

Evaluating Performance of Photographs for Marine Citizen Science Applications

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Conflict of interest statement

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

Author contribution statement

All authors contributed to this study. Authors KN, BT, and AC contributed to the collection of data. Authors KN and BT compiled and analyzed all data, with suggestions and help from AC and GR. AC and GR provided financial support for the project. KN and BT wrote the manuscript. AC and GR edited the manuscript. KN prepared for submission.

Keywords

Photographic methods, marine, Invertebrates, non-native, citizen science, Taxonomy

Abstract

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Long-term measurements are imperative to detect, understand, and predict changes in coastal biological communities, but can be both costly and difficult to implement. Here, we compare measurement methods used to document community structure and assess changes in marine systems, and explore potential applications in citizen science. The use of photographs for species identifications and monitoring has become a popular and useful data collection tool, but its use requires evaluation of its effectiveness in comparison to data collected from live examinations. We used settlement panels in San Francisco Bay, a well-studied and vital coastal ecosystem, to compare standardized measures of the invertebrate fouling community through examination of live organisms in the field and via photographs. Overall, our study found that live measurements were more accurate and better represented these marine communities, having higher richness and diversity measurements than photographic measurements. However, photographic analyses accurately captured the relative abundances of some species and functional groups. We suggest that highly recognizable target taxa or broad scale comparisons of functional group composition are easily tracked through photographs and offer the best potential for research conducted by citizen scientists.

Contribution to the field

Citizen scientists have historically been undervalued as data collectors, however rising interest and increased attention to data quality have proven that properly managed public programs can collect robust and trustworthy data. Citizen scientists offer a potential solution to the problem of finding new invasive species, as professional taxonomists cannot unremittingly watch the world's coastline. Few studies have sought to verify whether marine invertebrates could be successfully monitored using public programs, such as photo-based surveys. This study took steps to identify potential future invasive species monitoring opportunities by ascertaining the best possible data collection opportunities from photographs and untrained taxonomists. Photography provides ample opportunity to extend monitoring programs to search for known invasive species and to survey communities for coastal ecosystem shifts. Our findings suggest citizen scientists can be employed to take and analyze photographs. Additionally, citizen scientists could be a potential resource to track target species and identify organisms to functional group. However, we report that species-specific measurement tools, like diversity and richness, cannot be approximated from photographs reliably for marine invertebrates.

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Ethics statements

(Authors are required to state the ethical considerations of their study in the manuscript, including for cases where the study was exempt from ethical approval procedures)

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In review

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7 **Keywords: Photographic Methods¹, Marine², Invertebrates³, Non-Native⁴, Citizen Sciences⁵,**
8 **Taxonomy⁶.**

9 **Abstract**

10 Long-term measurements are imperative to detect, understand, and predict changes in coastal
11 biological communities, but can be both costly and difficult to implement. Here, we compare
12 measurement methods used to document community structure and assess changes in marine systems,
13 and explore potential applications in citizen science. The use of photographs for species
14 identifications and monitoring has become a popular and useful data collection tool, but its use
15 requires evaluation of its effectiveness in comparison to data collected from live examinations. We
16 used settlement panels in San Francisco Bay, a well-studied and vital coastal ecosystem, to compare
17 standardized measures of the invertebrate fouling community through examination of live organisms
18 in the field and via photographs. Overall, our study found that live measurements were more accurate
19 and better represented these marine communities, having higher richness and diversity measurements
20 than photographic measurements. However, photographic analyses accurately captured the relative
21 abundances of some species and functional groups. We suggest that highly recognizable target taxa
22 or broad scale comparisons of functional group composition are easily tracked through photographs
23 and offer the best potential for research conducted by citizen scientists.

24 **1 Article type**

25 Methods

26 **2 Introduction**

27 Due to challenges presented by large-scale research efforts and the growing need to monitor our
28 coastal communities for threats from climate change, pollution, and invasive species (Ruiz et al.
29 1997; Stachowicz et al. 2002; Thiel et al. 2014), scientists have begun to develop and identify areas
30 where collaborations with citizen scientists would be most helpful (Dickinson et al. 2010). Citizen
31 science, or the involvement of the general public in collecting and analyzing scientific data, is an
32 increasingly important and useful approach to research that also broadens public engagement in
33 science. Though work of citizen scientists has historically been undervalued among academics
34 (Delaney et al. 2008), recent advances in communication technologies have made engaging citizen

35 scientists much easier, contributing to increased use (Bonney et al. 2014). Some past bias against
36 citizen scientists may be attributed to under-reporting of their efforts in research (Silverton 2009),
37 resulting in a lack of evidence supporting the use of data generated (Cooper et al. 2014). However,
38 citizen science has long been prevalent in the fields of ornithology (Dickinson et al. 2010) and
39 agriculture, among others (Miller-Rushing et al. 2012). As the use of citizen science has risen in the
40 past quarter century (Miller-Rushing et al. 2012), there is growing consensus that new citizen science
41 projects should carefully design questions and perform detailed analyses of data accuracy (Darwall &
42 Dulvy 1996; Boudreau & Yan 2004; Delaney et al. 2008; Fore et al. 2008; Silverton 2009; Dickinson
43 et al. 2010).

44 In order to create a lasting engagement with citizen scientists, it is necessary to use a method that is
45 both easily repeatable and quickly executed, such as the use of photographs to survey biological
46 communities. Photographs of organisms have been successfully used as a reliable tool to track
47 individuals over time (Frisch & Hobbs 2007; Carpentier et al. 2016). Some photographic
48 identification methods have become so advanced that computer-aided recognition methods allow for
49 automated comparisons (e.g. Melancon et al. 2011) or have inspired web and smartphone
50 applications to assist citizen scientists in identifying organisms in real-time (e.g. Kumar et al. 2012;
51 iNaturalist 2016). Not only could photographic comparisons give scientists the ability to identify
52 species or trends without the time constraints inherent to examination of live organisms in-situ, but
53 such approaches would allow anyone with a camera and enough interest to participate and contribute.

54 Monitoring for non-native species in particular has been identified as a good venue for citizen
55 scientists to make substantial contributions (Lodge et al. 2006; Cooper et al. 2014). Invasive species
56 are a leading threat to ecosystems across the globe (Stachowicz et al. 1999; Bax et al. 2001);
57 however, knowledge of the extent and effects of invasions in marine and coastal realms is still
58 deficient (Ruiz et al. 1999, 2011, 2015). Monitoring programs for biological invasions often have one
59 of two priorities: to be precise enough to detect arrivals of new species, which often initially appear
60 in small numbers, or to be broad enough to show changes over time, while remaining straightforward
61 in application and financial feasibility (Mangin 2001; Bax et al. 2001; 2003; Mantelatto et al. 2013;
62 USFWS 2015).

