

1    **Fishing degrades size structure of coral reef fish communities**

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3    Running head: *Fishing alters reef fish size structure*

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47 **Abstract**

48 Fishing pressure on coral reef ecosystems has been frequently linked to reductions of  
49 large fishes and reef fish biomass. Associated impacts on overall community structure  
50 are, however, less clear. In size-structured aquatic ecosystems, fishing impacts are  
51 commonly quantified using size spectra, which describe the distribution of individual  
52 body sizes within a community. We examined the size spectra of coral reef fish  
53 communities at 38 US-affiliated Pacific islands, spanning from near pristine to highly  
54 human populated. Reef fish community size spectra slopes 'steepened' steadily with  
55 increasing human population and proximity to market due to a reduction in the relative  
56 biomass of large fishes and an increase in the dominance of small fishes. In contrast, total  
57 fish community biomass was substantially lower on inhabited islands than uninhabited  
58 ones, regardless of human population density. Comparing the relationship between size  
59 spectra and reef fish biomass, we found that on populated islands size spectra steepened  
60 linearly with declining biomass, whereas on uninhabited islands size spectra and biomass  
61 were unrelated. Size spectra slopes also were steeper in regions of low sea surface  
62 temperature but were insensitive to variation in other environmental and geomorphic  
63 covariates. In contrast, reef fish biomass was highly sensitive to biophysical conditions,  
64 being influenced by oceanic productivity, sea surface temperature, island type, and  
65 habitat complexity. Our results suggest that community size structure is more robust than  
66 total fish biomass to increasing human presence and that size spectra are reliable  
67 indicators of exploitation impacts across regions of different fish community  
68 compositions, environmental drivers, and fisheries types. Size-based approaches that link  
69 directly to functional properties of fish communities, and are relatively insensitive to

70 abiotic variation across biogeographic regions, offer great potential for developing our  
71 understanding of fishing impacts in coral reef ecosystems.

72

73 **Introduction**

74 Overexploitation of marine species can cause system-wide shifts in species  
75 abundances and interactions (Bascompte et al., 2005; Britten et al., 2014), which in turn  
76 alter the structure and function of marine ecosystems (Jackson et al., 2001; Travis et al.,  
77 2014). Selective fishing of large consumers can produce trophic cascades (Bascompte et  
78 al., 2005; Baum & Worm, 2009) and destabilize predator-prey dynamics (Britten et al.,  
79 2014), while sustained exploitation at lower trophic levels can collapse prey populations  
80 (Essington et al., 2015). In temperate systems, broad fishing impacts are often evaluated  
81 using complex ecosystem-based models that require high-resolution ecological and  
82 exploitation data (Thorpe et al., 2015). However, when ecosystems are characterized by  
83 high ecological diversity or limited catch data these approaches are infeasible. Instead,  
84 community-level indicators that are simple to estimate, grounded in ecological theory,  
85 and generalizable across ecosystems can provide informative assessments of fishing  
86 impacts (Rochet & Trenkel, 2003; Thrush & Dayton 2010). Gaining such insights is of  
87 paramount importance for subsistence coral reef fisheries, which provide important  
88 sources of protein and livelihoods to millions of people across the world's tropical island  
89 nations (Sadovy, 2005; Newton et al., 2007). Coral reef fish assemblages are highly  
90 diverse (Kulbicki et al., 2013) and their fisheries are multi-species and multi-gear (Hicks  
91 & McClanahan, 2012), but catch and effort data are typically limited (Sadovy, 2005;  
92 Zeller et al., 2015). As a result, exploitation impacts can be particularly difficult to  
93 quantify (McClanahan et al., 2015; Nash & Graham, 2016), underscoring the need for  
94 simple community-level indicators of exploitation impacts.

95

96 In aquatic systems, trophic interactions are size-based and body size and  
97 individual trophic level are tightly linked (Jennings et al., 2001; Barnes et al., 2010).  
98 Size-based approaches that generalize across species but preserve links to community-  
99 level traits may provide significant insights into the impacts of exploitation in complex  
100 systems such as coral reefs (Nash & Graham, 2016). Body size also scales predictably  
101 with a number of important ecological processes, from metabolic rate at the individual  
102 scale (West et al., 2001) to biomass turnover at the population scale (Brown et al., 2004).  
103 Therefore, size-based approaches offer powerful methods of assessing ecological  
104 structure across distinct communities, and link directly to functional traits that are  
105 otherwise difficult to estimate in data-poor systems (Taylor et al., 2014). One metric, the  
106 size spectrum, describes the distribution of individuals across body sizes irrespective of  
107 species (White et al., 2007; Trebilco et al., 2013). The size spectrum has been used to  
108 assess fishing impacts across a range of temperate marine (Blanchard et al., 2005; Daan  
109 et al., 2005; Sweeting et al., 2009) and freshwater fish communities (Sprules, 2008),  
110 where community size structure is represented by the slope of the relationship between  
111 abundance and body size on logarithmic scales (White et al., 2007). Size-selective fishing  
112 causes the spectrum slope to decrease or ‘steepen’ as large fishes are depleted and prey  
113 species are released from predation (Daan et al., 2005; Shin et al., 2005; Fung et al.,  
114 2013). Metabolic and size-based theory predicts that a reduction in large fishes will  
115 produce shifts in size-linked life history traits such that overexploited communities are  
116 characterized by a greater dominance of small individuals, and concomitant higher  
117 productivity and faster biomass turnover times (Jennings & Blanchard, 2004; McCann et  
118 al., 2016).

119 In small-scale, artisanal coral reef fisheries, overexploitation is a pervasive issue  
120 that threatens the sustainability of a vital food resource for developing coastal countries  
121 (Newton et al., 2007; Cinner et al., 2009; Johnson et al., 2013). Standing stock biomass is  
122 widely used as metric of fishery health and of exploitation impacts at regional scales  
123 (Cinner et al., 2009; Cinner et al., 2012a; MacNeil et al., 2015) and, although declines in  
124 the abundance of large fishes on coral reefs are well documented (Sandin et al., 2008;  
125 Williams et al., 2011), analyses of associated impacts on coral reef fish community size  
126 structure have been infrequent (Nash & Graham, 2016). Steepening of size spectra slopes  
127 due to overfishing of large fishes has thus far been detected only in Fijian small-scale reef  
128 fisheries, and across only moderate gradients in exploitation pressure (Dulvy et al., 2004;  
129 Graham et al., 2005; Wilson et al., 2010). Other direct comparisons between fished and  
130 protected areas have found that community size structure is highly variable and unrelated  
131 to exploitation, which may reflect unmeasured environmental influences (McClanahan &  
132 Graham, 2005; Graham et al., 2007). As a result, it remains unclear whether degradation  
133 in overall community size structure occurs across extreme gradients in exploitation  
134 pressure, such as from pristine to overexploited reef communities, and if these patterns  
135 are dependent on the fisheries' species composition. At regional and global scales, recent  
136 macroecological analyses of coral reef fish trophic structure and life history traits indicate  
137 that biomass and ecological functions may be broadly preserved in lightly exploited  
138 communities (McClanahan et al., 2011; MacNeil et al., 2015; McClanahan et al., 2015).  
139 Similar examination of reef fish community size structure across large spatial scales and  
140 gradients in fished biomass would provide additional insights into the state of coral reef  
141 fisheries relative to unexploited ecosystems.

