 44.49'	End
MOC 09	289.896	42° 28.70'	68° 45.00'	Begin
	289.949	42° 30.97'	68° 46.69'	End

Table 2

Citations and parameters for the acoustic backscattering models used in the Forward Problem calculations. β_D is the length to width ratio ($\frac{L}{D}$) of the animal, R is the reflection coefficient.

Scatterer	Scattering Model Citation, Parameters
Euphausiids	Stanton et al. (1993a); Stanton and Chu (2000), $\beta_D = 5.3576$, R = 0.058, “head-up” orientation distribution ($20^\circ \pm 20^\circ$)
Copepods and Larval Crustaceans	Stanton et al. (1993a); Stanton and Chu (2000), $\beta_D = 2.5497$, R = 0.058, uniform orientation distribution ($0 - 360^\circ$)
Amphipods	Stanton et al. (1993a); Stanton and Chu (2000), $\beta_D = 3.0021$, R = 0.058, uniform orientation distribution ($0 - 360^\circ$)
Polychaetes and Chaetognaths	Stanton et al. (1993a); Stanton and Chu (2000), $\beta_D = 17.151$, R = 0.058, uniform orientation distribution ($0 - 360^\circ$)
Limacina Pteropods	Stanton et al. (1994), R = 0.5
Siphonophore Nectophores	Monger et al. (1998), R = 0.028
Siphonophore Pneumatophores	Anderson (1950)
Salps	Monger et al. (1998), R = 0.028
Microstructure	Seim (1999)

674 **Figure Captions**

675 Fig. 1. Total biomass estimated from MOCNESS # 7 (a) and 9 (b) data with
676 the relative contributions of the different taxonomic groups. Several taxa are
677 grouped together in the plot including: amphipods and other crustaceans in-
678 cluding larvae (“Crustaceans”); polychaetes, chaetognaths (“Worms”), pteropods
679 and gelatinous zooplankton (“Others”). The vertical thickness of the bar cor-
680 responds to the depth range sampled. For MOCNESS #7 the lower depths
681 were composed primarily of copepods and euphausiids with some gelatinous
682 animals, while the surface layers also contained small amounts of polychaetes,
683 chaetognaths, and siphonophore fragments. MOCNESS #9 sampled a large
684 sub-surface layer of salps (over 2500 animals m^{-3}) which dominated the biomass
685 sample. The remaining nets were composed of primarily copepods and eu-
686 phausiids.

687 Fig. 2. BIOMAPER-II echograms for 43, 120, 200, and 420 kHz for MOCNESS
688 # 7 (a) and 9 (b). The white line indicates the path of the net system, black
689 circles indicate where nets were opened and closed. Different regions of the
690 water column have different measured backscattering strengths for the various
691 frequencies. Remnants of an internal wave (undulating backscattering layers)
692 were observed during MOCNESS # 7, while MOCNESS # 9 measured a the
693 strong near-surface layer at 20 m depth which is seen most strongly in the 43
694 kHz echogram.

695 Fig. 3. Temperature, salinity and density profiles collected by the ESS system
696 onboard MOCNESS # 7 (a) and 9 (b). Regions of potential mixing and tur-
697 bulent microstructure are indicated by nearly-vertical or unstable sections of
698 the density profile such as 30 - 80 m for MOCNESS #7 and 0 - 20 m, 60 - 100
699 m, and 140 - 180 m for MOCNESS #9.

700 Fig. 4. Forward problem calculations for each class of scatterer for MOCNESS
701 #7. The different acoustic frequencies (43, 120, 200 and 420 kHz) are repre-
702 sented by squares, stars, circles, and diamonds respectively. Small copepods
703 had body lengths less than 2.5 mm. Data points that lie above the diago-
704 nal line indicate that the FP underestimates the scattering, while points be-
705 low the line are overestimates. Microstructure, siphonophore pneumatophores,
706 chaetognaths, and euphausiids are the strongest contributors to the predicted
707 levels of backscattering.

708 Fig. 5. Forward problem calculations for each class of scatterer for MOCNESS
709 #9. The different acoustic frequencies (43, 120, 200 and 420 kHz) are repre-
710 sented by squares, stars, circles, and diamonds respectively. Small copepods
711 had body lengths less than 2.5 mm. Data points that lie above the diagonal line
712 indicate that the FP underestimates the scattering, while points below the line
713 are overestimates. Microstructure, salps, siphonophore pneumatophores and
714 euphausiids are the strongest contributors to the predicted levels of backscat-
715 tering.