63 The largest and most diverse component of marine introduced species is comprised of invertebrates
64 (Ruiz et al. 2000), which, through local elimination of native species, are one of the significant
65 threats to marine ecosystems (Ruiz et al. 1997, 2000; Stachowicz et al. 1999, 2002; Grosholz et al.
66 2000; Carlton 2001; Blum et al. 2007). Citizen scientists have contributed to a broad range of
67 published and unpublished aquatic invasive species research (Boudreau & Yan 2004; Delaney et al.
68 2008; Crall et al. 2010; Azzurro et al. 2013; Zenetos et al. 2013; Scyphers et al. 2014; Maistrello et
69 al. 2016), and photographic methods have proven successful for many larger taxa in both terrestrial
70 and aquatic habitats (e.g. Darwall & Dulvy 1996; Bray & Schramm 2001; Fore et al. 2008).
71 However, many studies of marine invertebrate communities have relied on photographic analyses
72 without assessing the accuracy of this method compared to live examination or traditional measures.
73 Notably, citizen science surveys of smaller marine invertebrate communities are rare, though studies
74 on groups like Porifera and Tunicata do exist (Thiel et al. 2014). Many scientists have expressed
75 skepticism of taxonomic identifications via photographs without examination of physical specimens
76 (e.g. Ceríaco et al. 2016). Due to this uncertainty, further research on such performance and possible
77 constraints is useful before launching a marine invertebrate-focused citizen science effort, to align
78 objectives and results.

79 In this study, we assessed the use of photographs to accurately characterize marine invertebrate
80 communities in order to design a citizen science program with the purpose of 1) detecting non-native
81 species (i.e. new arrivals) and 2) documenting whole community response to change (species
82 introduction, environmental disturbance, etc.). We tested the accuracy of photographs in comparison
83 to live, field-based analyses and evaluated different research questions to determine which are best
84 answered by volunteers without specific taxonomic expertise. We analyzed five years of data from
85 live examinations of marine invertebrate fouling communities on settlement panels from San
86 Francisco Bay and compared their performance to data gathered from photographs of the same panels
87 for multiple common ecological measurements: species richness and diversity, functional group
88 richness and diversity, relative abundance, and detection rates of known non-native species. Species
89 and functional group diversity and richness were evaluated for both live and photographic methods,
90 and the latter method was expected to be less comprehensive than live analyses, as well as skewed
91 towards over-representation of large, conspicuous species. We expected functional group
92 composition to be similar between methods but lower detection of target taxa for photographic versus
93 live analysis methods.

94 **3 Materials and Methods**

95 **3.1 Study area and field methods**

96 Most marine invertebrate invasions occur in hard substrate habitats (Ruiz et al. 1999, 2011, 2015),
97 and a common method for assessing the status of marine invasions is to use settlement panels that
98 serve as standardized, passive sampling devices (e.g. Wisely 1959; Sutherland & Karlson 1977; Dean
99 & Hurd 1980; Osman & Whitlatch 1995; Stachowicz et al. 1999). Settlement panels (hereafter
100 “panels”) have been widely adopted in fouling community and biological invasion surveys
101 (Sutherland 1974; Bax et al. 2003; Blum et al. 2007; Tracy & Reynolds 2014; Marraffini & Geller 2015;
102 Newcomer et al. 2018) and are ideally suited for photographic analyses, as they offer a relatively
103 small, standardized, and flat area that is easily photographed. Panels are also ideal for use by citizen
104 scientists, since their deployment is both simple and repeatable with minimal prior experience.

105 We deployed replicate panels (n=10) at ten sites (Supporting Information) per year throughout San
106 Francisco Bay, California, USA (37°42'30"N, 122°16'49"W) over a five-year timespan. Panels were
107 cut from grey 0.5 cm thick polyvinyl chloride (PVC) sheets to 14 x 14 cm squares, lightly sanded,
108 attached to bricks, and suspended horizontally (“face-down”) one-meter below floating docks during
109 the season of high larval recruitment and biomass accumulation (June to September) each year from
110 2012 through 2016. Of the 100 panels deployed each year, we randomly chose 40 panels per year
111 (n=200, across all years), without regard to site, for comparison.

112 After three months in the water, panels were removed and photographed with a Canon® EOS Rebel
113 T5 camera. Three measurement methods were compared on each of the 200 panels: live point counts,
114 photo-based point counts, and an exhaustive live search. Species lists were therefore compiled from
115 the three methods, which were conducted as follows.

116 **3.2 Live in-field settlement panel point count**

117 Once photographed, each panel was examined live with a point count grid under a dissecting
118 microscope. Individual organisms attached to the panel directly underneath grid intersections were
119 morphologically identified to lowest taxonomic level, for a total 50 recorded points. Any sessile
120 species under the point was recorded. Points with more than one organism settled atop of each other
121 were recorded as two or more points, giving some panels >50 points.

122 3.3 Settlement panel photographic analysis

123 To mimic the live point count protocol, photographs were cropped to contain just the panel and
 124 scaled so that all images were of equal dimensions and resolution. A ‘digital point count grid’
 125 consisting of uniform intersections that mirrored the physical point count grid was then overlaid on
 126 the panel photograph using image-processing program ImageJ 1.8.0 (Abràmoff et al. 2004). We
 127 identified organisms directly underneath grid intersections to lowest taxonomic level possible (i.e.
 128 species, genus, family, etc.). When evident that one organism was settled atop of another under the
 129 point, both species were recorded as points, giving some panels >50 points.

130 3.4 Total species list verification

131 For each live panel, a researcher from our team of trained scientists (invertebrate parataxonomists)
 132 conducted an exhaustive search of the entire panel and identified and vouchered all discernable taxa.
 133 This search was verified by a second investigator. Vouchered samples were later re-verified by
 134 taxonomic experts. So-called ‘total observed’ species lists were then compiled from these expert
 135 identifications and reflect the best possible identification for every sessile species identified on each
 136 panel, including species that were not observed during point counts.

137 3.5 Evaluating potential citizen science research questions

138 Previous studies have noted that citizen science efforts to identify species might be better directed
 139 into functional groups based on multiple, easily recognized characteristics (Lodge et al. 2006; Thiel
 140 et al. 2014). Therefore, species were also classified into coarse functional groups (Supplementary
 141 Material). Functional groupings allowed researchers to compare within and between groups and
 142 calculate a conservative estimate for richness and diversity scores. Additionally, classifying
 143 identifications by functional group allowed researchers to compare the usefulness of group-level to
 144 species-specific scoring, as identifications generated by future citizen science projects are likely to be
 145 of lower resolution (less specific) than those collected by expert taxonomists.