142 Here, we use a large-scale dataset of Pacific reef fish abundances spanning from  
143 remote near-pristine islands and atolls to highly populated ones, to examine how human  
144 impacts alter the size structure of reef fish communities. The reefs included in our  
145 analyses also span strong gradients in environmental covariates (Williams et al., 2015),  
146 and differ substantially in their species compositions (Kulbicki et al., 2013) and  
147 exploitation history (Dalzell et al., 1996; Houk et al., 2012). We estimated size spectrum  
148 slopes to assess shifts in community structure across a body size range from tiny  
149 planktivores (20g) to large piscivores (> 1kg), and quantified the biomass of large fishes  
150 relative to the total fish community to determine whether exploitation was size selective.  
151 To examine how changes in size structure corresponded with more conventional  
152 biomass-based indicators, we also compared trends in size spectra with trends in total  
153 community biomass.

154

## 155 **Materials and Methods**

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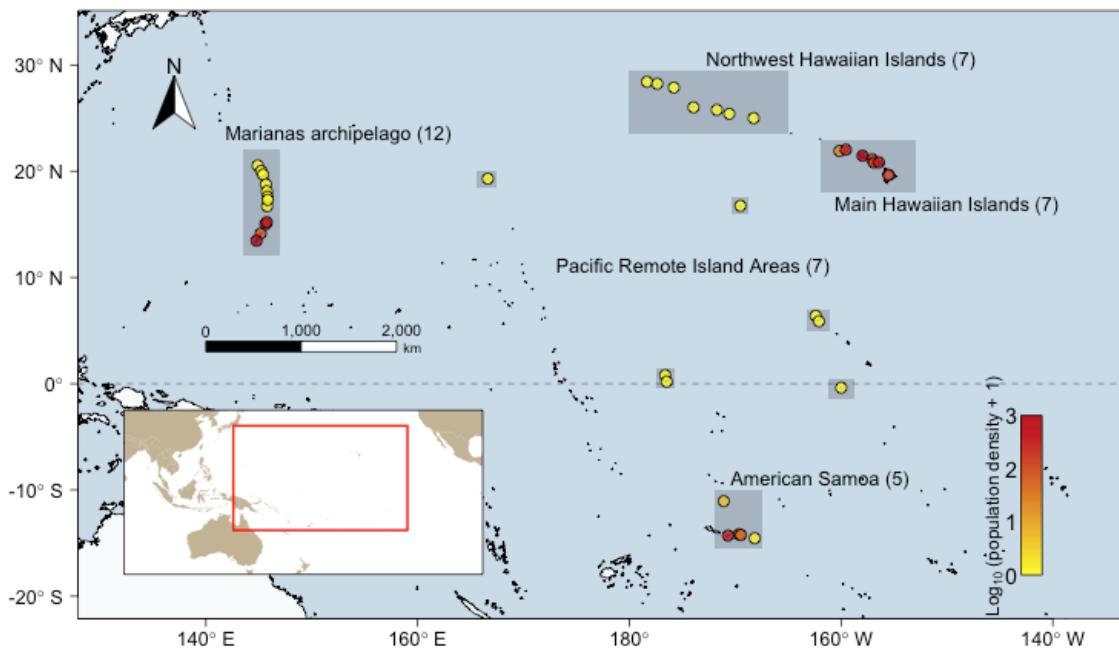
### 157 *Study location and survey data*

158 We examined reef fish communities at 2,124 sites located on 38 U.S.-affiliated Pacific  
159 islands, atolls, and banks (hereafter islands) (Fig. 1), that were surveyed between 2010  
160 and 2014 by the Pacific Reef Assessment and Monitoring Program (Pacific RAMP) of  
161 NOAA's Coral Reef Ecosystem Program (CREP). Surveyed islands encompass  
162 substantial gradients in biodiversity, productivity and temperature, and span human  
163 population densities from uninhabited atolls to densely populated islands supporting up to  
164 2,235 people/km<sup>2</sup> forereef habitat (Table S1) (Williams et al., 2015).

165 The survey data (Coral Reef Ecosystem Program; Pacific Islands Fisheries  
166 Science Center 2015) consist of observations of individual fish made during underwater  
167 visual censuses (UVCs) by CREP's highly trained scientific divers. Two divers  
168 conducted stationary point counts (SPC), with each surveying one of two adjacent  
169 visually estimated 15 m diameter cylinders along a 30 m transect (survey area = 353 m<sup>2</sup>).  
170 Each diver identified every fish species present in or transient through their cylinder,  
171 before enumerating and sizing (total length to the nearest cm) all observed fishes (Ayotte  
172 et al., 2011). CREP surveys were stratified by depth bin, into shallow (0-6 m), mid (6-  
173 18 m) and deep (18-30 m) zones, and we only examined surveys conducted on forereef  
174 habitat. The number of surveys at each island was proportional to the total forereef area.  
175

176 We considered each individual UVC survey recorded by a pair of divers (two  
177 CREP cylinders) as a unique site. To analyse fishing impacts at the community level we  
178 aggregated all sites sampled in each year across each island (n = 70 island x year  
179 combinations). We converted the length estimate from each individual fish to body mass  
180 (to the nearest gram) using published length-weight relationships for species or families  
181 (Kulbicki et al., 2005; Froese & Pauly, 2016). Because UVC methods of coral reef fish  
182 communities can be subject to several potential biases (Bozec et al., 2011), we excluded  
183 all fish < 20 g body mass to avoid underestimating the abundance of small cryptic fishes  
184 (Ackerman & Bellwood, 2000; Wilson et al., 2010). In addition, large mobile piscivores  
185 (i.e. sharks and jacks) are often overestimated in small-scale non-instantaneous  
186 underwater visual surveys (Ward-Paige et al., 2010), and may also be attracted to divers  
187 at remote islands (Parrish et al., 2000; Richards et al., 2011). Both biases can

188 substantially inflate biomass estimates and we therefore followed other recent large-scale  
189 studies of reef fish communities by excluding sharks and jacks from our analyses  
190 (MacNeil et al., 2015; Williams et al., 2015).



191  
192 Fig. 1. Map of Pacific islands surveyed by CREP ( $n = 38$ ) with each coloured by human  
193 population density. Human population density is population per forereef area ( $\text{km}^2$ )  
194 within a 20 km radius on a  $\log_{10}$  scale.