716 Fig. 6. Percentage breakdown of Forward Problem calculations for each net and
717 acoustic frequency of MOCNESS #7. Turbulent microstructure contributes
718 large amounts to the total backscattering in the mid-water regions, while eu-
719 phausiids backscattered a majority of the sound in the deeper water. Siphonophores
720 contribute greatly at the lower frequencies, but not at the higher ones. The
721 surface region (0 - 20 m) had a very diverse group of scatterers.

722 Fig. 7. Percentage breakdown of Forward Problem calculations for each net and
723 acoustic frequency of MOCNESS #9. Euphausiids dominate the backscatter-
724 ing in the mid-water depths, while turbulence contributes strongly both near
725 the bottom and near the surface. The salps which dominated the biomass in
726 the near-surface net, contribute only 30% to 60% to the total backscattering
727 for that region.

728 Fig. 8. The relationship between the logarithm of biomass and measurements
729 of acoustic backscatter strength for MOCNESS #7 (squares) and 9 (circles),
730 excluding net #7 from MOCNESS #9. Most frequencies show little correlation
731 between these two variables except for the highest frequency (420 kHz).

732 Fig. 9. The relationship between the logarithm of biomass and predictions of
733 biologically-caused acoustic backscatter for MOCNESS #7 (squares) and 9
734 (circles), excluding net #7 from MOCNESS #9. All the acoustic frequencies
735 show improved regressions and fairly linear relationships for the data.

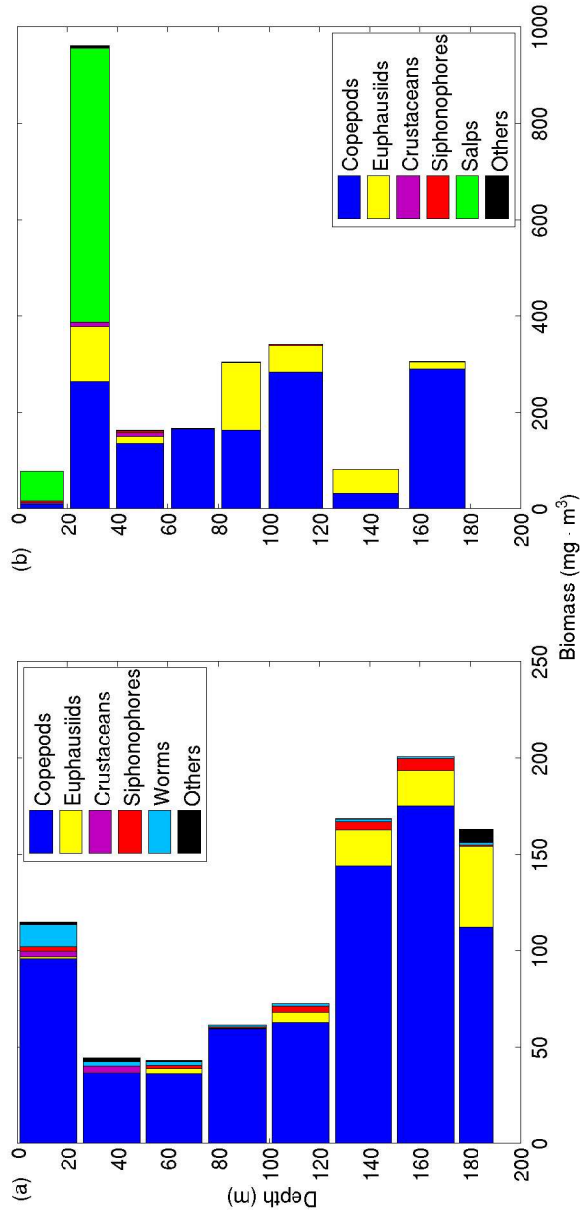


Fig. 1.