146 We identified four non-native target taxa, or species of interest to scientists and policy-makers that
 147 are known to occur in San Francisco Bay and are spreading to other global regions. Previous studies
 148 completed by citizen scientists have used such targeted species search lists successfully (Darwall &
 149 Dulvy 1996; Boudreau & Yan 2004; Delaney et al. 2008). We chose target taxa that span four major
 150 functional groups and represent known species of interest: ‘Encrusting Bryozoa’, *Watersipora*
 151 *subatra* (Ortman, 1890); ‘Soft Bryozoa’, *Amathia verticillata* (delle Chiaje, 1822); ‘Solitary
 152 Tunicata’, *Ciona* spp. Fleming, 1822; and ‘Colonial Tunicata’, Botryllinae.

153 3.6 Data and model analysis

154 We compared common ecological measurements, including richness, diversity, abundance, and
 155 detection rate, which are often used in community surveys and citizen science-led research (Canning-
 156 Clode et al. 2008; Cooper et al. 2014; Thiel et al. 2014).

157 Statistical analyses were performed in the R statistical computing environment (R Core Team 2015)
 158 with the lme4 package (Bates et al. 2014), MuMIn package (Bartoń 2015), lsmeans package (Lenth
 159 2016), boot package (Davison & Hinkley 1997; Canty & Ripley 2017), and stats package (R Core
 160 Team 2015). We performed analyses with the methods outlined below on identifications made to
 161 functional group-level, as well as the lowest possible level (usually species level). Shannon-Wiener
 162 diversity indices (Shannon & Weaver 1948) and taxonomic richness were evaluated for each panel at

163 both group level as well as the lowest possible level.

164 We evaluated species and group-level diversity indices for each panel as a function of the fixed effect
 165 measurement method and used linear mixed models in a normal distribution. We evaluated species
 166 richness and functional group richness with a generalized linear mixed model (GLMM) using a
 167 Poisson error distribution. Panel was included as a random factor in richness and diversity models to
 168 account for the “repeated” measurement, as richness and diversity scores of the same panel would
 169 predictably be more related than different panels. Panel was nested within Year, another random
 170 effect in the model. Diversity scores were calculated from photograph and live point count data, but
 171 cannot be obtained from total observed species lists, as species list data does not supply the relative
 172 abundance of species, which is needed to calculate the Shannon-Wiener diversity index. Therefore,
 173 while richness measurements are compared in the models between photographic, live, and total
 174 observed scores, diversity models only compare the two point count methods. We used the lsmmeans
 175 package to conduct a three-way pairwise analysis on richness measurement types. Abundance was
 176 analyzed by functional group in a GLMM with a Poisson distribution, with each group’s field
 177 abundance evaluated against the fixed effect of photo abundance. Site and year were included as
 178 random factors. For all models, we calculated pseudo marginal and conditional r^2 values with an
 179 adapted r-squared formula for GLMMs in R package MuMIn. Residuals were plotted to verify model
 180 fit. Models were compared by their Akaike Information Criterion (AIC) value (Sakamoto et al.
 181 1986).

182 Tukey’s mean difference analyses, or Bland-Altman agreement analyses, were also used to assess the
 183 agreement and the strength of the relationship between our two point count methods (Bland &
 184 Altman 1999; 2003; Giavarina 2015). In the case of non-normally distributed differences, confidence
 185 intervals (95%) and the limits of agreement (1.97 x SD) were bootstrapped (DiCiccio & Efron 1996).
 186 In order for the photo-based method to have been considered comparable to the live method, 90% of
 187 the sample needed to fall within the limit of agreement (LOA).

188 For each year, live and photograph-based point count methods were compared using species
 189 accumulation curves. Estimated richness was calculated using the second-order jack-knife variant
 190 (Canning-Clode et al. 2008). Species accumulation curves were compared graphically in R statistical
 191 computing environment (R Core Team 2015) using the vegan package (Oksanen et al. 2007).

192 **4 Results**

193 **4.1 Species Richness and Diversity**

194 Photo-based analyses produced lower richness counts than field-based analyses (Table 1; Figure 1),
 195 and measurement type was responsible for 55% of differences in richness scores, according to best fit
 196 models (2919 AIC, -0.38 estimate, $p < 0.01$; Supplementary Material). Species richness was highest
 197 in total observed measurements, while live point count measurements detected 78% of distinct
 198 species across all years, and photos found 41%, significantly less in each case according to best fit
 199 models (Supplementary Material). Nearly two-thirds of richness measurements from photos counted
 200 only 40% of total species or fewer (Figure 2). When directly compared, photo-based richness scores
 201 were representative of live point count richness scores, though significantly different with an average
 202 of 2.5 species not counted in photos (94.5% within LOA; Supplementary Material). The performance
 203 of both point count methods declined as total richness increased (Figure 2).

204 Increased sampling effort (i.e. more panels) would likely not increase richness found via photographs
 205 to levels observed in live point counts (Figure 3). Live point count analyses can be used to accurately

206 estimate total observed richness using extrapolation, as extrapolated richness from the live point
 207 counts were not statistically different from in-situ richness, however photo-based point count
 208 analyses cannot approximate true richness, as the extrapolated scores remained statistically different
 209 (Table 1; mean $29.8 \pm \text{CI } 13.8$, mean $-4.4 \pm \text{CI } 10.9$).

210 Across all years analyzed, 31 species were observed in the field that were not identified in
 211 photographs (Supplementary Material). ‘Kamptozoa’ were completely absent from photographic
 212 point counts. Furthermore, functional groups ‘Anthozoa’, ‘Cirripedia’, and ‘Hydrozoa’, as well as
 213 families Sabellidae, Serpulidae, Spirorbidae, and Terebellidae, could not be identified to lower
 214 taxonomic resolution via photographs, thus missing at least 20 distinct species that were identified
 215 from live point counts and total observed methods.

216 Diversity measurements from photographs were on average 0.36 times lower than live measurements
 217 (Figure 1; Supplementary Material). Measurement type explained 13% of the difference in diversity
 218 scoring, according to best fit models (474 AIC, $p < 0.01$; Supplementary Material). The two point
 219 count based diversity scores were found to be statistically different, with only 75% of scores within
 220 the LOA, thus failing to meet the standard to consider photo-based diversity scoring as representative
 221 of live scores (Supplementary Material).