195

#### 196 *Reeffish community analyses*

197 We used size spectra to quantify reef fish community structure. The size spectrum  
198 is usually fitted to frequencies of body sizes and predicted to approximate a power law  
199 distribution (Eq. 1) (Vidondo et al., 1997; Andersen & Beyer, 2006). Here, we used  
200 maximum likelihood estimation to estimate the size spectrum slope,  $b$  (Vidondo et al.,

201 1997; Edwards, 2008). We fitted body size data for individual fishes from each island, for  
202 each year, to a bounded power law distribution with probability density function:

203 
$$f(x) = \frac{(b+1)x^b}{x_{\max}^{b+1} - x_{\min}^{b+1}} \quad (1)$$

204 where  $x$  is body mass,  $b$  is the scaling exponent, and the distribution is bounded by the  
205 minimum and maximum possible body sizes ( $x_{\min}, x_{\max}$ ) (White et al., 2008). Equation 1  
206 is undefined for  $b = -1$ , but this value does not occur for our data. The log-likelihood of a  
207 bounded power law is:

208 
$$\log[L(b \mid \text{data})] = n \log\left(\frac{b+1}{x_{\max}^{b+1} - x_{\min}^{b+1}}\right) + b \sum_{j=1}^n \log x_j \quad (2)$$

209 (Edwards et al., in revision) and was numerically optimized to estimate  $b$  (Edwards,  
210 2008; Edwards et al., 2012). Unlike binning-based approaches to fitting frequency data,  
211 this method has the benefit of producing accurate estimates of  $b$  (Edwards et al., in  
212 revision). In our maximum likelihood estimation,  $x_{\min}$  and  $x_{\max}$  are the minimum (i.e. 20  
213 g) and maximum observed values at each island within a single survey year (Edwards et  
214 al., 2012). In most empirical analyses of the aquatic size spectrum, binning-based  
215 methods are used to estimate  $b$ , such that the regression slope is the parameter of interest  
216 and a ‘steepening spectrum’ is predicted following the selective exploitation of large  
217 body sizes (i.e. the regression slope, or  $b$ , becomes more negative as the abundance of the  
218 largest size classes is depleted relative to small size classes) (Daan et al., 2005; Blanchard  
219 et al., 2005; Graham et al., 2005; Petchey & Belgrano, 2010). For consistency with these  
220 studies, we refer to the power law exponent  $b$  as the size spectrum slope (see also  
221 Edwards et al., in revision).

222

223 We used a Monte Carlo resampling procedure to correct for differences in  
224 sampling effort (i.e. number of UVCs) at each island. Size spectrum slopes were  
225 estimated for a random sample (without replacement) of 1000 individual fish at each  
226 island in each survey year, and the size spectrum slope was the mean slope estimate from  
227 10,000 replicate random samples. Each island included in the analysis had at least 1000  
228 individual fish observations (Table S2).

229

230 In addition to size spectra, we examined two biomass–based fisheries indicators.  
231 First, we quantified overall community fish biomass ( $\text{kg ha}^{-1}$ ) by averaging biomass  
232 across all UVCs at each island for each year. Second, to investigate the extent to which  
233 size-selective fishing was responsible for the observed patterns in size spectra slopes and  
234 overall community biomass, we estimated the proportion of large fish at each island using  
235 a large fish indicator (LFI) (Greenstreet et al., 2011). We defined the LFI as the biomass  
236 of fish  $> 1\text{kg}$  divided by the total biomass of the fish community, averaged across all  
237 UVCs at each island for each year.

238

239 *Explanatory covariates*

240 We examined variation in community size spectra and fish biomass in relation to  
241 two anthropogenic and six environmental covariates (Tables 1, S1). No standard measure  
242 of fishing effort was available across all islands sampled. Instead, we estimated both  
243 human population density, expressed as number of people within a 20 km radius divided  
244 by the forereef area (Williams et al., 2015), and distance to market (defined as the  
245 distance to provincial capital) (Cinner et al., 2012a) as distal metrics of exploitation

246 pressure on coral reef fish communities (Appendix S1). Although human population  
247 density is often strongly correlated with a loss of reef fish biomass (Mora, 2008;  
248 Williams et al., 2011; Cinner et al., 2012a; Williams et al., 2015), distance to market,  
249 which is less commonly employed, may be a more sensitive indicator of fishing intensity  
250 on sparsely populated coral reefs (Brewer et al., 2012; Cinner et al., 2012a; Maire et al.,  
251 2016). Sea surface temperature (SST) and oceanic productivity also can both positively  
252 influence reef fish biomass (Williams et al., 2015), but their influence on community size  
253 structure remains unclear. We used remote sensing data to calculate time-averaged  
254 estimates of SST ( $^{\circ}\text{C}$ ) and oceanic productivity ( $\text{mg C m}^{-2} \text{ day}^{-1}$ ) at each site (Appendix  
255 S1). In addition to oceanographic factors, coral reef fish communities may be influenced  
256 by a suite of other biophysical characteristics (Table 1). For example, reef area and island  
257 type have been shown to influence reef fish biomass (Cinner et al., 2012a) while, at the  
258 site level, reefs of high complexity are thought to offer extensive prey refugia that  
259 support greater densities of small-bodied fish and steeper size spectra (Wilson et al.,  
260 2010; Alvarez-Filip et al., 2011; Rogers et al., 2014). We estimated land area and reef  
261 area within 75 km radius of each site (Appendix S1), classified each island as an atoll  
262 (e.g. Kure, Palmyra), island with lagoon or pseudo-lagoon ('low' island, e.g. Saipan), or  
263 island without a lagoon ('high' island, e.g. Oahu) following D'Agata et al. (2014), and  
264 quantified habitat complexity with both *in situ* (habitat complexity) and remotely sensed  
265 (bathymetric slope) estimates at each site (Appendix S1). All site-level explanatory  
266 covariates were averaged to give estimates for each island (Table S1).

Covariate	Definition	Source	Size spectrum			Fish biomass
			-ve	+ve	-ve	
Human population density	Total population within a 20 km radius per reef area (km <sup>2</sup> )	SEDAC	1, 2, 3	-	6, 7, 8, 9, 10, 11, 12	
Proximity to market	Distance to nearest provincial capital (km)	ARC GIS	-	-	10, 11	
Minimum SST	Mean of weekly minimum SST (°C) values over 1982-2009 at 4x4 km resolution	CoRTAD	-	12	13	
Mean productivity	Weekly mean of productivity (mg C m <sup>-2</sup> day <sup>-1</sup> ) values over 2002-2013 for at least 3 1x1 km cells	NOAA CoastWatch	-	12, 13	-	
Habitat complexity	Mean substrate height within point count cylinder	CREP				
Bathymetric slope	Bathymetric slope extent (0 – 90°) at 1x1 km resolution	MARSPEC	3, 4, 5	12, 14, 15	-	
Island type	Atoll, low (island with lagoon or pseudo-lagoon), high (island without lagoon)	D'Agata et al. (2014)	-		Highest at atolls (11)	
Land area	Land area within 75 km radius (km <sup>2</sup> )	Millennium/Coral Reef Habitat Map	-	-	-	
Reef area	Total reef area <30 m depth within 75 km radius (km <sup>2</sup> )	Millennium/Coral Reef Habitat Map	-		No effect (11)	

268 Source references and methodological details in Appendix S1. Example references: 1. Dulvy et al. (2004); 2. Graham et al. (2005); 3. Wilson et al.  
 269 (2010); 4. Alvarez-Filip et al. (2011); 5. Rogers et al. (2014); 6. Jennings et al. (1995); 7. Jennings & Polunin (1997); 8. Mora et al. (2011); 9.  
 270 Williams et al. (2011); 10. Brewer et al. (2012); 11. Cinner et al. (2012a); 12. Williams et al. (2015); 13. Barneche et al. (2014); 14. Friedlander et  
 271 al. (2003); 15. Graham & Nash (2012)

272 Table 1. Anthropogenic and environmental covariates included in size spectra and  
273 biomass models. Previous studies that examined the influence of each covariate on  
274 size spectra and biomass are numbered and categorized by the direction of the  
275 relationship they observed (positive, +ve; negative, -ve).

