

Local human impacts disrupt depth-dependent zonation of tropical reef fish communities

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Nature Ecology and Evolution

DOI: 10.1038/s41559-023-02201-x

Published: 01/11/2023

Peer reviewed version

Cyswllt i'r cyhoeddiad / Link to publication

Dyfyniad o'r fersiwn a gyhoeddwyd / Citation for published version (APA): Richardson, L., Heenan, A., Delargy, A., Neubauer, P., Lecky, J., Gove, J. M., Green, M., Kindinger, T., Ingeman, K., & Williams, G. J. (2023). Local human impacts disrupt depth-dependent zonation of tropical reef fish communities. *Nature Ecology and Evolution*, 7(11), 1844-1855. https://doi.org/10.1038/s41559-023-02201-x

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1 TITLE

2 Local human impacts disrupt depth-dependent zonation of tropical reef fish communities

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33 ABSTRACT

- 34 The influence of depth and associated gradients in light, nutrients, and plankton on the ecological
- 35 organisation of tropical reef communities was first described over six decades ago but remains
- 36 untested across broad geographies. During this time humans have become the dominant driver of
- 37 planetary change, requiring we re-visit historic ecological paradigms to ensure they capture the
- 38 dynamics of contemporary ecological systems. Analysing >5,500 in-water reef fish surveys between
- 39 0–30 m depth on reef slopes at 35 islands across the Pacific, we assess whether a depth gradient
- 40 consistently predicts variation in reef fish biomass. We reveal predictable ecological organisation at
- 41 unpopulated locations, with increased biomass of planktivores and piscivores, and decreased
- 42 primary consumer biomass with increasing depth. Bathymetric steepness also had a striking
- 43 influence on biomass patterns, primarily for planktivores, emphasising potential links between local
- 44 hydrodynamics and the upslope propagation of pelagic subsidies to the shallows. However, signals
- 45 of resource-driven change in fish biomass with depth were altered or lost at populated islands, likely
- 46 due to depleted fish biomass baselines. While principles of depth zonation broadly held, our findings
- 47 expose limitations of the paradigm for predicting ecological dynamics where human impacts
- 48 confound connections between ecological communities and their surrounding environment.

49

50 KEYWORDS

- 51 Ecological zonation; paradigm; spatial scales; biophysical gradients; depth; bathymetric steepness;
- 52 fish biomass; macroecology; coral reefs; human impacts

54 MAIN TEXT

55 Ecological paradigms inform the understanding and management of natural systems but are limited 56 by two fundamental issues. First, biophysical processes governing ecological organisation often occur at regional and continental scales¹⁻³, inherently introducing scale-dependent patterns and 57 58 heterogeneity in observed local community structure^{4–6}. To understand ecological generality, a large 59 enough lens across a land- or seascape is required to encompasses these processes⁷. However, many 60 influential paradigms were developed from single-point studies in the mid-20th century (e.g. ⁶⁻⁸). Until recently our ability to test generalisable predictions on ecological organisation in hierarchically 61 62 structured ecosystems has been limited by a lack of spatially comprehensive data and accessible 63 statistical tools^{7,11}. Second, escalating anthropogenic impacts confound natural drivers of ecological 64 organisation¹²⁻¹⁴ such that humans are now considered the dominant force of planetary change¹⁵. 65 Where theories are founded on a premise that ecological organisation occurs in isolation of anthropogenic forcing^{16,17}, the predictive capacity of historical paradigms can breakdown (e.g. island 66 biogeography theory¹⁸), requiring thorough reassessment of their applicability in this era of rapid 67 change^{16,19}. 68

69 Ecological zonation—the distribution of organisms across space—represents one of the oldest 70 ecological concepts^{8,9,20}. Here we revisit this basic principle in the context of resource driven depth 71 zonation of tropical coral reef communities-the distribution of reef fish biomass among distinct $trophic \ groups. \ Depth \ was \ recognised \ as \ a \ fundamental \ structuring \ force \ over \ six \ decades \ ago^{10,21-23},$ 72 73 and recently shown to be an important predictor of fish diversity²⁴. Ocean-facing reefs are formed 74 along a bathymetric depth gradient with covarying gradients in availability of sunlight, water 75 temperature, surface wave energy, and nutrients^{21,25–27}. With increasing depth, there are predictable changes in energetic resource supply. Light for primary producers diminishes²⁵ but particulate foods 76 77 and nutrients derived from deeper water that support higher trophic levels such as planktivorous 78 predators, secondary consumers, and piscivores^{28–30}, increase with depth with increased proximity 79 to sources of upwelling^{26,31,32}. The degree to which upwelling can boost shallow-water (<30 m 80 depth³³) primary production is, in turn, determined by the steepness of the reef slope — either 81 facilitating or impeding the upslope propagation of deeper nutrient-rich waters to the shallows^{26,31,32}. Where bathymetry mediates local hydrodynamics around islands, upwelling 82 processes can concentrate in specific areas^{34,35}, creating intra-island variation in pelagic resource 83 84 supply³⁶. At larger spatial scales, cross-regional gradients in primary production³ likely underscore background levels of local depth-dependent resource supply. However, despite these important 85 86 structuring forces and a general acceptance of an effect of depth, we have maintained a limited 87 understanding of resource driven depth zonation on shallow coral reefs. Early observations were 88 born of single-location point studies and to date the applicability of the theory remains untested 89 across broad geographies, limiting our understanding of how this ubiquitous physical characteristic 90 of tropical reefs influences natural ecological biomass baselines³⁷. Modern-day island reefs span vast 91 ocean expanses and are among some of the most biodiverse, socio-economically important, but also 92 human-impacted ecosystems on our planet^{19,38}. Traversing numerous biophysical gradients that 93 mediate ecological organisation across scales⁶, they provide a unique focal system to explicitly test 94 early ecological theories across scales, and assess whether classic paradigm-based science and 95 management of contemporary coral reefs should be adapted^{16,19,39}. Reefs are exposed to ocean 96 warming and acidification, and a suite of local human impacts that deplete biological communities and degrade habitats^{16,19}. Some of these human activities are also stratified by depth. Fishing, for 97 98 example, tends to concentrate in shallower depths and disproportion ately targets distinct trophic groups of reef fishes such as large-bodied piscivores, herbivores, and planktivores⁴⁰⁻⁴². With human 99

- impacts on coral reefs globally widespread³⁸, it is unknown how anthropogenic forcing interacts with
 natural energetic resource-supply across depth, and therefore whether a classic depth zonation
- 102 model is effective for predicting the ecological organisation of modern reefs.