222 4.2 Functional Group Richness and Diversity

223 Photo-based measurements of functional group richness were significantly lower than live point
 224 count and total observed richness scores, and measurement type explained 33% of this difference
 225 between scores, according to best fit models (2214 AIC, $p < 0.01$; Supplementary Material). When
 226 directly compared, photo point count functional group richness scores were significantly different
 227 than live point count scores, but fell within the limit of agreement (97%; Supplementary Material).
 228 Photographic functional group richness scores were 0.24 times lower on average than those taken in
 229 the field (Supplementary Material; Figure 1).

230 Of the highly specious functional groups, identifications within ‘Branching Bryozoa’, ‘Encrusting
 231 Bryozoa’, and ‘Solitary Tunicata’ noted similar numbers of unique species from both photo and live
 232 analyses. For these four groups, >90% of measurements fall within the LOA for the number of
 233 species within a functional group (Supplementary Material).

234 Functional group identification lists were similar for all years with the exception of ‘Soft Bryozoa’,
 235 ‘Anthozoa’, and ‘Hydrozoa’, which appeared in some years but were absent in others, and
 236 ‘Kamptozoa’, which never appeared in photos. Functional groups were sampled less frequently from
 237 photographs when compared to live point counts, except for ‘Branching Bryozoa’ and ‘Soft
 238 Bryozoa’, which are oversampled in photographs due to their dominant presence (e.g. greater relative
 239 height, broad canopy). The abundances of functional groups, as well as associated presence detection
 240 rate, are comparable between live and photographic methods, with notable similarities in the most
 241 abundant categories in all years. ‘Branching Bryozoa’, ‘Soft Bryozoa’, ‘Solitary Tunicata’, ‘Colonial
 242 Tunicata’, and ‘Cnidaria’ were all accurately captured from photos based on their limits of agreement
 243 (Supplementary Material). ‘Branching Bryozoa’ and ‘Soft Bryozoa’ abundances were not statistically
 244 different between point count methods (Supplementary Material). The two methods could not be
 245 evaluated for the remaining groups (‘Bivalvia’, ‘Encrusting Bryozoa’, ‘Cirripedia’, ‘Hydrozoa’,
 246 ‘Polychaeta’, and ‘Porifera’), as the abundances of those groups did not meet assumptions of the
 247 mean difference tests, likely due to sparse abundance.

248 ‘Solitary Tunicata’ (estimate = 0.03, $r^2 = 0.22$, $p < 0.01$), ‘Colonial Tunicata’ (0.04, $r^2 = 0.20$, p

249 <0.01), and ‘Branching Bryozoa’ (0.05, $r^2 = 0.19$, $p < 0.01$) abundances were most correlated
 250 between methods, according to best fit models (Supplementary Material; Figure 4). ‘Cnidaria’
 251 models for abundance did not improve upon the null. All other functional group abundances were not
 252 correlated ($r^2 < 0.10$; Supplementary Material).

253 Photo-based measurements of functional group diversity were 0.18 times lower on average than those
 254 gathered from live point counts (Figure 1; Supplementary Material). Measurement type explained 6%
 255 of this difference in diversity score, according to best fit models (240 AIC, $p < 0.01$; Supplementary
 256 Material). Methods were found to be statistically different, though 94.5% of scores were within the
 257 limits of agreement (Supplementary Material).

258 4.3 Target Taxa

259 For every year of this study they appeared, all target taxa were found using all three methods.
 260 However, detection rates of *Ciona* spp. and *A. verticillata* were similar between point count methods,
 261 while photos captured significantly less *W. subatra* and Botryllinae than live measures (Figure 5).
 262 The detection rate of *Ciona* spp. by either point count method most closely approximated its true
 263 frequency compared to any of the other target taxa in San Francisco (Figure 5).

264 Non-native bryozoan *W. subatra* was found on an average of 47% of panels per year, according to
 265 the total observed species lists. Live point counts identified the bryozoan in 56% of these occurrences
 266 (SE 9%) and photo-based point counts 25% of the occurrences (SE 4%). Non-native tunicates *Ciona*
 267 spp. were found on an average of 77% of panels per year. Live point counts also identified the
 268 tunicates 96% of the time (SE 2%) and photo-based point counts 92% of the time (SE 5%).
 269 Botryllinae, a Tunicata subfamily and common known non-native species, were found on an average
 270 of 90% of panels per year. Live point counts also identified the tunicates 87% of the time (SE 3%)
 271 and photo-based point counts 74% of the time (SE 5%). The non-native bryozoan *A. verticillata* was
 272 found on an average of 30% of panels per year for the three years it appeared in San Francisco. Live
 273 point counts also identified the bryozoan 56% of the time (SE 9%) and photo-based point counts 43%
 274 of the time (SE 6%).

275 5 Discussion

276 5.1 Richness and diversity

277 Our results indicate that richness and diversity scores recorded from photographs are not fully
 278 representative of the richness and diversity recorded by experts using microscopic examination of
 279 live samples. There was not a simple reduction in overall diversity that would allow researchers to
 280 use photographic analyses to consistently and accurately estimate the live diversity measured by
 281 microscopic examination. Although species richness was related between photo and live point count
 282 scores, photos were not representative of the in-situ total measurements. The relationship of diversity
 283 and richness scores between live and photo analyses might be influenced by the functional groups
 284 that make up a sample. Some groups were systematically underrepresented in photographic point
 285 counts compared to live analyses, particularly those organisms that have inconspicuous or small
 286 mature individuals, like kamptozoans, while abundances of larger, easily discernable groups, like
 287 arborescent bryozoans and colonial and solitary tunicates, were well approximated.

288 5.2 Species composition

289 Species lists amassed from photographic methods are likely to omit significant numbers of taxa. In

290 our analysis, photograph-based point counts accurately captured just 41% of distinct species found
 291 from our total species analysis method. Critically, species accumulation curves constructed from
 292 photographic data did not predict more species discovery with continued effort. Live point counts
 293 were more representative of total richness, at 78% of total distinct species identified, but often
 294 overlooked rare organisms. An increase in effort for live point counts (more panels) might increase
 295 the number of species found closer to the total observed, though our analyses found that extrapolated
 296 richness estimates from live point counts already produced comparable estimates to the total
 297 observed richness. Some variation in the number of distinct species could be attributed to the
 298 individual bias of the observer. However, these results suggest that live point counts can be a useful
 299 tool for rapid surveys.