276

277 *Statistical modeling*

278 Prior to analyses, we applied  $\log_{10}$  transformations to distance to market (km),  
279 population density per island ( $\log_{10}(\text{density} + 1)$  per  $\text{km}^2$ ), and reef area ( $\text{km}^2$ ) to  
280 reduce skewness. We also centered and standardized all continuous covariates  
281 (Schielzeth, 2010). Island type (atoll, low island, high island) was coded as two  
282 dummy variables before centering to a mean of zero. Distance to market and  
283 population density were strongly negatively correlated ( $r = -0.84$ ), so to avoid  
284 collinearity issues we fitted separate models for each human covariate.

285

286 We modelled size spectra slopes and reef fish biomass estimates against the  
287 eight anthropogenic and environmental covariates at the island level. The distribution  
288 of size spectra estimates  $b$  was normal (Shapiro-Wilk normality test:  $W = 0.992$ ;  $p =$   
289 0.934) so we used linear mixed effects models (*lme4* package in R; Bates et al., 2015)  
290 to examine variation amongst them. To account for instances of islands sampled in  
291 multiple years, survey year ( $j$ ) was included as a random effect ( $\rho_j$ ). We modeled reef  
292 fish biomass with a gamma distribution and a log link (Zuur et al., 2009), and the  
293 same fixed and random effects structure as the size spectra models. Prior to model  
294 selection procedures, we assessed evidence of collinearity with variance inflation  
295 factors (VIF), where variables with  $\text{VIF} > 10$  were considered evidence of strong

296 multicollinearity (Zuur et al., 2009). In the saturated size spectrum and reef fish

297 biomass models every explanatory variable had a VIF <6.

298

299 We used multimodel inference to examine models based on all possible  
300 subsets of our anthropogenic and environmental covariates using a dredge function  
301 (MuMIn package in R; Barton, 2015). We assessed model support with the Akaike  
302 Information Criterion adjusted for small sample sizes (AIC<sub>c</sub>) (Burnham & Anderson,  
303 2002) and found there was no single top model (i.e.  $\Delta\text{AIC}_c > 2$ ). Instead, following  
304 Cade (2015) we examined weighted absolute t-statistic values across all subset  
305 models as a measure of covariate importance. The t-statistic can be used as a measure  
306 of effect size within models as it is the parameter estimate divided by the standard  
307 error. We weighted each t-statistic by the corresponding model probability (i.e. AIC<sub>c</sub>  
308 weight for each model  $i$ ,  $w_i$ ), and estimated the weighted sample variance ( $\sigma^2$ ) for  
309 each t-value ( $x_i$ ) for the weighted mean t-value ( $\mu$ ):

$$310 \sigma_{\text{weighted}}^2 = \sum_{i=1}^N w_i (x_i - \mu)^2 \quad [3]$$

311 In this way, the variables that were most important in predicting the given response  
312 (i.e., had the strongest effects in the more probable models) had the largest weighted  
313 absolute t-statistic.

314

315 To visualize how the most important explanatory covariates influenced size  
316 spectra and reef fish biomass, we examined model predictions for each explanatory  
317 covariate across the range of observed values while holding all other predictor  
318 covariates at their means. We plotted the model-averaged prediction across the top  
319 model set ( $\Delta\text{AIC}_c < 7$ ) weighted by the corresponding model probabilities (Burnham  
320 & Anderson, 2002), and estimated the weighted sample variance as a measure of

321 variability in predictions across the top model set. We visualized the predictions  
322 concerning distance to market models in the same direction as human population  
323 density by plotting predictions against the inverse of distance to market (hereafter  
324 'proximity to market', i.e. for the scaled covariates, islands with high population  
325 estimates also had high proximity to market estimates).

326

327 We also examined if changes in size spectra corresponded with changes in  
328 reef fish biomass, and if those relationships differed between populated and  
329 uninhabited islands. We fitted linear mixed models to examine how size spectra  
330 changed across a gradient of reef fish biomass, treating populated and uninhabited  
331 islands separately and including survey year as a random effect. To explicitly test for  
332 size-selective fishing of large body sizes, we used the same approach to examine the  
333 relationship between size spectra and the LFI at populated and uninhabited islands  
334 (Fig. S1).

335

336 Finally, we conducted sensitivity analyses to test the robustness of our results  
337 to different treatments of the datasets. UVC methods provide estimates of length  
338 rather than mass, and previous studies of reef fish communities have generally fitted  
339 length spectra (Dulvy et al., 2004; Graham et al., 2005; Wilson et al., 2010). As such,  
340 we also estimated size spectra slopes using reef fish lengths and refitted our statistical  
341 models. Model averaged predictions and weighted mean t-statistic ratios for reef fish  
342 length spectra models were similar to results from mass spectra models (Figs. S2, S3,  
343 Table S5). Estimates of mass spectra facilitate comparisons with our analyses of reef  
344 fish biomass and, as a result, we decided to present mass spectra rather than length  
345 spectra as our main results.

346

347 All analyses were conducted using R version 3.2.0 (R Core Team, 2015), and  
348 we provide our code at an open source repository (github.com/baumlab/robinson-  
349 reefs-spectra).