103 Establishing meaningful baselines from which to effectively measure change requires spatially 104 comprehensive replication at the un-impacted end of the intact-to-degraded spectrum³⁷. Using a standardised Pacific-wide set of reef fish surveys⁴³, composite data on bathymetric steepness, and 105 106 hierarchical statistical models, we test whether depth zonation patterns in fish biomass are generalisable on tropical coral reefs across broad geographies, and compare patterns in locations 107 108 with and without local human populations. To explicitly assess generality, we isolate the study focus 109 to test a framework of *a-priori* defined hypotheses of the effects of depth, bathymetric steepness, 110 and human population status on the biomass of reef fishes across a broad spatial extent characterised by known environmental and anthropogenic variation^{3,44}. In doing so, we intentionally 111 112 exclude other known influential biophysical and anthropogenic covariates on reef fish biomass (e.g. ^{44,45}) to test the predictive capacity of depth at an ocean-basin scale on the biomass of fishes 113 grouped by their major dietary sources⁴⁶ – primary consumers, planktivores, secondary consumers, 114 115 and piscivores. We link data from 5,525 visual surveys at 35 islands and atolls (hereafter 'islands') across five distinct ecoregions⁴⁷ spanning approximately 4,600 km latitude and 6,800 km longitude in 116 117 the Pacific (Supplementary Table 1) with prior information on unfished biomass baseline estimates 118 to integrate existing global-scale research⁴⁸. We do this to (1) quantify gradients in fish biomass across shallow reef depths (1–30 m) and steepness; (2) compare depth zonation patterns at 119 120 unpopulated versus human populated locations; and (3) estimate the scale -dependency of observed

121 patterns by quantifying variability in fish biomass across reefs, islands, and ecoregions.

122 **RESULTS**

123 Hierarchical Bayesian regression estimates showed that at unpopulated islands there was evident

- depth zonation in the biomass of all trophic groups of reef fish across or within the 0–30 m depth
- range (Fig. 1 and 2; Table 1; Extended Data Fig. 1; Supplementary Tables 4 and 5). However, the
- effect of depth on fish varied by trophic group (Fig. 2), evidenced by differences in estimated effect
 sizes (Fig. 1; Supplementary Table 4) and probability from model posterior draws (Table 1;
- 128 Supplementary Table 5). The biomass of planktivores and piscivores increased across the 0–30 m
- depth gradient with high probability ($P(\beta depth > 0) = 0.98$, both), while primary consumer biomass
- decreased with increasing depth from 0–30 m ($P(\beta depth>0) = 0.93$; Fig. 2; Table 1; Supplementary
- 131 Table 4). The biomass of secondary consumers increased from 0–10 m depth, plateaued from 10–20
- m, and then decreased from 20–30 m (Table 1; Fig. 2). Combining all trophic groups, total fish
- biomass increased from 0–20 m depth ($P(\beta depth>0) = 0.75$; Supplementary Table 5), and plateaued
- from 20–30 m (Table 1). Overall, human population status had a negative effect on the biomass of all
- trophic groups (all: $P(\beta population status < 0) = 1.00$; Fig. 1; Supplementary Table 5), with posterior
- estimates of fish biomass at populated islands consistently lower than at unpopulated islands across
 0-30 m depth (Fig. 2). The greatest negative effect of human population status was on the biomass
- 138 of piscivore reef fish (Fig. 1 and 2; Supplementary Table 4).
- 139 After establishing the direction of change in fish biomass per trophic group over an increasing depth
- range (Fig. 2; Table 1; Supplementary Table 4), we next sought to determine the magnitude of
- 141 change in fish biomass across the study depth range, across nominally shallow (0–10 m), mid-depth
- 142 (10–20 m), and deep (20–30 m) sites. We examined the density distributions from the model
- posteriors of predicted biomass changes across each depth bin and for each trophic group (Fig. 3A),
- 144 and compared the difference in depth zonation measured as absolute change in biomass at

145 populated versus unpopulated islands (Fig. 3B). At populated islands, the magnitude of change in 146 fish biomass across depth was reduced (total biomass, planktivores, piscivores, and secondary 147 consumers) or not observed (primary consumers) relative to patterns observed at unpopulated 148 islands (Fig. 2, Fig. 3A and B; Supplementary Table 7). Total biomass increase was predominantly 149 lower at populated locations across 0–20 m, piscivore and planktivore biomass increases were lower 150 across 0-30 m, and secondary consumer biomass increase was reduced within the shallow 0-10 m 151 range (Fig. 3A and B; Supplementary Table 7). We observed little change in primary consumer 152 biomass across depth at populated islands (Fig. 3A and B; Supplementary Table 7). Examining 153 zonation as a function of proportionate change in biomass across depth, there was greater observed 154 depth zonation at populated islands in the biomass of secondary consumers and piscivores than at 155 unpopulated islands (Fig. 3C; Extended Data Fig. 2; Supplementary Table 8). These inverse trends in zonation, indicating greater proportionate change with depth at populated islands, were likely 156 157 driven by lower biomass baselines and higher incidences of zero-count observations across all fish 158 groups, but most notably for piscivores, at populated islands and in shallower depths than at 159 unpopulated islands (Supplementary Table 9).

160 Incorporating site-level derived estimates of mean bathymetric steepness (°) from within a 400 m

161 buffer radius into fish biomass models revealed confounding and variable effects of forereef

162 steepness on the biomass of planktivores, primary consumers, and total fish biomass

163 $(P(\beta steepness > 0) \ge 0.92$; Fig. 4; Extended Data Fig. 1; Supplementary Tables 4 and 5). Increased reef

164 steepness between 0 to an initial threshold of approximately 5–10° strongly correlated with a two-

165 fold increase in biomass of planktivores, 25% increase in biomass of primary consumers, and 50%

increase in total fish biomass (Fig. 4). Planktivore biomass increased between 0–30° with the
 sharpest increase between 0 to 5–10°, and plateauing around 30°. Conversely, total biomass and

sharpest increase between 0 to 5–10°, and plateauing around 30°. Conversely, total
 primary consumer biomass plateaued around 10°, then declined from 30° (Fig. 4).

169 The proportion of variation in fish biomass explained by each spatial scale, quantified by extracting 170 the posterior standard deviations of these modelled random effects, varied among trophic fish 171 groups (Fig. 5; Supplementary Tables 10 and 11; see Extended Data Fig. 3 for variation in island-level 172 depth effects among ecoregions). For all trophic groups except secondary consumers, there was 173 greater variation in biomass at the site-scale (25–52% variance) and ecoregion-scale (25–66%) than 174 at the island-scale (8–22%), suggesting that intra-island and inter-regional processes are more 175 influential on fish biomass distributions than those occurring among islands (Fig. 5; Supplementary 176 Table 10). The biomass of secondary consumers was most variable at the site scale (63%), relative to the island and ecoregion scales (20%, 17% respectively; Fig. 5; Supplementary Table 10). There were 177 178 high probabilities across all fish groups that variation was greater at the site-scale than island-scale 179 $(P(sdsSITE > sdsISLAND) \ge 0.99;$ Supplementary Table 11). For planktivores, variation in biomass was 180 proportionately greatest at the ecoregion-scale (66% variance). For all other groups except 181 planktivores and piscivores, site-scale variance was greater than at the ecoregion-scale (P(sdsSITE >

182 sdsECOREGION) \geq 0.85) (Fig. 5, Supplementary Tables 10 and 11).