300 Photographic methods performed best with easily recognizable species, including many of interest to
 301 scientists and managers (e.g. colonial tunicates of the family Didemnidae; Valentine et al. 2009;
 302 McCann et al. 2013; Ojaveer et al. 2015). Target taxa examined in this study showed that
 303 presence/absence trends follow the same pattern between photos and live analyses, and every species
 304 was found in photographs, though each species was detected less frequently from photos than from
 305 live analyses. Taxa having the closest correlation between live and photograph abundances and
 306 detection rates were usually larger-bodied species, particularly tunicates and arborescent bryozoans.
 307 This high correlation could be due to their size, but could also be partially attributed to their ability to
 308 “stand out” from fouling community counterparts (e.g. distinct coloration and shape), making them
 309 easier to recognize and capture in data from a photograph. For these reasons, we expect that detection
 310 of target taxa is generally most reliable among highly recognizable groups (‘Solitary Tunicata’,
 311 ‘Colonial Tunicata’, and ‘Branching Bryozoa’).

312 Studies utilizing target taxa span a wide breadth of ecological purposes – from conserving
 313 endangered species (Greenemeier 2017) to monitoring water quality (Carroll et al. 2009; Zuykov et
 314 al. 2016). Many sessile marine invertebrates preferentially inhabit very specific environmental
 315 conditions (Chiarelli & Roccheri 2014; RAC/SPA 2015). If specific invertebrate species are
 316 identified as target taxa in a region, their use as bioindicators could help both scientists and managers
 317 to understand changes, or impending changes, in environmental conditions and ecosystem health
 318 (Ward & Larivière 2004). Our results suggest selection of target taxa from highly conspicuous and
 319 recognizable function groups will serve to provide the most reliable data and enable engagement of
 320 citizen scientists.

321 More broadly, we recognize many areas where photographs are useful in ecological research by
 322 scientists with taxonomic training. Other studies have used photographs to study the succession and
 323 growth rate of species onto bare space, usually by monitoring individual colonies over time (e.g.
 324 Tracy & Reynolds 2014). Since these studies do not rely on photography for taxonomic identification
 325 (colonies are identified to species level using live microscopic examination at some point during the
 326 study), they are an example of the successful use of photographs for species-specific analyses. Other
 327 studies have successfully used multiple images stitched together to enhance resolution (e.g. Lindeyer
 328 & Gittenberger 2011; Newcomer et al. 2018). Our study found that some organisms within well-
 329 photographed functional groups (e.g. ‘Branching Bryozoa’ and ‘Solitary Tunicata’) can be reliably
 330 separated into species from photographs. Moreover, it is also important to note that we focused on an
 331 area where fouling communities have high three-dimensional growth, and locations with less upright
 332 growth may have different results. We expect that photograph-based studies could create more
 333 accurate species lists if the communities were younger (organisms are smaller, e.g. Valentine et al.
 334 2009) or morphologically smaller (like in high latitudes) with very little physical overlap occurring
 335 between species, though smaller individuals would require high-resolution photographs and would

336 still lack microscopic examination of key species traits.

337 **5.3 Large scale trends and relative abundance**

338 Large abundance trends in functional groups were reliably captured by the photograph method. The
 339 best example of this in our data is the observed increase in abundance of ‘Solitary Tunicata’,
 340 specifically *Ciona* spp., in 2013 and an increase in ‘Soft Bryozoa’, specifically *A. verticillata*, in
 341 2015 in San Francisco Bay. Both changes appear to represent an organismal response to a significant
 342 increase in salinity during a major drought (2013–2015; Swain 2015). In this case, abundances for
 343 these species drastically affected the relative abundances of their functional groups in the respective
 344 years, allowing researchers to explore the change in community composition, identify the species
 345 responsible, and infer that the salinity shift was a potential cause (e.g. Chang et al. 2017). Future
 346 studies could profitably compare the classification of communities identified using photographic
 347 methods to those identified using live analyses. Such compositional analyses rely heavily on
 348 abundance information, which is one of the more reliable metrics that can be derived from
 349 photographs.

350 **5.4 Applications for citizen scientists**

351 Many citizen projects have adopted a high replicate model, finding that increased effort will
 352 compensate for less precise and less accurate identifications, eventually leading to comparable results
 353 (Kosmala et al. 2016; Swanson et al. 2016). However, our results indicate that in the case of marine
 354 invertebrates, photographs will continue to miss many rare and small species, even with increased
 355 effort (Figure 3). Additionally, professional scientists (parataxonomists) were used in this study to
 356 calculate richness from photographs, which suggests that non-expert citizen scientists would likely
 357 identify fewer species, resulting in even lower richness scores (Fore et al. 2008; Kremen et al. 2011).
 358 Thus, in programs that intend analysis by citizen scientists, we recommend that projects focus on
 359 gathering information at the level of functional groups, on limited target species, or on species within
 360 well-sampled large and conspicuous groups. Measures and experiments that rely on citizen scientists,
 361 or groups with variable taxonomic experience and training, must carefully design questions that do
 362 not rely on species-level community analyses.

363 **5.5 Recommendations for Photograph Use by Citizen Scientists:**

- 364 1. Invasive Species Monitoring
 - 365 a. Surveys for a limited number of known target taxa within the highly recognizable
 - 366 functional groups (e.g. ‘Colonial Tunicata’, ‘Branching Bryozoa’, and ‘Solitary
 - 367 Tunicata’)
 - 368 b. Surveys for a limited number of known target taxa of any functional group when
 - 369 panels are <1-month-old, or have little overlapping growth
- 370 2. Whole Community Surveys
 - 371 a. Surveys for species within one highly recognizable functional group (e.g. ‘Colonial
 - 372 Tunicata’, ‘Branching Bryozoa’, and ‘Solitary Tunicata’) when species are already
 - 373 known and readily described to volunteers
 - 374 b. Surveys for functional group abundance excluding challenging groups (‘Bivalvia’,
 - 375 ‘Encrusting Bryozoa’, ‘Cirripedia’, ‘Porifera’, ‘Kamptozoa’, ‘Hydrozoa’, ‘Anthozoa’,
 - 376 and ‘Polychaeta’)

377 The most reliable uses for photographic analyses identified by our study are the identification of
 378 specific target taxa, such as possible known invasive species, and the documentation of large shifts in

379 community structure. We suggest that photographs could be used for identifying recognizable
 380 invasive species, or for monitoring large community shifts over time that may serve as indicators of
 381 drastic environmental change, as functional groups are easily identified from photos. These best uses
 382 also reduce the expectation of citizen scientists to learn many species, and reduce the amount of
 383 training needed for new volunteers.

384 **6 Figures and Tables**

385 **6.1 Figure Legends**

386 **Figure 1.** Comparison of the spread of richness scores found by species identification (A) and
 387 functional group identification (B) and diversity scores found by species (C) and functional group
 388 (D) from all panels for both point count methods.

389 **Figure 2.** Comparison of richness scores from live (A) and photo (B) point count methods,
 390 represented as the percent of true richness found by either method, plotted against true richness.