350

## 351 **Results**

### 352 *Size spectra analyses*

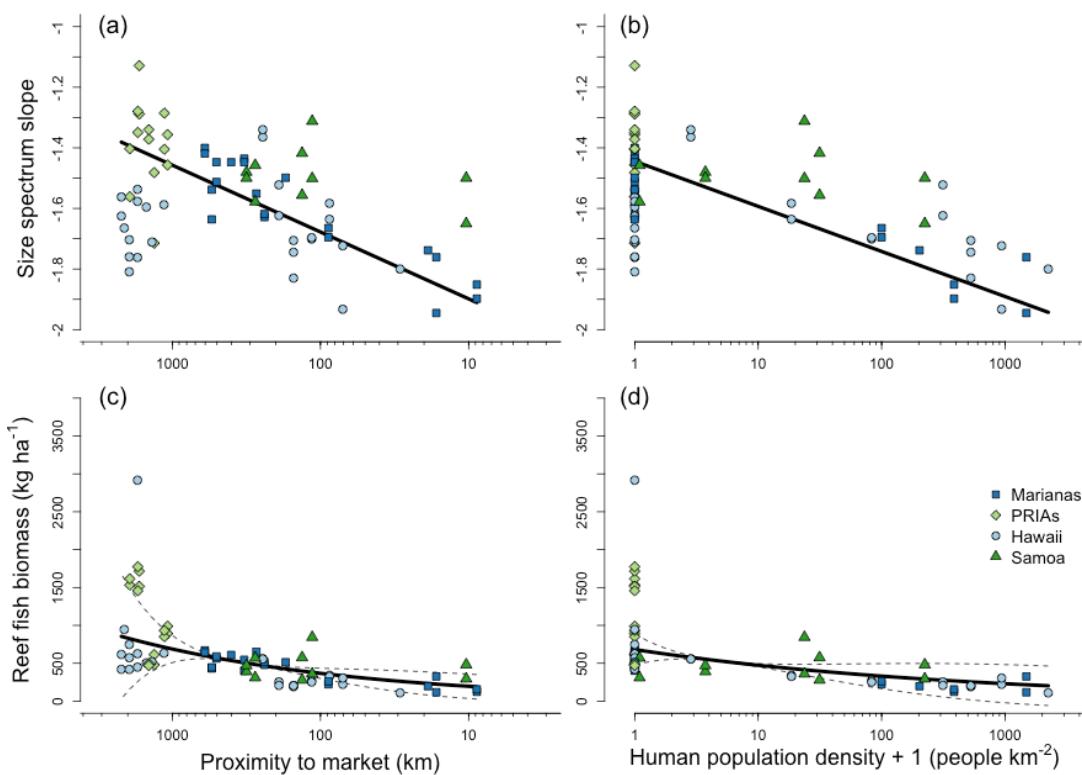
353 Reef fish community size structure varied considerably across the gradient of  
354 human impacts over 38 Pacific islands, with size spectra slopes ( $b$ ) ranging from -1.13  
355 down to -1.95 (Fig. 2a,b). The human disturbance and several environmental variables  
356 explained a large proportion of the variation in size spectra across islands. Across the  
357 top model set (all models  $< 7 \Delta AIC_c$ ), the range in  $R^2$  was 0.56 – 0.59 and 0.61 – 0.65  
358 for the proximity to market and human population density models, respectively (Table  
359 S3). Regardless of the metric used, human disturbance had the strongest effect on the  
360 size spectrum at a given island (Fig. 3a). Size spectra slopes decreased linearly with  
361 increasing proximity to market (model averaged t-statistic = 6.87) (Fig. 2a) and with  
362 increasing human population density (model averaged t-statistic = 7.85) (Fig. 2b).  
363 The steepest size spectra ( $b < -1.8$ ) were generally observed only at reefs with high  
364 human population density, which typically also were close to market centres (Pearson  
365 correlation = 0.84) (Fig. 2a,b). Apart from human impact covariates, minimum SST  
366 ( $^{\circ}C$ ) had a strong positive effect on size spectra slopes in top model sets for both  
367 proximity to market (model averaged t-statistic = 6.23) and population density (4.93).  
368 The remaining environmental and biogeographic covariates had relatively weak  
369 effects on size spectra (all model averaged t-statistics  $< 2.4$ ) (Fig. 3a).

370

371 *Biomass analyses*

372 Reef fish biomass varied across islands from an estimated  $110 \text{ kg ha}^{-1}$  to over  
373  $2900 \text{ kg ha}^{-1}$ , and was lowest at islands with high human presence. Across the top  
374 model set,  $R^2$  ranged from  $0.54 - 0.59$  and  $0.53 - 0.58$  for the proximity to market and  
375 human population density models, respectively (Table S4). As with the size spectra  
376 models, human disturbance covariates were the strongest drivers of reef fish biomass  
377 (Fig 3b): reef fish biomass decreased non-linearly with increasing proximity to market  
378 (Fig. 2c) and human population density (Fig. 2d), and only the remote, unpopulated  
379 islands supported biomass levels  $>1000 \text{ kg ha}^{-1}$ . The lowest biomass levels ( $<200 \text{ kg}$   
380  $\text{ha}^{-1}$ ) were observed only at reefs with high human population density, which typically  
381 were also close to market centres (Fig. 2c,d). Several environmental covariates were  
382 also important drivers of reef fish biomass. Generally, islands with higher minimum  
383 SST ( $^{\circ}\text{C}$ ) and higher productivity supported greater biomass (Fig. 3b). However, the  
384 relative effects of SST and productivity on biomass differed slightly between model  
385 sets, with SST the stronger driver in the proximity to market model set (model  
386 averaged t-statistic = 4.33 for proximity to market; 3.40 for human population  
387 density) and productivity the stronger in the human population density model set  
388 (3.16 and 5.11). For both model sets, atolls supported greater biomass than high (t-  
389 statistics: 1.53 and 2.56 for proximity to market and human population density,  
390 respectively) and low islands (0.41 and 1.08), and more complex habitats were  
391 associated with higher biomass levels (2.32 and 1.23) (Fig. 3b).