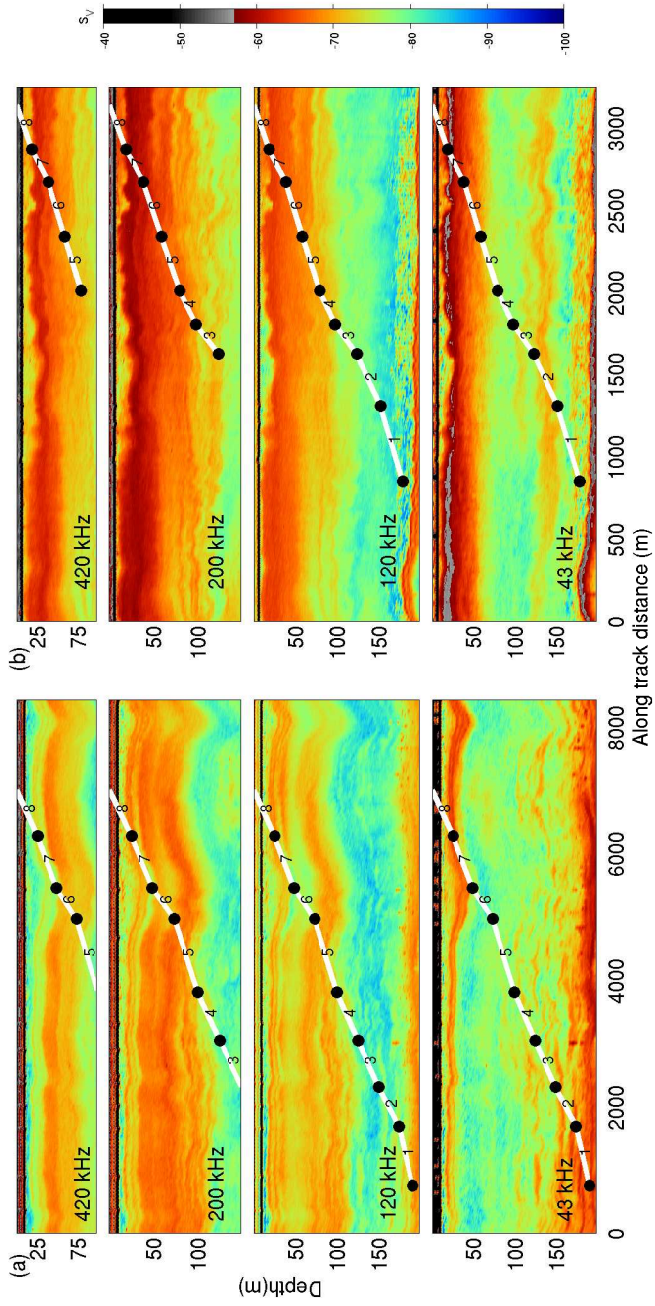


Fig. 2.