183

184 **DISCUSSION**

185 While the structuring force of depth on reef ecology featured among the earliest descriptions of

186 tropical coral reefs (for example, zonation in species composition)^{10,22,23}, these observations were

187 restricted to single-point locations. To date, the generality of resource driven depth zonation in fish

188 biomass remains untested across broad geographies. Here, we show that in the absence of local

189 human populations there are predictable changes in tropical fish biomass with depth that track expected gradients in energetic resource supply to reefs^{25,28}. These patterns hold true across the 190 study area which spans distinct biogeographic regions, with high spatial consistency across islands 191 192 and ecoregions (Extended Data Fig. 3) despite varying spatial heterogeneity in fish biomass among 193 trophic groups, suggesting the role of distinct scale-dependent drivers. Recent work details declining 194 patterns of reef fish diversity with increasing depth from the shallows to the mesophotic zone (max 195 150 m depth)²⁴. We build on these findings by revealing a common degree of ecological organisation 196 in relation to both depth and bathymetric steepness across geographically distinct reefs. However, 197 while there was evidence of resource-driven depth zonation in some groups at human-populated 198 islands, the absolute change in biomass with depth relative to unpopulated islands was much 199 reduced (planktivores, secondary consumers, piscivores) or was absent (primary consumers). Where 200 humans have fundamentally re-set standing biomass baselines⁴⁸, changes to naturally observed zonation may signal biological depletion confounding the predictive capacity of depth-dependent 201 202 gradients in resource supply. These findings support calls for revisiting and potentially updating 203 twentieth century ecological paradigms (e.g. island biogeography theory¹⁸) that may no longer 204 capture ecological patterns and processes in a human-dominated world¹⁶.

205 At geographically distinct unpopulated islands, we show that reef fish biomass of all broad trophic 206 groups correlated predictably and relatively consistently across depth despite underlying variation in biophysical drivers known to affect standing reef fish biomass^{3,29,45}. Focussing on the shallowest 0–30 207 m, we show that secondary consumer biomass increased between 0–15 m then plateaued. This 208 209 diverse trophic group includes macro- and sessile invertivores and omnivores whose biomass can 210 vary differentially with depth at local scales^{33,49}. However, broad energetic pathways are governed by primary productivity, generally with nutrient limitations in the shallows³² but greater productivity 211 212 with depth, which at broader scales may cause the more consistent zonation pattern of this 213 group^{25,50}. Planktivore and piscivore biomass increased across 0–30 m depth, likely reflecting the 214 increased proximity to pelagic energetic subsidies^{25,26,32} delivered by upwelling that support the 215 growth of planktonic prey for planktivorous fishes^{51,52}. These planktivorous fishes are in turn prey for 216 higher trophic level piscivores²⁸, such that the distribution of piscivores can be indirectly governed by the availability of pelagic energetic subsidies^{29,44,53}. Notably, we observed an approximate two-217 218 fold increase in the biomass of planktivores and piscivores between 0–10 m depth. If the biomass of 219 these groups tracks the availability of their dietary targets, then this trend might be driven by the limited subsidies reaching the shallows. The upslope delivery of pelagic subsidies from deeper 220 221 waters can be highly variable and upwelled waters are often depth-restricted to below 10 m depth²⁵. 222 This can be due to friction caused by the reef topography slowing the propagation of these nutrient-223 rich waters up the reef slope⁵⁴ and limiting their positive benefits on the concentration of 224 zooplankton to greater depths²⁵. In contrast, primary consumer biomass decreased with increasing 225 depth, likely limited by the rapid attenuation of light available for photosynthetic algal production 226 with depth²⁵. Across latitudinal gradients, primary consumer biomass is higher in areas of greater 227 irradiance²⁹. Their distribution across depth is therefore also likely driven by the enhanced benthic 228 primary production that occurs in shallower well-lit waters.

Reef fish biomass also exhibited a striking and varied relationship with bathymetric steepness, primarily in planktivores and primary consumers. For this study, we measured average steepness at the site-scale using a 400 m radial-buffer. The correlation between steepness at this scale and fish biomass may reflect the role of localized hydrodynamics³⁴ and physical hydrodynamic interactions with the benthos⁵⁴ in determining the delivery of nutrient-rich subthermocline water up onto shallow reefs²⁶. While the presence of a depth gradient is a fundamental physical feature of every tropical coral reef island and atoll in the world, the steepness of this gradient can vary. As such, the 236 two can combine to determine the influx of pelagic subsidies to otherwise nutrient-poor tropical 237 waters³¹ and likely set natural limits on the distribution and productivity of reef fishes^{55–58}. Previous 238 studies employing estimates of reef steepness derived at broader spatial scales (e.g. 10 km site 239 buffer radius) have found inconsistent effects of reef steepness on fish productivity⁵², possibly highlighting critical scale-dependency in localised upwelling processes created by physical features 240 like internal waves³⁶. Planktivores and primary consumers are strongly influenced by energetic 241 subsidies to coral reefs^{29,44,59}, and their biomass is naturally higher in areas of higher oceanic primary 242 243 production^{45,52}. Our results show that planktivore biomass increased between 0–30° with the 244 sharpest increase observed within the initial 0 to 5–10°. Indeed, an increase in reef steepness from 245 just 0 to 5–10° yielded twice the biomass of planktivores and a 25% increase in the biomass of 246 primary consumers. Notably, a threshold of approximately 0-10° steepness is required for 247 enhanced nearshore primary production around these islands and atolls³¹. At steepness levels of 30-44°, planktivore biomass plateaued, and primary consumer biomass declined. This may indicate a 248 249 threshold of critical slope steepness where internal waves rich in deep-water pelagic nutrients 250 become more likely to be reflected back offshore than to propagate upslope and/or break at steeper 251 topography^{35,60,61}. We thereby provide novel ocean-basin scale evidence quantifying the influence of 252 local-scale reef steepness on patterns of reef fish biomass. Combined, the results suggest the 253 existence of lower and upper local-scale thresholds in critical reef steepness in mediating delivery of 254 allochthonous subsidies into the shallows and that these effects propagate through to determine the 255 natural carrying capacity of specific trophic groups of reef fishes. Previous studies document variable 256 peaks in planktivorous and piscivorous fishes at mesophotic depths beyond the 30 m limit of this 257 study^{49,62}. These variable peaks may be indicative of spatial variation in upwelling, potentially linked 258 to-among other oceanographic factors and associated changes in benthic composition³³-differences 259 in local bathymetric steepness among those study locations.