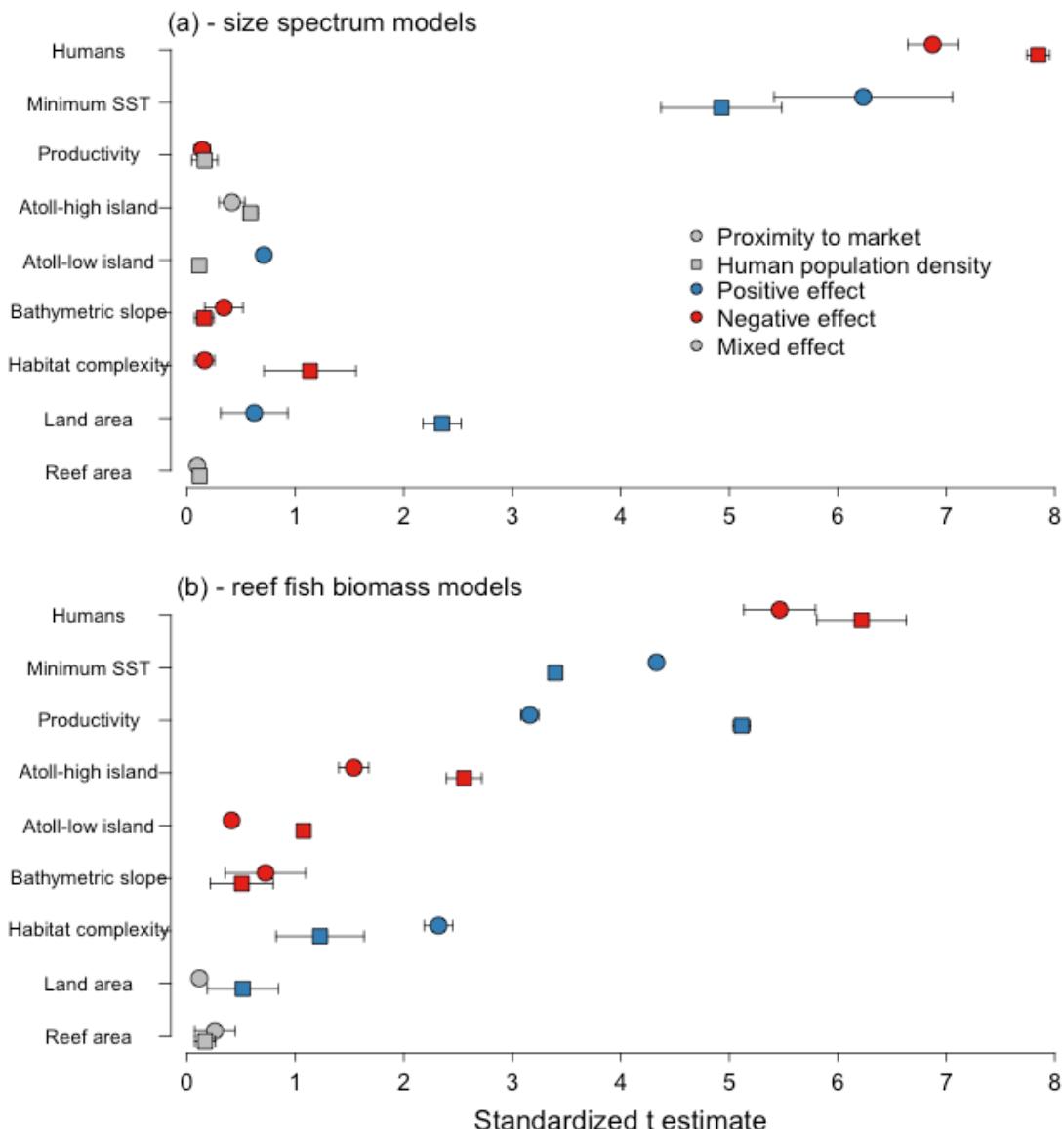
392



393

394 Fig. 2. Human drivers of coral reef fish community size structure and biomass ( $\text{kg ha}^{-1}$ )<sup>1</sup>. Size spectra (a) and reef fish biomass (b) relationships are model averaged predictions across the standardized range of observed  $\log_{10}$  proximity to provincial capital (km) and  $\log_{10}$  human population density per forereef area ( $\text{km}^2$ ) (b, d respectively). Predictions were made across the top model set ( $\Delta\text{AICc} < 7$ ) and weighted using model probabilities (Tables S3, S4), while holding all other relevant covariates to their mean observed value. Dashed lines are the weighted sample variance at each value of human covariate (though these are indistinguishable from the model predictions in the size spectra analyses). For visualization purposes, we included the observed data as points plotted against the untransformed human covariates and coloured by region (dark blue squares = Marianas archipelago; light blue circles = Hawaiian archipelago, light green diamonds = Pacific Remote Island Areas, dark green triangles = American Samoa).

407



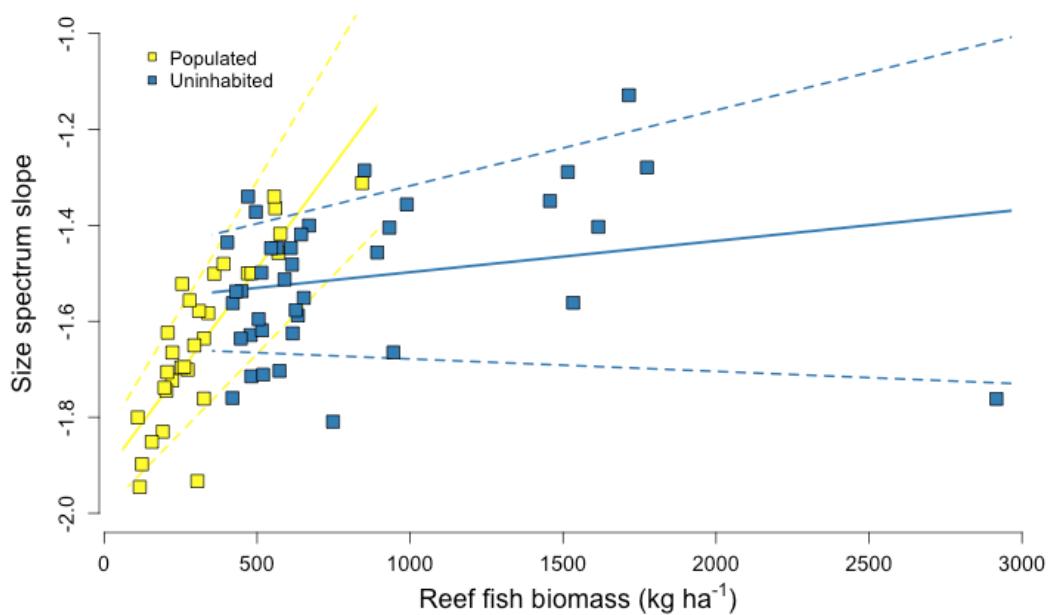
408

409 Fig. 3. Human and environmental drivers of reef fish size structure and biomass. Size  
 410 spectra (a) and reef fish biomass (b) are presented for the distance to market (circles)  
 411 and human population density (squares) full model sets. Points are the weighted  
 412 absolute t-values for each explanatory covariate, with weighted sample variance as  
 413 error bars. T-values indicate the magnitude of each covariate effect, and colors  
 414 indicate the direction of each covariate effect (blue = positive; red = negative; grey =  
 415 mixed). See Tables S3 and S4 for further details.

416

417 *Populated vs. uninhabited reef fish community structure*

418 At the populated islands, there was a strong relationship between size spectra  
419 and reef fish biomass ( $P < 0.001$ ,  $R^2 = 0.70$ ) (Fig. 4). This relationship appeared to be  
420 explained by the disproportionate exploitation of large-bodied fishes, since the most  
421 negative (i.e. steepest) spectra slopes were associated with particularly low values for  
422 the large fish indicator (i.e. low relative biomass of large-bodied fish;  $P < 0.001$ ,  $R^2 =$   
423 0.30) (Fig. S1). In contrast, despite substantial variation in size spectra slopes (-1.81  
424 to -1.13) and reef fish biomass (402 to 2917 kg ha<sup>-1</sup>), size spectra did not consistently  
425 covary with either fish biomass ( $P = 0.172$ ,  $R^2 = 0.05$ ) (Fig. 4) or the large fish  
426 indicator ( $P = 0.316$ ,  $R^2 = 0.03$ ) at the remote, uninhabited islands (Fig. S1).