Fig. 3.
40

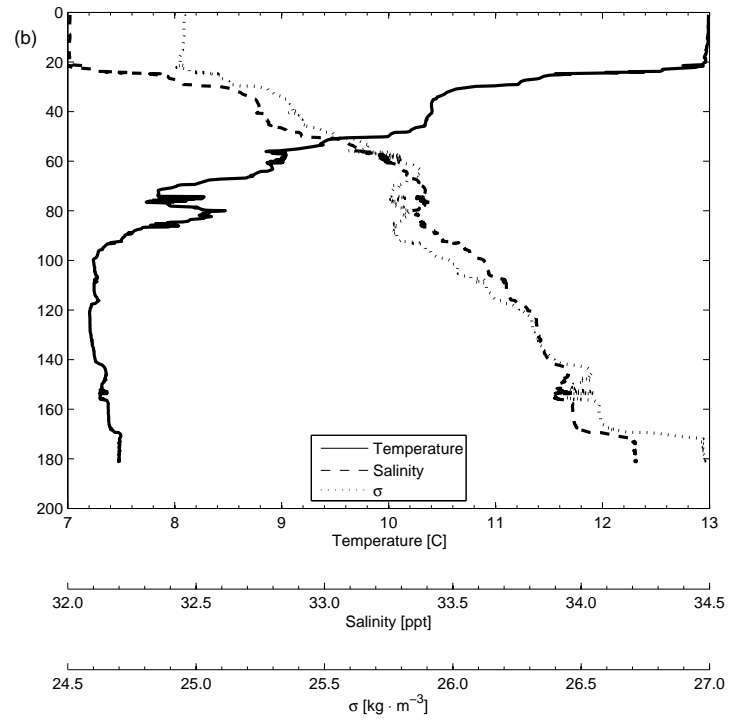
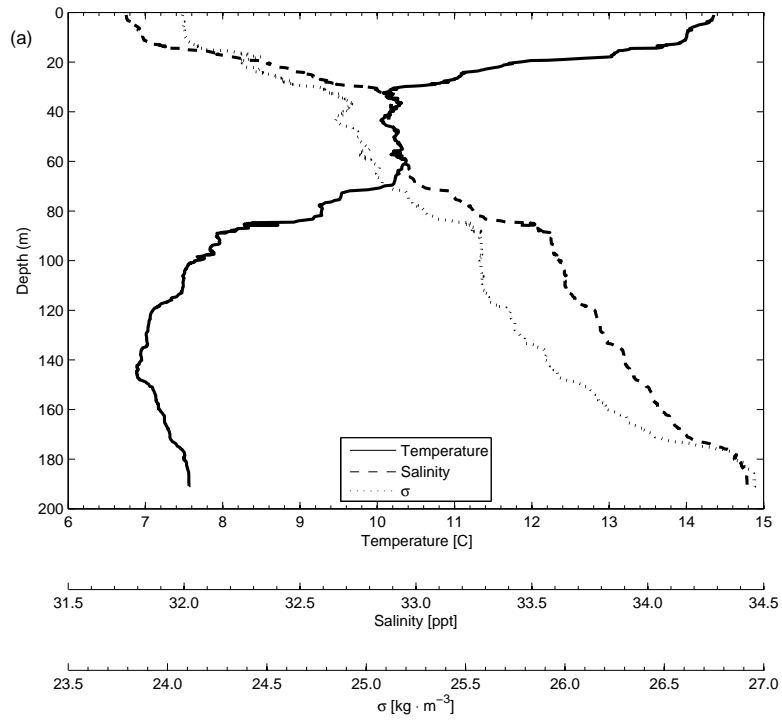


Fig. 4.