391 **Figure 3.** Species accumulation curve for the two point count methods.

392 **Figure 4.** Abundance scores per panel found from live and photographed point counts for six
 393 functional groups: Solitary Tunicate (A), Branching Bryozoa (B), Encrusting Bryozoa (C), Colonial
 394 Tunicata (D), Cirripedia (E), and Soft Bryozoa (F).

395 **Figure 5.** Target taxa detection rate, shown as frequency found per point count method compared to
 396 true frequency, with asterisks that denote statistical differences between method for each target taxa.

397 **6.2 Tables**

398 **Table 1.** Richness scores calculated per year for the two measured point count methods, along with
 399 the extrapolated richness scores for the point count methods, the respective percentage of the total
 400 observed species those measurements reflect, and the measured total species present.

401

	Year	Observed Richness	% of Total	Extrapolated Richness	% of Total	Total Species Present
Photo	2012	22	28.9%	26.9	35.4%	76
	2013	25	49.0%	34.7	68.0%	51
	2014	28	40.6%	38.7	56.1%	69
	2015	31	33.0%	37.9	40.3%	94
	2016	27	28.1%	38.6	40.2%	96
Live	2012	47	61.8%	63.7	83.8%	76
	2013	36	70.6%	47.8	93.7%	51
	2014	46	66.7%	81.7	118.4%	69
	2015	51	54.3%	89.8	95.5%	94
	2016	40	41.7%	64.8	67.5%	96

402

403 **7 Contribution to the Field**

404 Citizen scientists have historically been undervalued as data collectors, however rising interest and
 405 increased attention to data quality have shown that properly managed public programs can collect
 406 robust and trustworthy data. Citizen scientists offer a potential solution to the problem of finding new
 407 non-native species, as professional taxonomists cannot unremittingly watch the world's coastline.
 408 Few studies have sought to verify whether marine invertebrates could be successfully monitored
 409 using public programs, such as photo-based surveys. This study took steps to identify potential future
 410 invasive species monitoring opportunities by ascertaining the best possible data collection
 411 opportunities from photographs and untrained taxonomists. Photography provides ample opportunity
 412 to extend monitoring programs to search for known invasive species and to survey communities for
 413 coastal ecosystem shifts. Our findings suggest citizen scientists can be employed to take and analyze
 414 photographs. Additionally, citizen scientists could be a potential resource to track target species and
 415 identify organisms to functional group. However, we report that species-specific measurement tools,
 416 like diversity and richness, cannot be approximated from photographs reliably for marine
 417 invertebrates.

418 **8 Conflict of Interest**

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

421 **9 Author Contributions**

422 All authors contributed to this study. Authors KN, BT, and AC contributed to the collection of data.
 423 Authors KN and BT compiled and analyzed all data, with suggestions and help from AC and GR. AC
 424 and GR provided financial support for the project. KN and BT wrote the manuscript. AC and GR
 425 edited the manuscript. KN prepared for submission.

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620 <https://doi.org/10.1016/j.chemosphere.2013.05.001>

621

622 **13 Data Availability Statement**

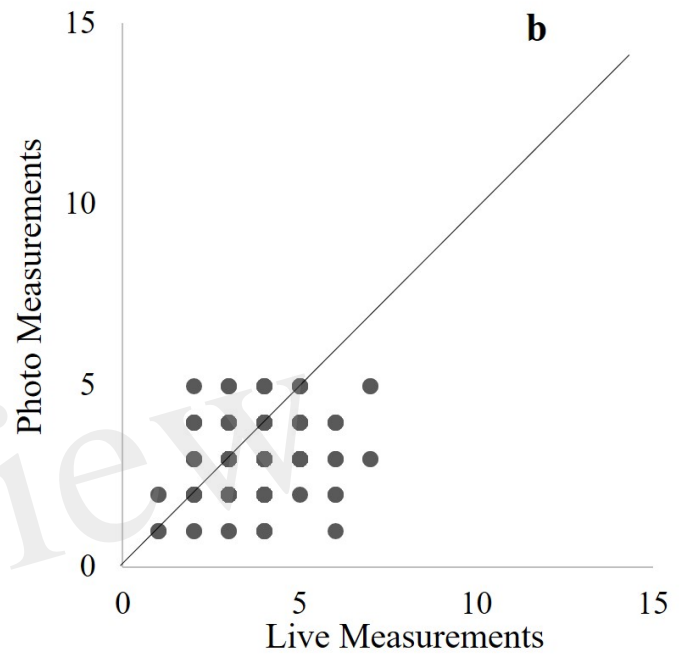
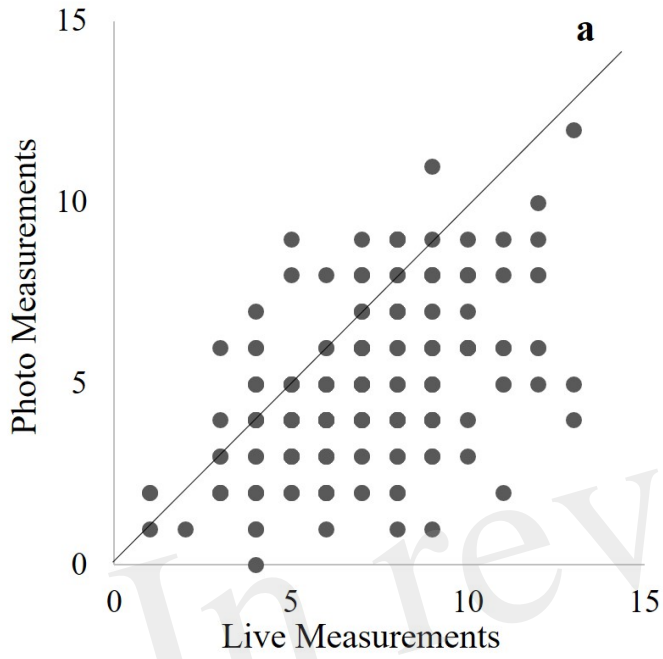
623 The raw data supporting the conclusions of this manuscript will be made available by the authors,
624 without undue reservation, to any qualified researcher.