260 Despite marked bathymetric gradients in fish biomass at unpopulated islands, we show that depth 261 related changes in biomass were altered by depleted biomass baselines at islands inhabited by 262 people. There was overall lower fish biomass across the depth gradient for all trophic groups at 263 populated locations. Further, the change in absolute biomass of planktivores, piscivores, and secondary consumers across depth was substantially reduced at populated islands, and depth 264 265 zonation in primary consumers was lost. Conversely, when measured as percent change in biomass, 266 depth zonation was greater on populated islands for secondary consumers and especially piscivores. However, at populated islands overall lower biomass baselines of all groups and frequent absence of 267 268 piscivores in shallow water surveys (i.e. zero-count survey observations) likely served to artificially 269 inflate proportionate change across depth. These findings add to mounting global evidence of 270 humans changing fundamental ecological organisation on tropical reefs^{16,19}. Human-driven declines 271 in reef fish biomass even at relatively low levels of human exploitation are well documented in the 272 Pacific^{44,63}. Fishing reduces the overall standing biomass of reef fishes across trophic groups⁴⁴, often with marked losses of piscivores and herbivores^{42,64–66}. Our findings of diminished biomass of 273 274 primary consumers from shallow depths and piscivores and planktivores between 0-30 m on $populated\ reefs\ does\ not\ exclude\ the\ possibility\ of\ mesophotic\ refugia\ for\ depth\ -generalists^{33,42,49,67}.$ 275 276 It does, though, underscore the vulnerability of herbivorous fishes that are largely restricted to 277 shallow reef zones⁶⁸. We note that human impacts on reef fish assemblages are not limited to the 278 effects of fishing^{16,63}. Global warming interacts with local threats such as land-use related 279 sedimentation and nutrient loading into watersheds, dredging, plastic pollution, and invasive 280 species^{16,69}. As a result of these multiple stressors, underlying relationships between reef organisms 281 and their surrounding environmental settings have been blurred across the region^{14,39}. Our findings 282 show that by using human population presence/absence as a simple binary predictor of these

- impacts, natural zonation signals of absolute fish biomass change across depth are substantially
 reduced or are lost at populated islands, with variable responses among trophic groups. Such
 evidence emphasises the critical need for greater protection for reef ecosystems from a suite of
 human impacts^{16,38}, and in particular for depth-constrained trophic groups that perform distinct and
- 287 important functions. 288 Natural resource management is most effective when applied at scales aligning with (or broader than) scales of ecological variance⁷⁰. This is because ecological communities exist in relation to their 289 290 environment at spatial scales where structuring biophysical processes dominate to limit or promote 291 the abundance of competitive organisms^{1,2,50}. For example, intra-island gradients in surface wave 292 energy and localised upwelling can determine the abundance and spatial-clustering of benthic 293 organisms on coral reefs^{36,71}. For reef fish assemblages, inter-island variability in oceanic productivity 294 and island geomorphology can mediate levels of species richness and functional redundancy⁵. 295 Conversely, finer-scale habitat quality (i.e. complexity and cover) can be more influential in 296 determining other aspects of fish assemblage diversity and the abundance of particular groups and
- species^{5,6}. As a result, variation in these biophysical processes through space can create inherent
- 298 ecological heterogeneity across those spatial scales^{4,70}.

299 While there was minimal observed deviation from the global depth effect across the study islands 300 and ecoregions, our results show that spatial variation in fish biomass—across site, island, and 301 ecoregion scales—was differentially and unevenly distributed among trophic groups, underscoring the importance of scale of observation in ecological enquiry^{1,72}. The 35 islands and atolls in this study 302 303 span much of the western and central Pacific and encompass substantial biophysical gradients that 304 influence ecological organisation — ranging from local variation in live coral habitat availability among 305 reefs for example, to regional disparities in temperature, irradiance, and primary production^{3,14,29,44}. 306 These scale-dependent gradients can influence the biomass of herbivorous, planktivorous, and 307 piscivorous reef fishes observed across the region^{44,45,52}. We found that the greatest spatial variance 308 was at the site-scale for primary and secondary consumers, piscivores, and total biomass. We note 309 that unmeasured temporal stochasticity at the site-level due to factors like fish recruitment, 310 mobility, or behaviour can influence small-scale single time point observations and their associated 311 variability at that scale⁶. Nonetheless, the importance of site-scale characteristics, indicated by this intra-island heterogeneity, supports numerous studies that identify habitat availability⁷³, local 312 313 hydrodynamics⁵, and local disturbances^{16,63,74} as predominant mediators of the biomass of those 314 groups⁶. Spatial variance at the site-level was particularly high (63%) for secondary consumers, emphasising trends in location-specific variability in their biomass-depth relationships described in 315 316 previous studies^{33,49}. Conversely, spatial variance in planktivore biomass was greatest at the 317 ecoregion-scale, pointing towards regional disparities in primary production³ and the availability of pelagic subsidies as a primary correlate in the distribution of planktivorous fish biomass^{30,52}. These 318 319 findings align with previous studies that describe habitat composition at the site-level to be the likely 320 dominant driver of reef fish metacommunity structure, including diversity and the biomass of most 321 trophic groups, while attributing greater prevalence of planktivores to larger-scale gradients in 322 overall ocean productivity⁶. That we observed lower variation at the island-scale than site and 323 ecoregion scales may be due, in part, to a dominant influence of local variation in habitat, 324 hydrodynamics, or disturbances and variable background levels of productivity across ecoregions, 325 over processes occurring at the island-scale. In this context, our findings suggest that management 326 of primary consumers, piscivores, and especially secondary consumers might achieve satisfactory 327 outcomes at local within-island scales with no-take areas⁷⁵ (assuming a source of larval supply), 328 habitat restoration, or better regulated destructive human activities⁷⁴. Local management of 329 planktivores is no doubt also important³⁰, but given the potential influence of regional-scale drivers

- 330 on planktivore biomass production and the importance of this group as the prey base for higher
- trophic levels²⁸, more nuanced, region-specific targets for recovery⁷⁶ or catch of planktivores may be
- advisable in areas of naturally lower primary production.
- 333 In revisiting one of the oldest ecological theories-energetic resource-driven depth zonation of
- tropical coral reef communities, to date untested at scale^{10,21–23}–we provide evidence of
- 335 generalisable depth zonation spanning islands across the Pacific. While the principle of resource-
- driven depth zonation held at both unpopulated and populated islands for some trophic groups (e.g.
- direction of change for planktivores, piscivores, secondary consumers), their magnitude of change
- with depth (i.e. absolute biomass) was substantially reduced at populated islands. For functionally
- important primary consumers, the depth zonation signal was conspicuously absent at populated
- islands. Therefore, while we broadly evidence sustained depth zonation in these contemporary reef
 systems, our findings expose limitations of the energetic resource-driven zonation paradigm for
- systems, our findings expose limitations of the energetic resource-driven zonation paradigm for
 predicting ecological dynamics where human impacts increasingly confound connections between
- 342 predicting ecological dynamics where human impacts increasingly confound connect
 342 acclosical communities and their surrounding environment^{12,14,39}
- ecological communities and their surrounding environment^{12,14,39}.
- 344