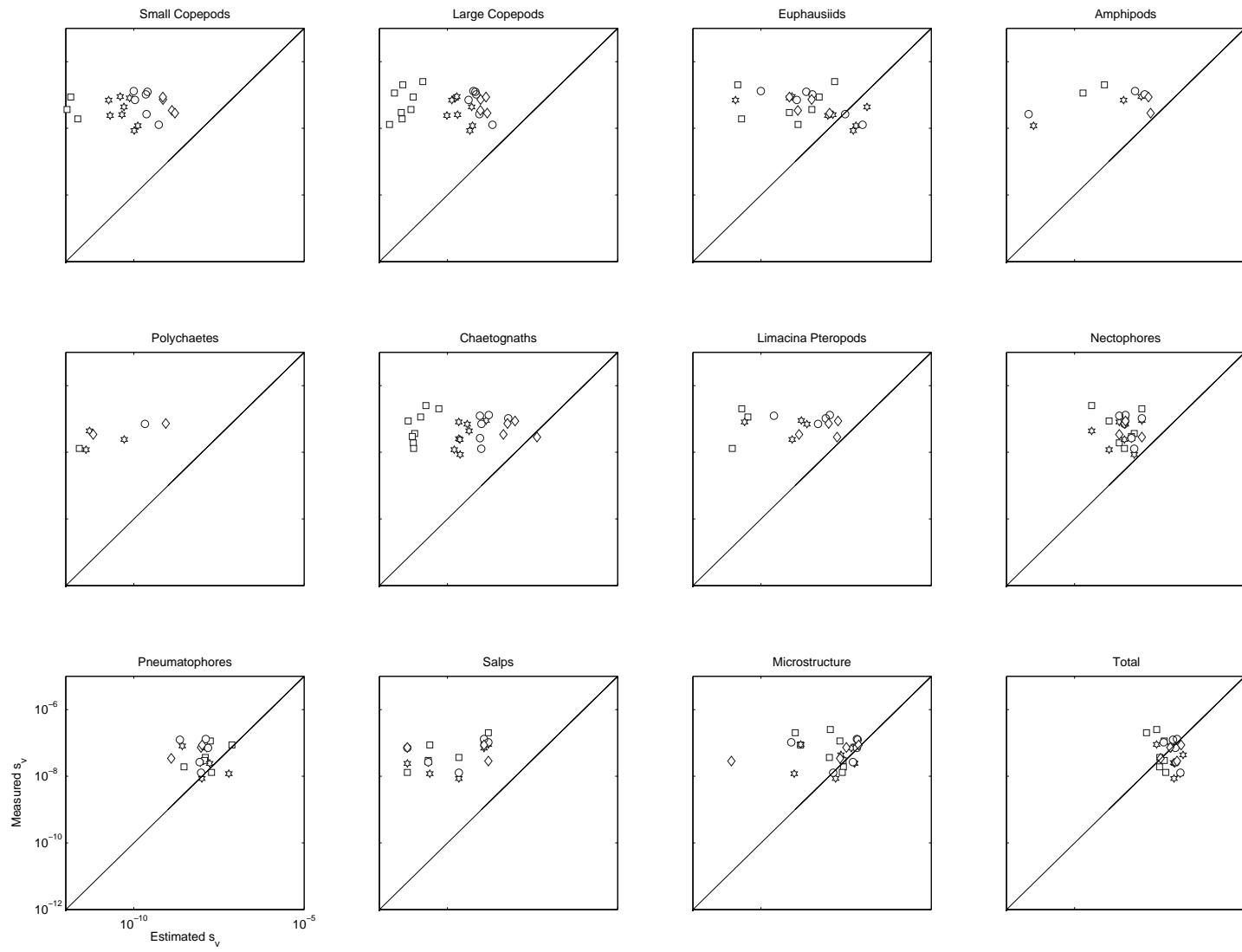
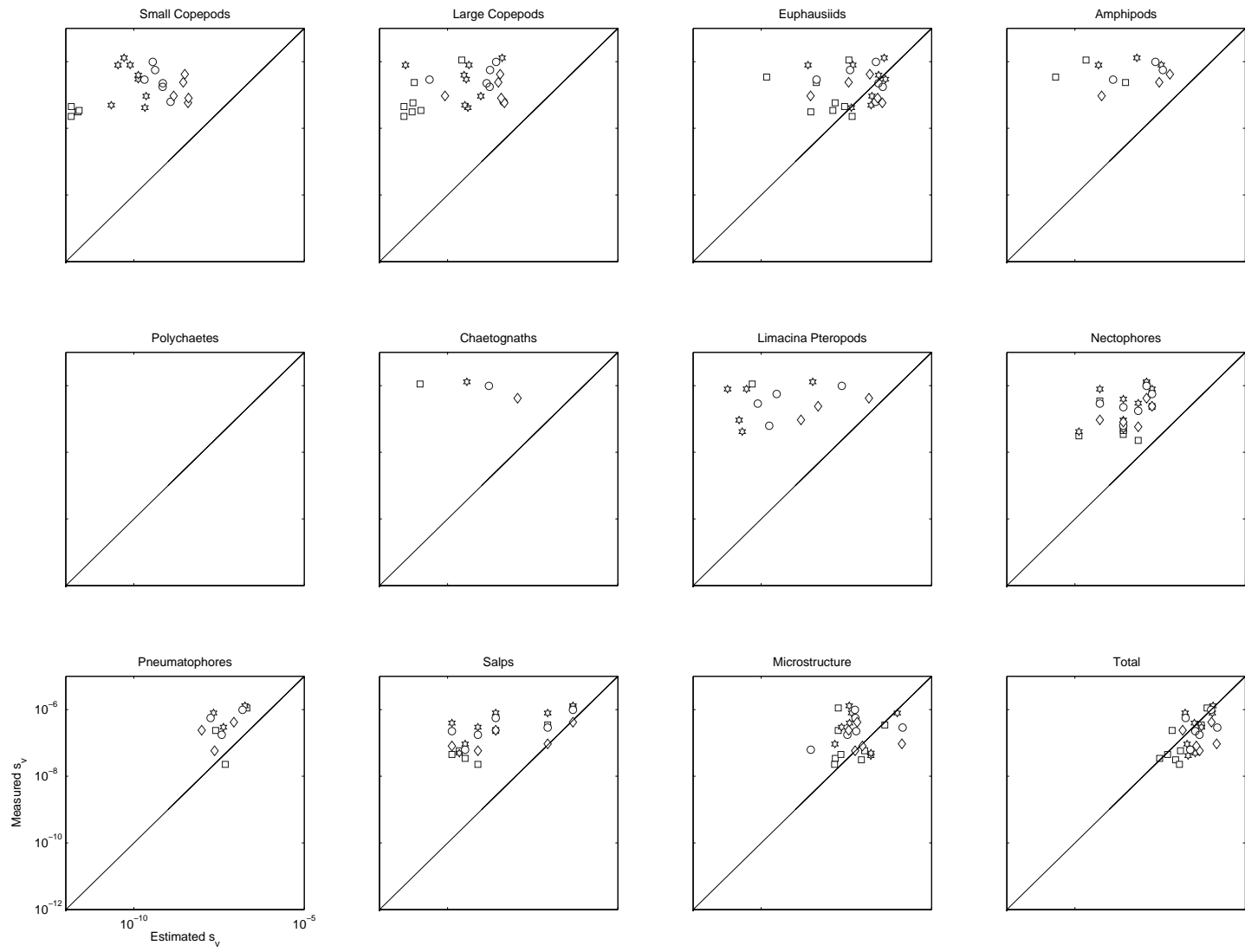