In review

SPECIES LEVEL IDENTIFICATIONS

FUNCTIONAL GROUP IDENTIFICATIONS

RICHNESS



DIVERSITY

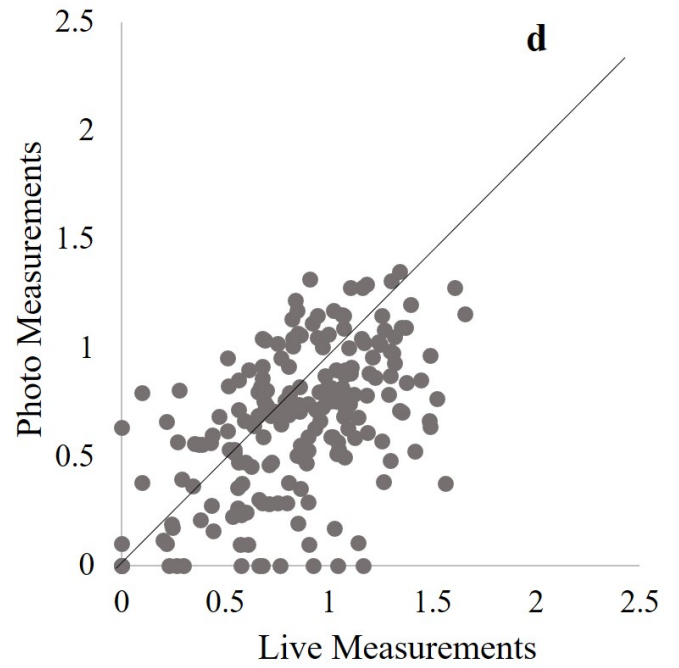
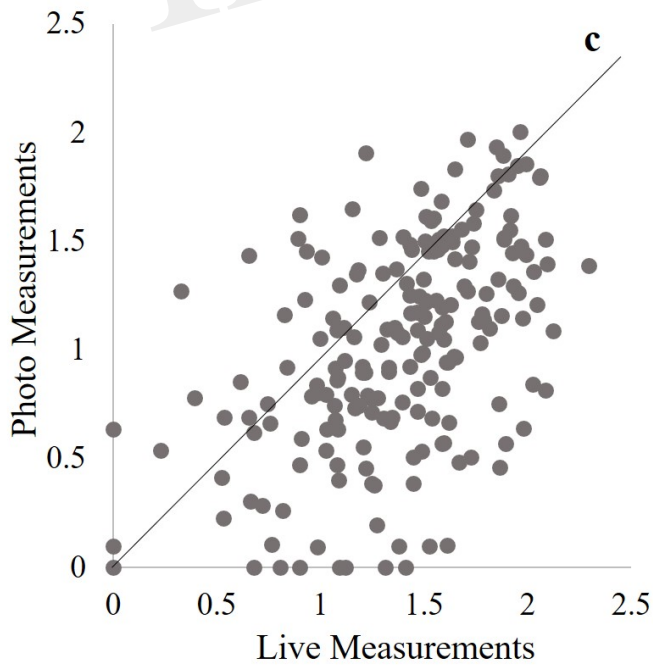


Figure 2.JPEG

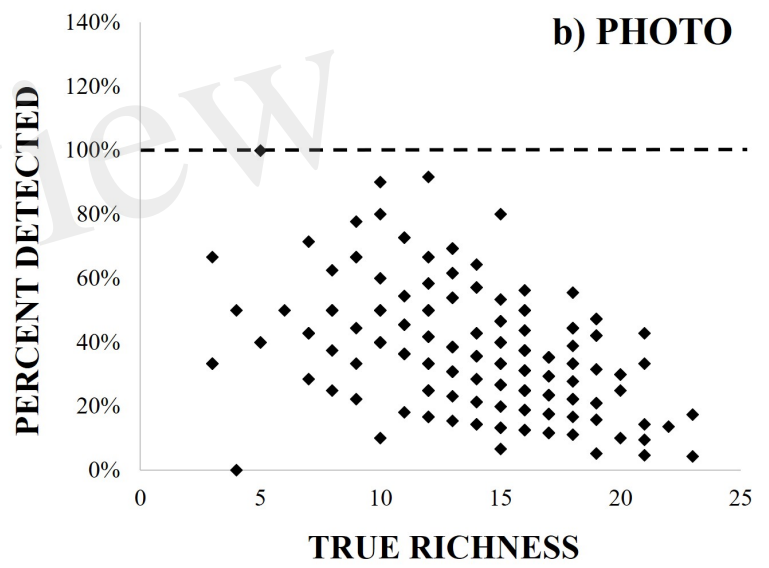
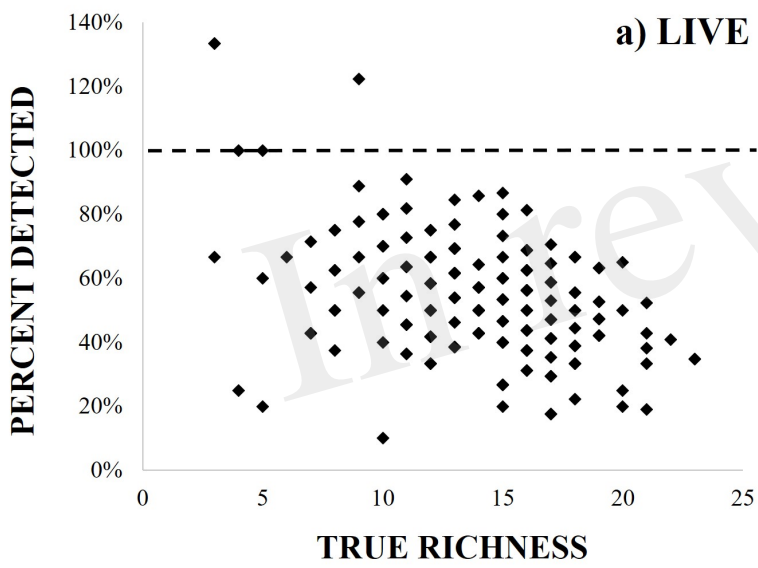