345 **METHODS**

346 Study location

- 347 To examine the fish zonation across depths and investigate how humans may impact natural
- 348 zonation on coral reefs, we used monitoring data from a standardized dataset that spans the central
- and western Pacific⁴³. Specifically, 5,525 distinct surveys from 2,253 forereef sites (\leq 30 m depth)
- 350 conducted on 35 US and US-affiliated islands and atolls across 42 degrees (°) of latitude (14° S to
- 28° N), and 62° of longitude (178° W to 145° E). The data were collected between 2010–2014 for the
- 352 National Oceanic and Atmospheric Administration (NOAA) Pacific Reef Assessment and Monitoring
- 353 Program (RAMP; 2010-2012) and NOAA's National Coral Reef Monitoring Program (NCRMP; 2013-
- 2019), conducted by the Ecosystem Sciences Division (ESD) of NOAA's Pacific Islands Fisheries
- 355 Science Center (PIFSC)⁴³. We classified sites around islands and atolls (hereafter 'islands') as
- 'populated' or 'unpopulated' based on unpopulated islands having <50 residents and located >100
 km from the nearest larger human settlement using the 2010 US census
- 358 (<u>www.census.gov/2010census</u>) (*sensu*^{14,63}). Of the 35 study islands, 21 were classified as
- unpopulated (*n* = 2,321 surveys, across 923 sites) and 14 as populated (*n* = 3,204 surveys, across
- 1,330 sites) (Supplementary Table 1). Islands were also classified by their location within ecoregions:
- 361 Hawaii Islands; Line Islands; Mariana Islands; Phoenix, Tokelau, Northern Cook Islands; and Samoa
- 362 Islands⁴⁷. The location of each replicate site was pre-selected by randomised stratified design per
- 363 sampling units of the Pacific RAMP/NCRMP protocol (island, group of small islands, or subsections of
- larger islands). The target sampling domain was hard-bottom substrate, with sampling effort
- stratified by reef zone and depth $(0-6 \text{ m}; 6-18 \text{ m}; 18-30 \text{ m})^{43}$. We constrained the dataset to
- 366 forereef habitat only to remove any possible confounding effects of habitat type on reef fish
- assemblages. Reef depth (m) was recorded by divers *in situ* at survey sites. An online map viewer of
 the study sites is available:
- 369 <u>http://noaa.maps.arcgis.com/apps/webappviewer/index.html?id=da5c18ea60d049588fca5feecb82f</u>
 370 e07
- 371 Reef fish survey data

372 The abundance and body-size of all diurnal, non-cryptic reef fishes were estimated using stationary point count (SPC) surveys (*sensu*^{5,43,44,77–79}). At each site, divers conducted simultaneous visual fish 373 374 counts within 1-4 adjacent, visually-estimated 15-m diameter cylindrical plots, extending from the substrate to the limit of vertical visibility⁴³. First, divers compiled lists of all species observed within 375 376 the survey area over a 5-min period, then counted and estimated the size (total length, TL, to the 377 nearest cm) of listed species present within the cylinder over approximately 30-mins. Surveys were 378 conducted by multiple observers across the study ecoregions and years. NOAA employs extensive 379 training and technical validation protocols to ensure consistency and avoid bias in survey technique, 380 fish species identification, and size estimation⁴³. Full details on SPC survey methods and technical validation steps are available in [⁴³]. To further mitigate any confounding effect of observer bias 381 382 among fish surveys, we included 'diver identity' as a random intercept in all statistical models 383 (described below) (sensu⁴⁸). We calculated individual species biomass from the SPC counts using the allometric conversion as $W = aL^b$, where W is the biomass for individual fishes, L is the estimated 384 385 body length of each individual, and parameters *a* and *b* are constants for each species (from⁸⁰). Fish 386 biomass (g m⁻²) was pooled into total fish biomass and four trophic groups: 'Primary Consumers' (herbivores and detritivores), 'Planktivores', 'Secondary Consumers' (omnivores and benthic 387 388 invertivores), and 'Piscivores'⁴⁶. Taxa that are not typically reef-associated were excluded from the 389 analyses, including tuna, bonito, and milkfish (families *Chanidae*, *Myliobatidae*, *Scombridae*; Supplementary Table 12). Sixteen species of shark, jack, and barracuda (families Carcharhinidae, 390 391 Carangidae, Sphyrnidae) were also excluded from the analyses as these highly mobile, large-bodied, 392 roving piscivores are known to be affected by the presence of stationary divers, typically resulting in systematic over-inflation of visual survey density estimates⁸¹ (*sensu*^{44,48}; Supplementary Table 12). 393 394 Further, the presence of divers among study locations also likely introduces a potential source of 395 differential bias of biomass estimates of these fishes, with 'mobbing' behaviour by jacks, sharks, and 396 barracuda known to occur particularly in remote, unpopulated areas such as the northwestern 397 Hawaiian Islands^{44,45}. Zonation patterns in piscivore biomass were comparable with and without this 398 filtering approach. This suggests that the reported patterns were not an artifact of the data handling 399 choice to exclude some species known to be affected and systematically overestimated by divers 400 (Extended Data Fig. 4). However, model outputs of population level effects of depth and bathymetric 401 steepness showed much greater biomass estimates at unpopulated than populated islands, 402 indicating that conservative exclusion of these species mitigated potential systematic bias associated 403 with the survey method among locations (Extended Data Fig. 4).

404 Bathymetric reef steepness

We derived site-level estimates of bathymetric steepness (°) from depth mosaics created from 405 406 multibeam SONAR, bathymetric LiDAR, and imagery derived depths in ArcGIS Pro v2.7 using the 407 'Slope' tool (Spatial Analyst) (Supplementary Table 2). Resulting mosaics ranged in spatial resolution 408 from 5–50 m. Steepness estimates were calculated by averaging steepness values within a 400 m 409 radial buffer around each replicate SPC site, and manually excluding backreef and lagoon areas and 410 land-mass elevation using NOAA PIFSC ESD habitat map information. All analyses were done in the 411 appropriate Universal Transverse Mercator zone for each island. A radial buffer-size of 400 m was 412 selected to encompass depths that would capture the propensity for pulsed delivery of nutrient-rich 413 sub-thermocline water by upwelling^{82,83} and for this upwelling to propagate onto shallow reef 414 habitats at depths \leq 30 m³¹ (maximum depth range within slope buffers: 596.2 m). Buffer-size extent 415 was also selected to cover a reliable gradient in steepness while avoiding buffers-sizes that would 416 easily extend across small island-masses to include bathymetry on opposite sides of the island. All 417 sites were visually cross-checked for island overlap, and those including >5% radial-buffer 418 bathymetry on the opposite site of a landmass were excluded from all analyses.