Fig. 5.



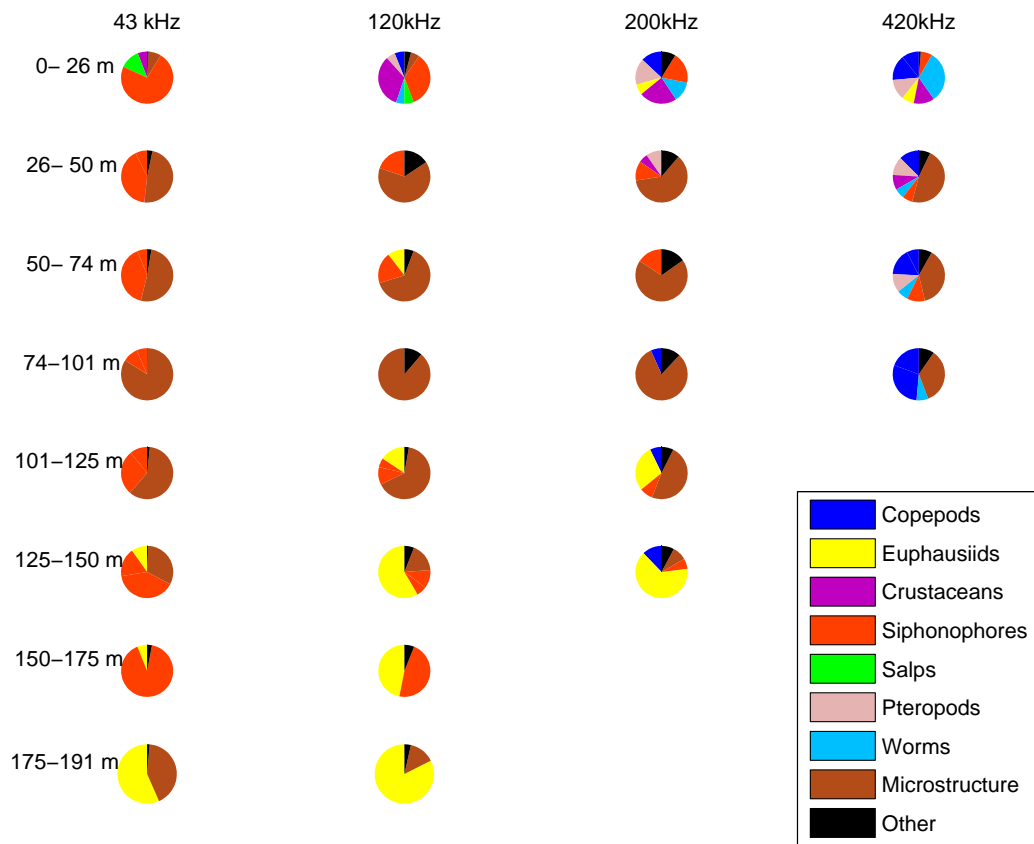


Fig. 6.