427  
428 Fig. 4. Change in size spectra across the gradient of reef fish biomass. Size spectra ~  
429 biomass relationships were fitted separately to uninhabited (blue) and populated  
430 (yellow) islands. Solid lines are linear regression slope estimates with 95%  
431 confidence intervals. Each point is a size spectrum slope and biomass estimate at one  
432 island in a single survey year.

433

434 **Discussion**

435 Our analyses reveal that, along a disturbance gradient from reefs of near-pristine  
436 wilderness to degraded reefs at developed population centres, increasing human  
437 presence causes a gradual degradation of coral reef fish community size structure. At  
438 populated islands, steeper size spectrum slopes were associated with a reduction in  
439 total fish biomass and the relative biomass of large-bodied fishes. The specificity of  
440 each ecological indicator to human impacts was markedly different, such that size  
441 spectra responded to solely to human presence and sea surface temperature whereas  
442 total biomass was highly sensitive to low levels of human presence as well as  
443 influences of temperature, oceanic productivity, and island geomorphology.

444

445 At populated islands, steepening size spectra represent a gradual shift in body  
446 size distributions from fish communities with a high relative proportion of large fish  
447 (shallow slopes) to ones dominated by small fishes (steep slopes). Large-bodied fishes  
448 play important roles in maintaining reef functions, suggesting that the loss of these  
449 individuals due to size selective exploitation may have disproportionate functional  
450 impacts on coral reefs. For example, many large herbivorous fishes are important  
451 bioeroders and control algal growth (Bellwood et al., 2011; Edwards et al., 2013).  
452 More generally, large predators can control the stability of prey populations across  
453 habitats (Rooney et al., 2006; Britten et al., 2014). Size-selective exploitation of these  
454 fishes may therefore impair the ability of reefs to recover from additional disturbances  
455 such as coral bleaching and hurricane damage (Cheal et al., 2013). Size spectra  
456 analyses of moderately exploited reef fisheries in Fiji (Dulvy et al., 2004: 1-100  
457 people per km reef front; Graham et al., 2005: 3-300 people per km reef front)  
458 previously suggested that harvesting of large-bodied fishes steepens size spectra at

459 small spatial scales. Fishing practices across the Pacific are, however, highly variable,  
460 with the gear and associated target species varying across islands and regions  
461 (Friedlander & Parrish, 1997; Craig et al., 2008; Houk et al., 2012). Our analyses  
462 encompass regions characterized by a high diversity of fishing gears (Dalzell et al.,  
463 1996; Fenner, 2012) and fish species (Kulbicki et al., 2013), and span a wider  
464 gradient in human population density (0 – 2,235 people per km<sup>2</sup> forereef area) than  
465 that of previous studies. As such, we show that size-selective exploitation is a  
466 pervasive issue on coral reefs at ocean-basin scales, which consistently alters reef  
467 community size structure.

468

469 Altered community size structure also may have important functional  
470 consequences that extend beyond a loss of large-bodied individuals. Size structuring  
471 of trophic interactions on coral reefs (Robinson & Baum, 2016) means that  
472 communities with steeper size spectra will have a lower mean trophic level (Jennings  
473 et al., 2002), consistent with evidence that the mean trophic level of reef fisheries  
474 catch is negatively correlated with human population density (Houk et al., 2012).  
475 Moreover, communities dominated by smaller individuals have faster rates of  
476 population growth (Brown et al., 2004; Blanchard et al., 2012) and biomass turnover  
477 (Jennings & Blanchard, 2004), and communities with lower mean trophic level may  
478 be less stable (Blanchard et al., 2012; Rochet & Benoit, 2012; Britten et al., 2014) and  
479 more sensitive to environmental change (Jennings & Blanchard, 2004). Exploitation  
480 of large size classes also may release prey populations from predation pressure and  
481 thus further steepen size spectra (Daan et al., 2005). However, such cascading effects  
482 may be difficult to detect in reef systems in which predator-prey interaction strengths  
483 are dampened due to apex predators feeding across large spatial scales and across

484 trophic levels (McCauley et al., 2012; Frisch et al., 2014; Frisch et al., 2016; Roff et  
485 al., 2016). In addition, exploited reef fisheries likely also target medium- and small-  
486 bodied fishes, thus depressing any compensatory growth by prey populations.  
487 Disentangling the combined effects of trophic release of prey populations and  
488 exploitation of smaller size classes therefore remains problematic, but shifts in  
489 community size structure along human disturbance gradients may provide an early  
490 warning of impacts on functional properties at the community level.

491

492 Human-associated declines in total biomass and large fish biomass have been  
493 documented globally across distinct coral reef regions (Roberts, 1995; Mora, 2008;  
494 Cinner et al., 2012a; MacNeil et al., 2015; Williams et al., 2015; Nash & Graham,  
495 2016), but the link between community size structure and biomass has not previously  
496 been examined. We found that gradual declines in size spectra slopes along either  
497 human covariate gradient contrasted with a rapid decrease in reef fish biomass from  $>$   
498  $1500 \text{ kg ha}^{-1}$  at unpopulated islands to  $< 600 \text{ kg ha}^{-1}$  at islands with the lowest human  
499 presence. These different patterns likely arose because biomass estimates are most  
500 strongly influenced by the number of large-bodied fish that are present (Nash &  
501 Graham, 2016), whereas size spectra respond to shifts across the entire distribution of  
502 body sizes from the smallest to largest fish, and treat each individual fish equally. At  
503 the most degraded reefs where large fishes are absent, fishing of medium- and small-  
504 sized fish would further deteriorate community structure but cause less dramatic  
505 reductions in total community biomass. In contrast, the size spectra of lightly fished  
506 reefs were similar that of an undisturbed size spectrum despite supporting biomass  
507 values typical of more heavily disturbed communities. The differential response of  
508 community size spectra and community biomass suggests that community size

509 structure may be more resilient than total biomass to light exploitation. These findings  
510 are consistent with patterns at coral reefs in the Indian Ocean where the functional  
511 composition of fished reefs remains partially intact at biomass levels  $> 600 \text{ kg ha}^{-1}$ ,  
512 despite total biomass falling far below that of neighbouring unexploited sites  
513 (McClanahan et al., 2015). Although recovery of reef fish biomass towards natural  
514 baseline levels is an important conservation target that aims to restore ecosystem  
515 properties by preserving functionally important species (Knowlton & Jackson, 2008;  
516 Bellwood et al., 2011; MacNeil et al., 2015), the maintenance of productive fisheries  
517 in populated regions is also a priority (Cinner et al., 2012b; Zeller et al., 2015).  
518 Rebuilding community size structure in exploited regions is a realistic management  
519 target that may be achieved without implementing the fisheries closures necessary for  
520 rebuilding pristine biomass (MacNeil et al. 2015). Management for the recovery of  
521 community size structure would also benefit from assessments of the influence of  
522 shark and jack populations on spectra slopes, as these top predators likely play  
523 important roles in structuring reef food webs (Bascompte et al., 2005; Rooney et al.,  
524 2006) but are largely absent in heavily exploited regions (Roff et al., 2016).