419 Statistical analysis

- To determine cross-spatial-scale depth zonation patterns in reef fish biomass, we fitted multi-level
 Bayesian regression models with brms⁸⁴. Fish biomass (biomass density; g m⁻²) was modelled
- 422 separately for each trophic group and total fish biomass, using the following general model:
- 423 $y_i \sim \text{Gamma}(\mu_i, \varsigma)$
- 424 $\log(\mu_{ij}) = \alpha + \beta X_i + Z_i \gamma$

425 where α is the trophic group (or total biomass) specific intercept, β is a vector of population-level

426 regression coefficients relating covariates X_i for observation *i* to the log of the expected biomass

427 density μ_j . Group-level coefficients γ are estimated for random effects encoded in design matrix Z.

428 We used depth and bathymetric steepness here as proxies for biophysical processes which influence

 $429 \qquad \ \ \text{coral reef fish assemblage structure. Population-level effects therefore included for ereef depth (1.3-$

430 30.0 m), bathymetric steepness (0.01–43.78 °; mean 10.53 °), and the interactions of depth with

each of bathymetric steepness and human population status. Near-island primary production can

432 decrease exponentially with increasing island bathymetric steepness³¹. Therefore, potential non-

433 linear effects of bathymetric predictors on fish biomass was included in the interaction of steepness

434 with depth by fitting it as a random effect with a cubic-basis spline³⁰.

435 To account for confounding effects of local human impacts on biomass density, we also included a 436 population level effect for human population status ('populated' or 'unpopulated'). Temporal 437 variability in reeffish survey estimates can be introduced by observers and can also reflect dynamic 438 processes such as those determining inter-annual variation in larval recruitment or nutrient availability across the region 85,86 . We therefore included group level effects for observation year (5 439 years; 2010–2014), and year nested within both ecoregion and island in our models to account for 440 441 this potential variation and avoid potential sampling bias. A group-level random intercept for 'diver 442 identity' was included to account for the potential effect of individual observer bias. By assuming an 443 inherent non-independence within divers and their observations that might affect the estimated 444 means and associated errors of fish biomass (sensu⁴⁸), we were then able to estimate isolated 445 population level effects (i.e. depth, human population status, bathymetric steepness) (*sensu*⁴³). 446 More broadly, by controlling these potential sources of variability, we can more accurately test a-447 priori hypotheses about ecological zonation occurring across spatial scales and with greater

448 inferential strength⁶.

To understand whether the ecological organisation of coral fish assemblages in relation to these 449 450 biophysical processes holds true across varying spatial scales, we first accounted for the inherent 451 hierarchical structure in the data by including random intercepts for ecoregions, islands within 452 ecoregions, and sites within islands ($sensu^4$). We hypothesized that patterns of fish biomass across bathymetric gradients may track scale-dependent biophysical drivers that regulate energetic 453 resource supply to shallow coral reefs^{25,36,87}. For example, regional-scale oceanographic currents and 454 455 sea-surface temperatures drive regional differences in primary production and net resource 456 availability^{3,87}. These net gradients in availability can be modified at smaller spatial scales by 457 oceanographic features interacting with local bathymetry³¹ such that, depending on the prevailing 458 direction of internal tidal energy, upwelling processes can drive strong intra-island gradients in 459 nutrient and planktonic resource supply among sites²⁵. As potential indicators of these scale-460 dependent processes, we then measured the variation in reef fish biomass at these three 461 hierarchical scales (sites within islands within ecoregions) and compared the proportion of the total 462 variation explained by those scales (sensu⁴). We quantified variation by extracting posterior standard deviations of random effects at these distinct geographical scales in the fish biomass models and
 compared them across trophic groups. We further included a random slope term for depth-within-

465 island to account for potentially variable depth zonation of biomass across islands.

466 All models were fitted with a Gamma response distribution, using a log-link function as biomass was positive, continuous, and overdispersed⁸⁸. Fish of each trophic group were not observed in every SPC 467 468 survey. To account for these zero-count observations, hurdle models were used, first fitting the 469 presence-absence of fish biomass as a function of the predictors described above, with a binomial 470 distribution and logit function, and then fitting the non-zero biomass data with the Gamma multi-471 level model outlined above. Where the proportion of zeros was too low to effectively estimate effect 472 sizes in the presence-absence component (i.e. an insufficient contrast between the number of zeros 473 and ones), the use of a hurdle structure affected model convergence and only added noise. This 474 occurred for primary consumers (1.09% zeros) and secondary consumers (0.05% zeros), so for these 475 groups the zero biomass replicates were removed from the analysis and the Gamma model detailed

- 476 above was fitted.
- 477 This study builds on existing knowledge established in previous research that estimated a global
- baseline of total resident reef fish biomass in the absence of fishing⁴⁸. We integrate this prior
- information by using their published posterior biomass estimate (1,013 kg ha⁻¹) as the mean of the
- 480 prior for log of total biomass (α ; converted to g m⁻²) (with standard deviation set at 1):

$$\alpha \sim N(\log(101.3), 1)$$

482 The intercept prior for each trophic group was estimated as a proportion of this total unfished global 483 biomass estimate as approximated in MacNeil et al. [⁴⁸]. The grouping of secondary consumers as

defined in this study (a coarse group based on diverse diet items typically targeted by species 484 485 including invertivores, corallivores, and omnivores⁴⁶) differed to those used in MacNeil et al. [⁴⁸]. Therefore, the intercept prior for this group was determined by the proportion of secondary 486 487 consumers in the total biomass from the present study data, applied to the unfished biomass estimate in [48]. MacNeil et al.48 and our study employ comparable data (i.e. in situ counts of 488 489 diurnally active, non-cryptic reef fish on forereef slopes, excluding sharks and semi-pelagics such as 490 jacks). However, to account for potential differentiating factors between the studies, such as species 491 filtering approaches, census method, or geographical representativeness, we inflated the prior 492 standard deviation in the intercepts for our models by an order of magnitude. Model priors are 493 detailed in Supplementary Table 3 and plotted with unpopulated posterior intercept estimations in 494 Extended Data Fig. 5. Marginal posterior distributions for model parameters were estimated by 495 Hamiltonian Monte Carlo (HMC) sampling, using 10,000 iterations across four chains, with a warm-496 up of 2,000 iterations, and a thinning factor of four. To ensure unbiased parameter estimates (i.e., 497 absence of divergent transitions), we set adapt delta to 0.995 and a maximum tree-depth of 12. 498 Model fits and convergence were assessed with graphical posterior predictive checks and via trace 499 and effective sample size plots, the Gelman-Rubin R-hat diagnostic and Bayesian adaptation of Rsquared⁸⁹. An effective sample size of >1,000 was chosen to determine stable parameter 500 501 estimates⁹⁰. Medians of posterior distributions were calculated to obtain a single point estimate and 502 75% and 95% credible intervals (CIs) were calculated from the respective quantiles of the posterior 503 distributions of all metrics presented. Non-independence of population-level predictors was 504 assessed by plotting bivariate correlations between the posterior samples (MCMC draws) of 505 predictor coefficients and quantifying Pearson correlation coefficients between paired samples 506 (Supplementary Figure 1)⁸⁹. Correlation coefficients were all <5%, bar one: a single pairwise 507 correlation coefficient for hurdle components depth and steepness in the planktivore model which 508 was still relatively low at 28%.