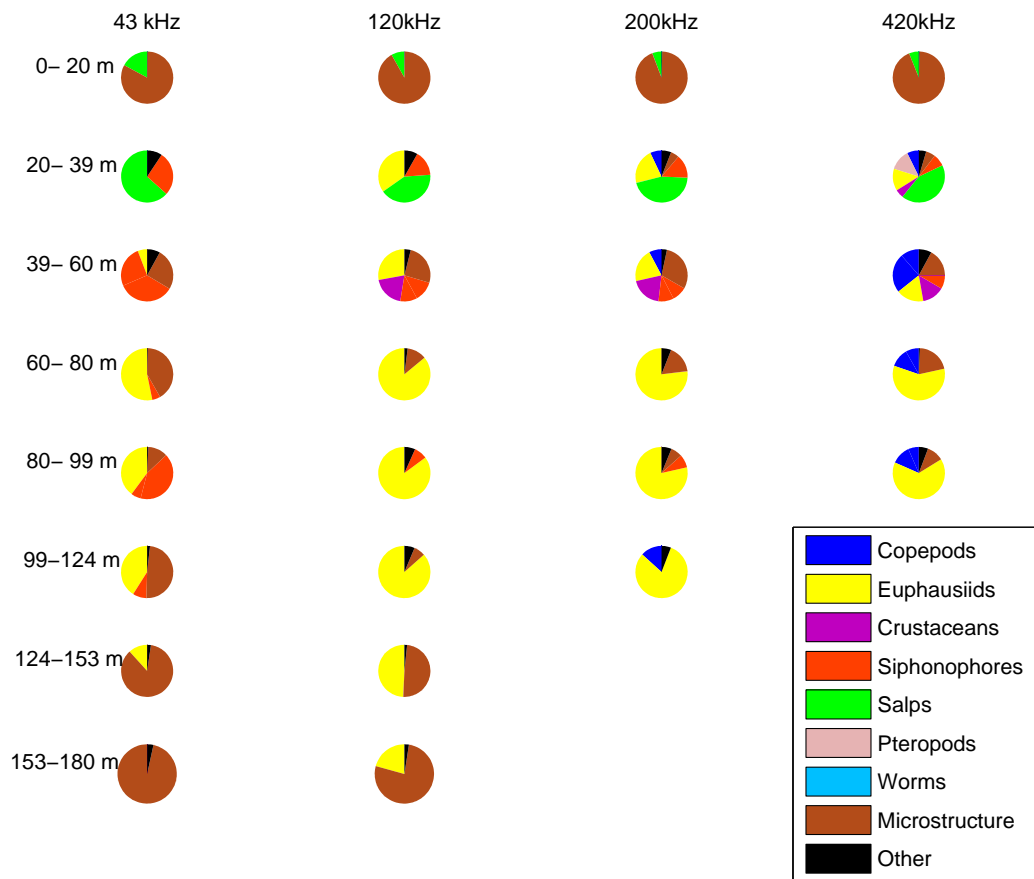


Fig. 7.