525

526       Although human covariates were the strongest predictors of size spectra,  
527 additional variation was attributed to differences in sea surface temperature.  
528 Metabolic principles predict that, in warmer environments, increases in individual  
529 energy demands drive greater per-capita consumption rates and strengthen top-down  
530 control of prey populations (Bruno et al., 2015; DeLong et al., 2015). Therefore, in  
531 agreement with our results, warmer islands should be characterized by shallower size  
532 spectra (lower abundance of small bodied fish relative to large bodied fish). However,  
533 difficulties with small-scale UVC methods in accurately enumerating large predator

534 populations (Ward-Paige et al., 2010) prevented the inclusion of some groups of large  
535 predators in our size spectra analyses; our results, therefore, can provide only  
536 incomplete evidence in support of stronger top-down control. Although metabolic  
537 approaches have provided valuable insights into environmental constraints on reef  
538 fish community biomass and trophic structure (Barneche et al., 2014, 2016),  
539 theoretical predictions of the effect of temperature on reef fish size distributions are  
540 lacking. Since size spectra were robust across gradients in other environmental  
541 covariates, improved understanding of temperature control of size spectra would help  
542 the development of predictions of natural baselines for reef fish community size  
543 structure. Such understanding also is increasingly important as climate change warms  
544 reef systems and degrades fish habitat, further stressing reef fish populations (Hoegh-  
545 Guldberg et al. 2007).

546

547 We also detected strong influences of oceanic productivity and habitat  
548 complexity on reef fish biomass. These patterns are broadly consistent with previous  
549 observations that high oceanic production promotes planktivorous fish abundance  
550 (Barneche et al., 2014; Williams et al., 2015) and high structural complexity promotes  
551 survival of small-bodied fishes (Graham & Nash, 2012; Rogers et al., 2014).

552 Subsequent increases in energy availability to upper trophic levels promote greater  
553 total community biomass (Friedlander et al., 2003; Cinner et al., 2009; Williams et al.,  
554 2015) though, interestingly, these apparent differences in energy availability did not  
555 affect size spectra. The lack of a strong response by size spectra at the island scale  
556 suggests that the extra biomass afforded by high productivity and habitat complexity  
557 may be equitably redistributed among all body sizes. Temperature was also a positive  
558 influence on biomass. A previous analysis of the CREP dataset detected this effect

559 only in planktivorous fishes (Williams et al., 2015), and other studies have variously  
560 noted positive (Richards et al., 2012) and negative (Barneche et al., 2014) effects of  
561 temperature on reef fishes, indicating that further study of the influence of  
562 temperature on biomass is warranted.

563

564 The apparent lack of environmental influences on size spectra - with the  
565 exception of temperature - across islands that varied greatly in environmental setting  
566 and biogeographic context supports the utility of size spectra as a robust ecological  
567 indicator of fishing. In temperate systems, size-based indicators have proven to be  
568 powerful methods of assessing exploitation effects across communities of different  
569 compositions (Bianchi et al., 2000; Shin et al., 2005). In reef fisheries, which typically  
570 lack adequate catch and survey data (Sadovy, 2005), UVC monitoring programmes  
571 can provide the body length information required for size spectra analyses (Graham et  
572 al., 2005; Nash & Graham, 2016). Size-based indicators also can effectively link  
573 patterns in community structure with less tangible community-level properties such as  
574 production and biomass turnover rates. Given their sensitivity to environmental  
575 influences and strong response at low levels of exploitation, biomass estimates may  
576 be less reliable as ecological indicators at large spatial scales.

577

578 Although we accounted for several potential sampling issues in our analyses,  
579 size spectra estimates derived from different UVC methods might vary substantially.  
580 Limitations of census methods can introduce error in the counts of small or large size  
581 classes (Bozec et al., 2011) that bias slope estimates or produce non-linear size  
582 spectra (Ackerman et al., 2004). Spectra estimated with biased binning-based methods  
583 (e.g. earlier reef spectra studies (Dulvy et al., 2004; Graham et al., 2005)) can also

584 introduce error in size spectra analyses, while subtle differences between these  
585 methods can even result in spectra slopes that differ by 1 (White et al., 2008).  
586 Difficulties in enumerating fishes accurately across the size spectrum suggest that it  
587 may be problematic to produce meaningful empirical estimates of baseline size  
588 spectra slopes, as has been done for temperate marine ecosystems (Jennings &  
589 Blanchard, 2004). Importantly, by removing some of the largest fish species (the  
590 sharks and jacks) that are heavily targeted by fishers, our results are almost certainly a  
591 conservative estimate of fishing impacts on reefs and are unlikely to match metabolic  
592 predictions for size spectra in which slopes are a simple function of predator-prey  
593 mass ratio and trophic energy transfer efficiency (Brown & Gillooly, 2003; Jennings  
594 & Blanchard, 2004; Trebilco et al., 2013). Instead, size spectra may be most  
595 informative if used to assess relative differences among communities in a space-for-  
596 time approach (as we did here) or to assess temporal changes in community size  
597 structure.

598

599 Across tropical Pacific coral reef ecosystems, islands with a strong human  
600 presence were characterized by degraded coral reef fish community size structure.  
601 Steepening size spectra suggest a shift in size-linked life history traits, implying that  
602 fished communities may have reduced resilience to further exploitation and future  
603 environmental change. Given comparative insensitivity to variation in environmental  
604 conditions, size spectra may prove to be effective ecological indicators of exploitation  
605 impacts on reef fisheries (Graham et al., 2005; Shin et al., 2005; Nash & Graham,  
606 2016). Extreme reductions in reef fish biomass can have potentially wide-ranging and  
607 pervasive consequences for reef ecosystems, particularly when species or trophic  
608 groups that provide key ecosystem functions are depleted (Bellwood et al., 2011;

609 McClanahan et al., 2011; Ruttenberg et al., 2011; McClanahan et al., 2015). However,  
610 despite the loss of biomass at lightly exploited islands, we detected weaker impacts on  
611 size spectra slopes that suggest that maintenance of ecological size structure is a  
612 tangible management target that could enhance the ecological resilience of coral reef  
613 ecosystems.

614

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1022 **Supporting Information**

1023 Appendix S1. Explanatory covariate processing.

1024 Figure S1. Relationship between size spectra slopes and the LFI at populated and  
1025 uninhabited reef areas.

1026 Figure S2. Length spectra slopes across proximity to market (a) and human  
1027 population density (b).

1028 Figure S3. Model parameter estimates for length spectra.

1029 Table S1. Covariate estimates for CREP reef areas.

1030 Table S2. Reef areas surveyed in the CREP dataset.

1031 Table S3. Parameter estimates and model fit for top size spectra model set ( $\Delta\text{AICc} <$   
1032 7).

1033 Table S4. Parameter estimates and model fit for top biomass model set ( $\Delta\text{AICc} < 7$ ).

1034 Table S5. Parameter estimates and model fit for top length spectra model set ( $\Delta\text{AICc}$   
1035  $< 7$ ).

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