- 509 All analyses were conducted in R 4.2.1⁹¹. Bayesian hierarchical models were implemented in
- 510 cmdstanr using *brms* 2.17.0⁹⁰; probability of covariate effect direction was estimated with
- 511 *bayestestR* 0.10.0⁹²; model information for querying posterior predictions was extracted with
- 512 *tidybayes* 3.0.2⁹³; cross-spatial model variance was plotted with TernaryPlot in *Ternary* 1.2.3⁹⁴;
- 513 model fits assessed using r2_bayes in *performance* 0.9.2⁹⁵, and independence of model predictors
- assessed with ggpairs in *GGally* 2.1.2⁹⁶. Fish symbols used in figures were created with *fishualize*
- 515 0.1.0⁹⁷.
- 516

517 DATA AND CODE AVAILABILITY

- 518 All data and R code used in this study are available at an open-source repository
- 519 (https://github.com/LauraERichardson/Depth-Fish).

520 ACKNOWLEDGEMENTS

- 521 We thank Ivor Williams and the staff at the Pacific Island Fisheries Science Center (NOAA) for
- 522 extensive survey data collection. NOAA's Coral Reef Conservation Program (CRCP) supported and
- 523 funded the National Coral Reef Monitoring Program (NCRMP); CRCP project #743. This research was
- funded by the European Commission supported by a Marie Skłodowska-Curie Sêr Cymru II COFUND 524
- 525 Fellowship (ref: BU191) and a Marie Skłodowska-Curie European Fellowship (ref: 844213 –
- 526 FISHSCALE) awarded to LER. We thank Sarah Bond and Alice Lawrence for useful discussions, Alastair
- Feather and Sivajyodee Sannassy Pilly for technical code assistance, Amanda Merritt for contributing 527
- 528 to figure preparation, Jan Hiddink and an anonymous internal NOAA reviewer for critical feedback
- 529 that contributed to preparation of the manuscript.
- 530

531 AUTHOR CONTRIBUTIONS STATEMENT

532 L.E.R., A.H., and G.J.W. conceived this study; L.E.R., A.H., G.J.W., J.M.G., J.L., and J.A.M.G. designed

- 533 the methodology; A.H. contributed to survey data collection; T.L.K. provided NOAA's fish survey
- 534 data; J.L., J.M.G. computed satellite derived bathymetric slope steepness estimates; L.E.R. and P.N.
- 535 conducted the analyses with input from A.J.D.; L.E.R. led manuscript writing with input from A.H.,
- 536 G.J.W., J.L., P.N., A.J.D., T.L.K., and K.E.I. All authors contributed significantly to the drafts and approved the final version for publication.
- 537
- 538

539 COMPETING INTERESTS STATEMENT

- 540 The authors declare no competing interests.
- 541
- 542 TABLES

543 Table 1 Probabilities of an increase in fish biomass across specified depths at unpopulated (U) and populated (P) islands. 544 Probability estimates are derived from posterior model distributions comparing biomass at one fixed depth versus a deeper

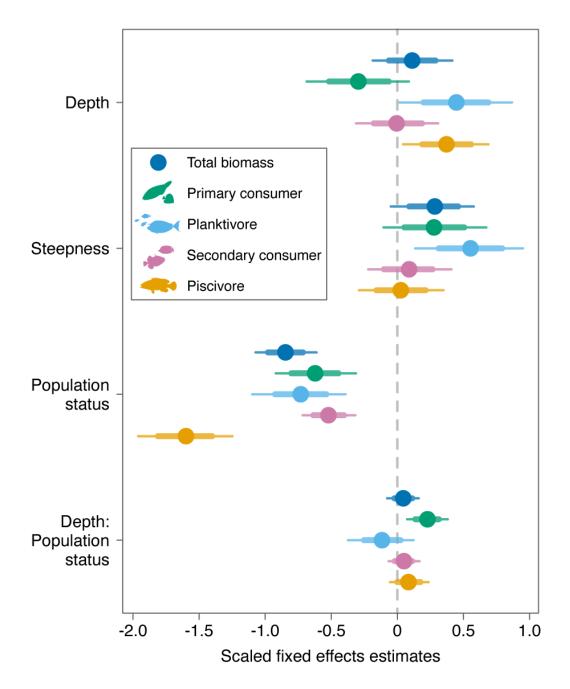
545 depth (i.e., at 0 vs 10 m depth), with steepness held at the study mean value.

Depth (m)	Population status	Total biomass	Primary consumer	Planktivore	Secondary consumer	Piscivore
0 vs 10	U	0.95	<u>0.81</u>	1.00	0.87	1.00
	Р	0.96	0.65	1.00	0.90	1.00
10 vs 20	U	0.95	<u>0.92</u>	1.00	0.63	1.00
	Р	0.98	0.58	0.98	0.73	1.00
20 vs 30	U	0.41	<u>0.95</u>	0.86	<u>0.81</u>	0.84
	Р	0.50	0.29	0.71	0.29	0.78

546 Probabilities ≥75% highlighted in bold. Probabilities ≥75% of the inverse difference (i.e., a high probability of a decrease in 547 biomass with increasing depth) are underlined.

548

549 FIGURE LEGENDS/CAPTIONS (FOR MAIN TEXT FIGURES)



550

551 Figure 1 Relationships between coral reef fish biomass of distinct trophic groups and overall effects of shallow reef depth, 552 bathymetric steepness, and human population status of islands (Population status indicates the effect of 'populated' by 553 humans versus 'unpopulated'). Effect sizes are scaled and includes the interaction of depth with population status 554 (Depth:Population status). Points represent posterior median estimates from Bayesian hierarchical models testing for an 555 effect of each explanatory variable on reef fish biomass, with 75% (thick lines) and 95% (thin lines) percentiles. Explanatory 556 variables were mean-centred and scaled by one standard deviation to facilitate comparisons of effect sizes among them. For 557 estimates of hurdle components (piscivore and planktivore models), see Extended Data Fig. 1 and Supplementary Table 4. 558 Unadjusted Bayesian conditional R² values and 95% CI: total fish biomass, 0.55 (0.44–0.67); primary consumers, 0.54 (0.51– 559 0.57); planktivores, 0.48 (0.24–0.67); secondary consumers, 0.37 (0.31–0.47); piscivores, 0.52 (0.38–0.62) (Supplementary 560 Table 6 for marginal unadjusted R² estimates). N = 5,525 stationary point count (SPC) surveys (across 2,253 forereef sites, 35 561 islands, five ecoregions).