Figure 3.JPEG

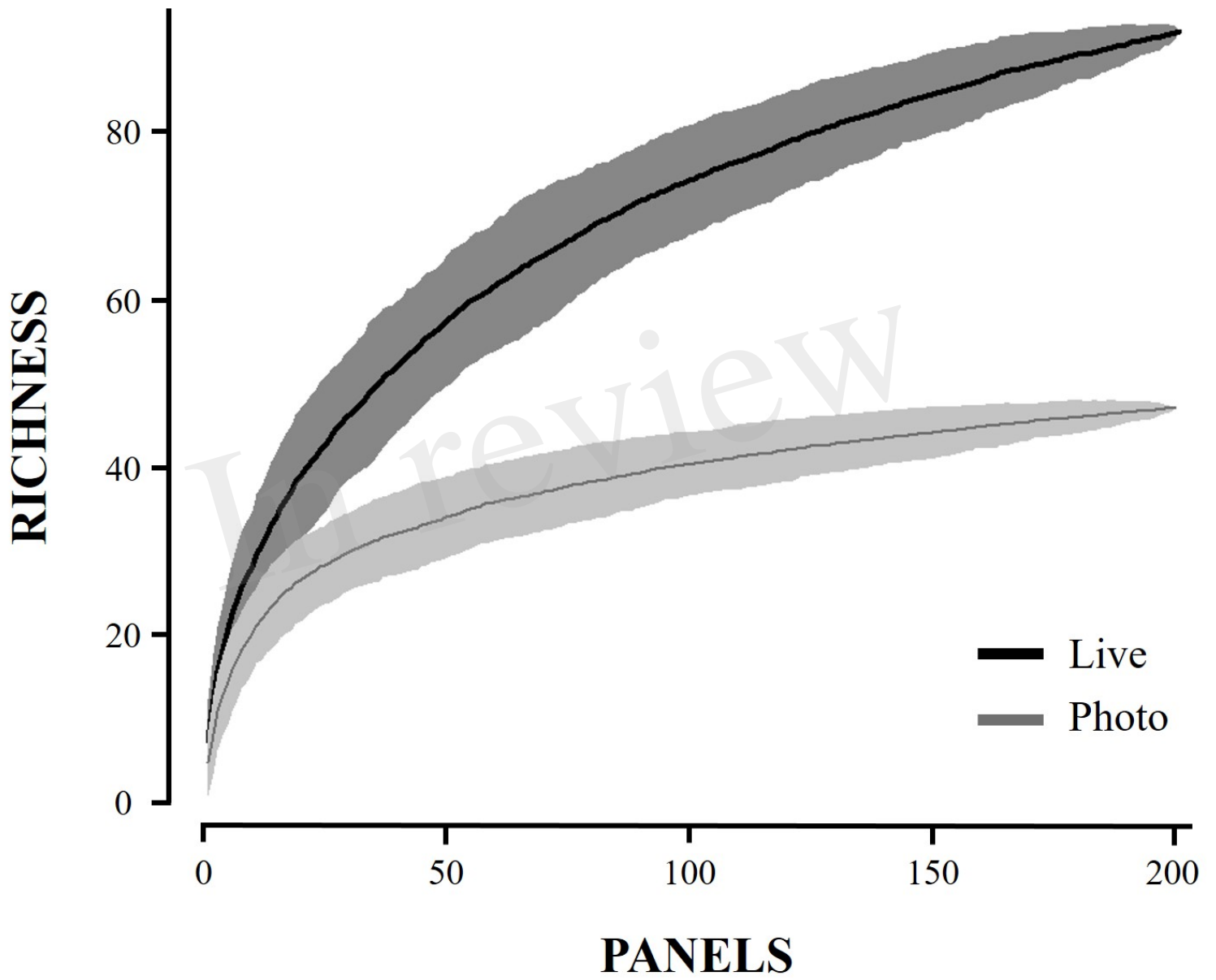


Figure 4.JPEG

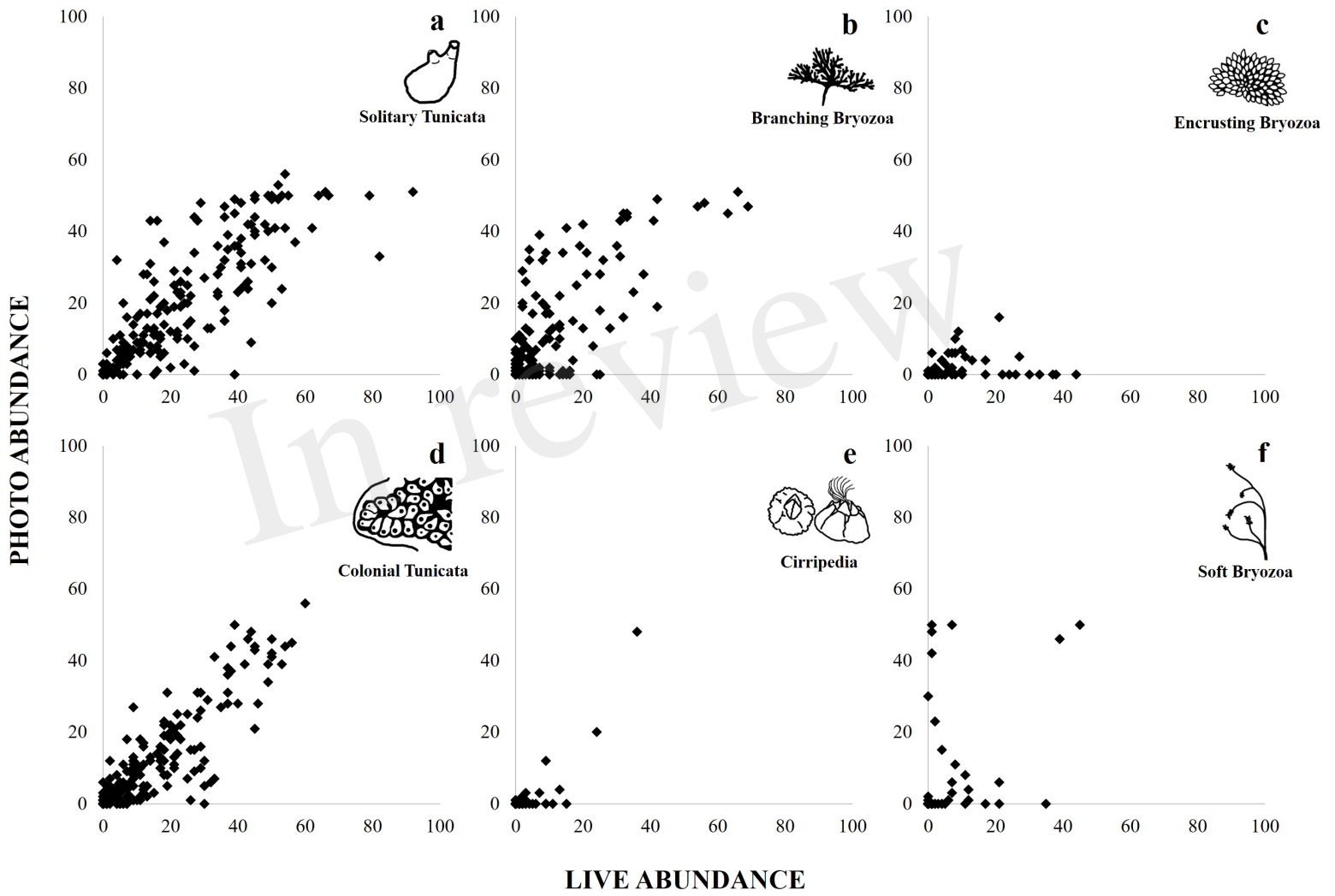


Figure 5.JPEG