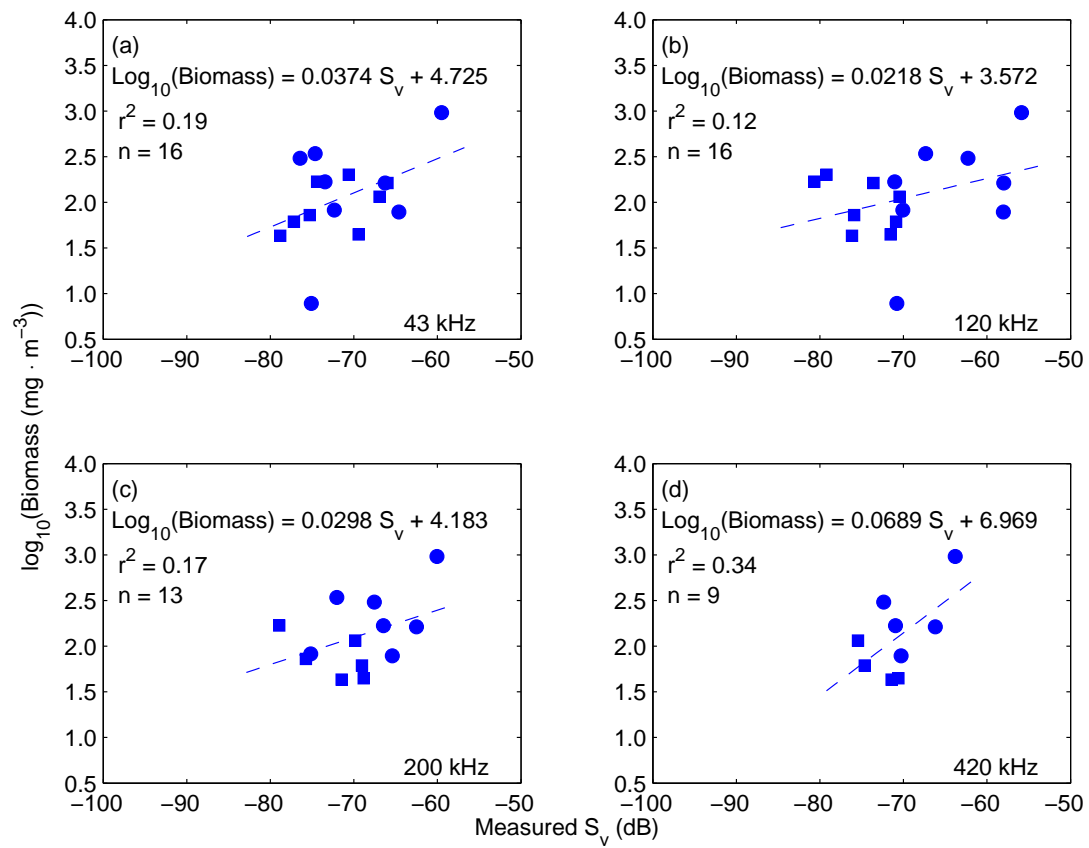


Fig. 8.

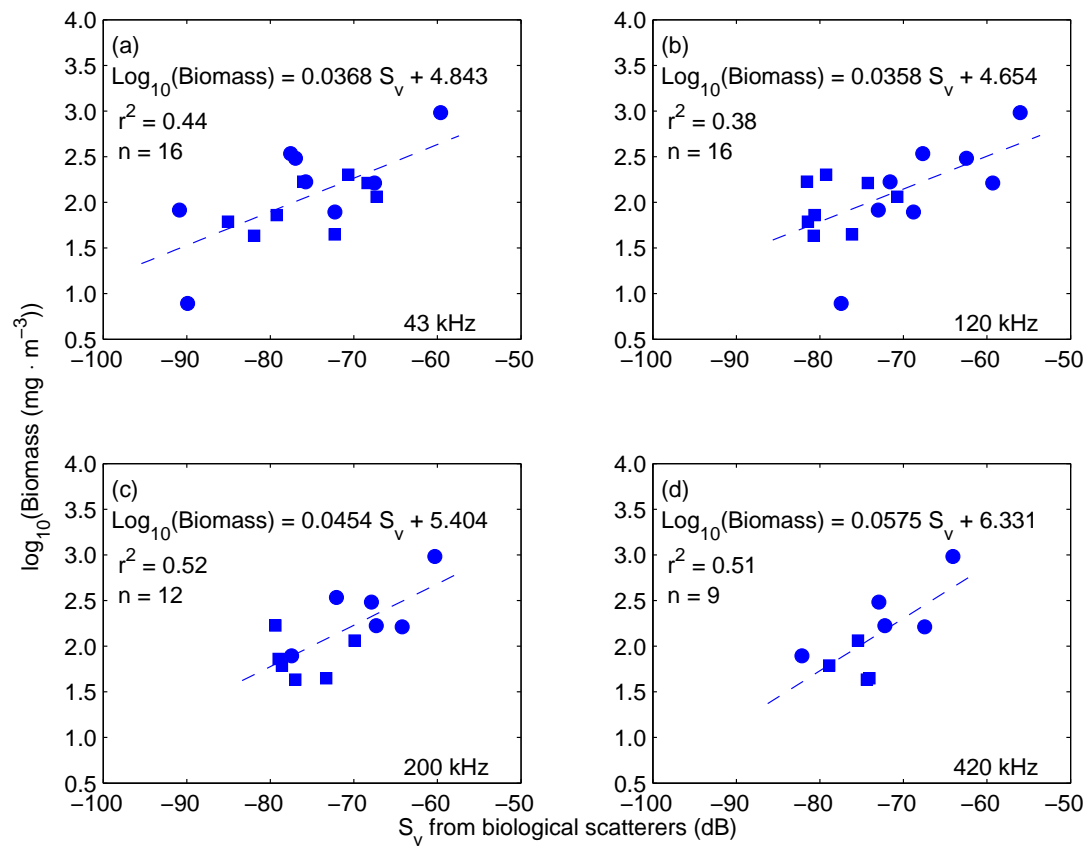


Fig. 9.