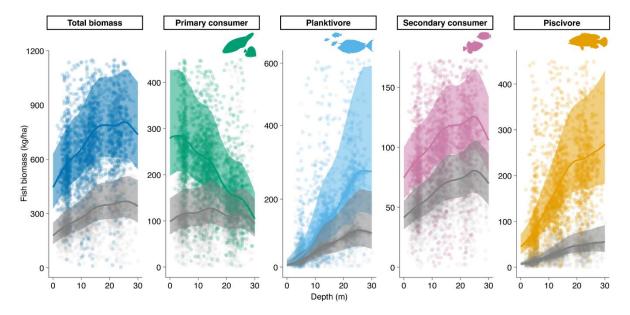
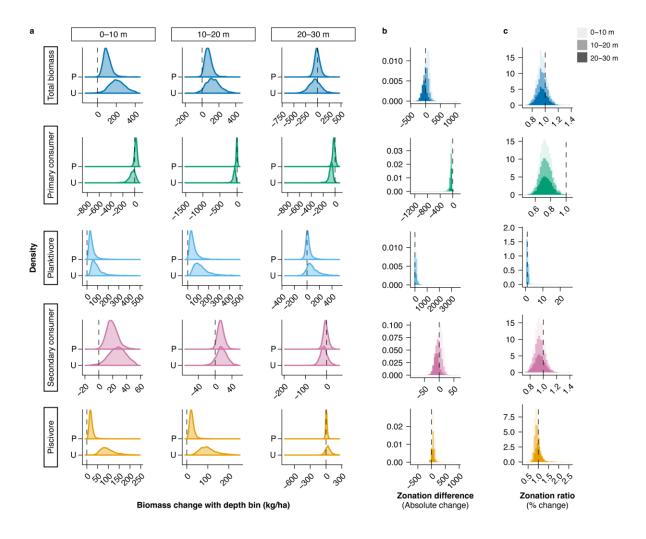


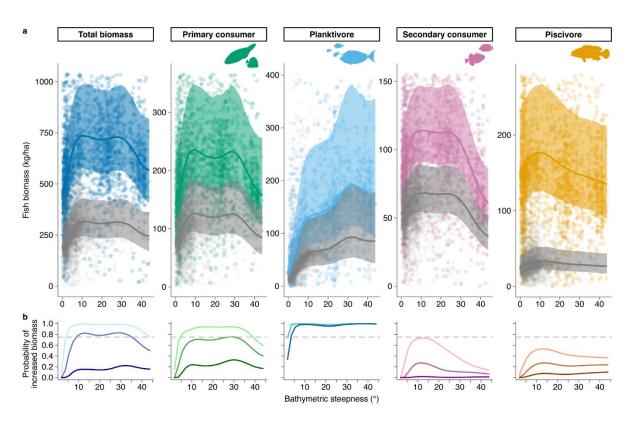


Figure 2 Coral reef fish biomass across a shallow depth gradient at unpopulated (colour) and populated (grey) islands.
 Estimates represent conditional posterior medians (lines), 75% percentiles (shaded areas), and partial residuals (points) at
 the study mean value of bathymetric steepness. The y axis is limited to 1.05x the maximum value of the 75% CI so partial
 residuals exceeding axis limits are not displayed. N = 5,525 stationary point count (SPC) surveys (across 2,253 forereef sites,
 35 islands, five ecoregions).



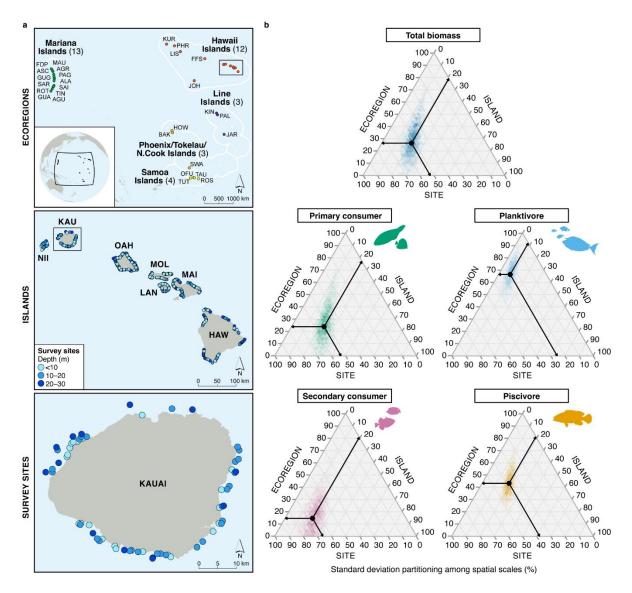
571 Figure 3 Changes in reef fish biomass across shallow reef depth: A) Conditional posterior distribution of changes in biomass 572 of each trophic group of fish (rows labeled on left) with each 10 m depth bin (columns labeled at top), at populated (P) and 573 unpopulated (U) islands. B) Conditional posterior distributions of differences in zonation, measured as differences in absolute 574 increase or decrease in biomass within each depth bin at populated versus unpopulated islands. C) Conditional posterior 575 distributions of zonation ratios, measured as the ratio of percentage change in biomass in each depth bin at populated versus 576 unpopulated islands. For example, panel A and B show a greater increase in absolute biomass of piscivores across depth bins 577 at unpopulated islands than at populated islands (further right of the dotted line), but panel C shows that the zonation ratio 578 of percentage change biomass is greater at populated islands in 0-20 m, spanning two bins (left of dotted line). All plots 579 display change in biomass with depth standardized at the study mean value of bathymetric steepness.

580





582 Figure 4 Coral reef fish biomass across a gradient of reef bathymetric steepness at unpopulated (colour) and populated 583 (grey) islands. A) Estimates represent marginal (integrated over depths from 0-30 m) posterior medians (lines), 75% 584 percentiles (shaded areas), and partial residuals (points) at the study-mean value of depth. The y axis is limited to the 585 maximum value of the 75% CI so partial residuals exceeding axis limits are not displayed). N = 5,525 SPC surveys (across 2,253 586 forereef sites, 35 islands, five ecoregions). B) Probability of increased fish biomass with increasing bathymetric steepness 587 from 0-44°. Coloured lines show the marginal posterior distribution derived probabilities of proportionate increases in reef 588 fish biomass with increasing bathymetric steepness (°) among trophic groups. Shading of coloured lines represent 589 probabilities of biomass increase by 25% (light), 50% (medium), and 100% (dark). Grey line dotted line highlights probability 590 threshold of 0.75.



592

Figure 5 The proportion of residual variation in coral reef fish biomass explained by the hierarchical structure of site, island, and ecoregion spatial scales across the central and western Pacific. A) Maps illustrate the spatial scales (from top to bottom): Ecoregions, example of islands within ecoregions (main Hawaiian Islands shown), and example of sites within islands (Kauai shown), B) Ternary plots of the relative posterior standard deviations explained by the spatial scales for total biomass and each trophic group. Black arrows indicate geometric mean percentage of standard deviations (SD) at each nested spatial scale (median SD estimates and CIs in Supplementary Table 10).

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