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Climate change is projected to reduce carrying capacity and redistribute species richness in North Pacific pelagic marine ecosystems

Running head: Climate change may reduce carrying capacity

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30 **Paper type:** Primary Research

31 **Abstract**

32 Climate change is expected to impact all aspects of marine ecosystems, including fisheries.
33 Here, we use output from a suite of 11 earth system models to examine projected changes in two
34 ecosystem-defining variables: temperature and food availability. In particular, we examine
35 projected changes in epipelagic temperature and, as a proxy for food availability, zooplankton
36 density. We find that under RCP8.5, a high business-as-usual greenhouse gas scenario,
37 increasing temperatures may alter the spatial distribution of tuna and billfish species richness
38 across the North Pacific basin. Furthermore, warmer waters and declining zooplankton densities
39 may act together to lower carrying capacity for commercially valuable fish by 2 – 5% per decade
40 over the 21st century. These changes have the potential to significantly impact the magnitude,
41 composition, and distribution of commercial fish catch across the pelagic North Pacific. Such
42 changes will in turn ultimately impact commercial fisheries' economic value. Fishery managers
43 should anticipate these climate impacts in order to ensure sustainable fishery yields and
44 livelihoods.

45
46 **Introduction**

47 Pelagic marine habitat is projected to experience a number of impacts from climate change (*e.g.*,
48 Bopp *et al.*, 2013). As earth system models improve with each model generation, confidence in
49 their projections has increased and a community consensus is coalescing around several
50 projected impacts. Of these, two of the most significant impacts to epipelagic habitat are likely
51 to be ocean warming (Bopp *et al.*, 2013) and the expansion of the oligotrophic subtropical gyres
52 (Sarmiento *et al.*, 2004; Steinacher *et al.*, 2010; Polovina *et al.*, 2011; Cabré *et al.*, 2015). Ocean
53 warming is a direct result of ocean heat uptake in response to atmospheric warming driven by
54 increasing greenhouse gas concentrations. Gyre expansion is projected as the result of two
55 physical mechanisms. Ocean heating leading to increased vertical stratification is expected to
56 further reduce nutrient concentrations in the euphotic zone of oligotrophic gyre waters (Xu *et al.*,
57 2012; Cabré *et al.*, 2015). Additionally, changes in atmospheric circulation may result in a
58 poleward displacement of both the descending branch of the Hadley circulation and of mid-
59 latitude storm tracks (Chang *et al.*, 2012; Scheff *et al.*, 2012; Yongyun *et al.*, 2012; Cabré *et al.*,

60 2015). These changes in atmospheric circulation will in turn alter ocean surface wind stress curl,
61 primarily along the gyres' poleward boundaries, contributing to gyre expansion.

62
63 Change in the biophysical marine environment will impact many marine organisms, as well as
64 fisheries and those who rely on fishery services. Catch from pelagic fisheries in the North
65 Pacific is largely comprised of tuna, including skipjack (*Katsuwonus pelamis*), yellowfin
66 (*Thunnus albacares*), and bigeye (*Thunnus obesus*), as well as other species (FAO, 2012). Tunas
67 occupy specific thermal habitats at different life stages and have a high metabolic demand
68 (Lehodey *et al.*, 2011, 2013). Thus, changes to either thermal habitat or ocean productivity are
69 likely to impact tunas, as well as other commercially valuable fish. We examine these climate
70 change impacts through a suite of models included in the fifth phase of the Coupled Model
71 Intercomparison Project (CMIP5; Taylor *et al.*, 2012), focusing on the two habitat parameters
72 that most directly influence ecosystem productivity and capacity: temperature and food
73 availability. We aim to make broad projections of climate change impacts on marine fisheries
74 that can be used by fishery managers when drafting ecosystem-based fisheries management
75 plans. Previous studies have suggested that climate change may have a substantial impact on
76 commercial fish catch, independent of fishing (Cheung *et al.*, 2010; Lehodey *et al.*, 2011, 2013;
77 Bell *et al.*, 2013; Howell *et al.*, 2013; Woodworth-Jefcoats *et al.*, 2015). Therefore, it is essential
78 that these potential impacts be incorporated into management plans so that both fishery resources
79 and livelihoods can be sustained well into the future.

80

81 **Materials and methods**

82 Earth system models used

83 We examine 11 earth system models included in CMIP5. Models used are presented in Table 1.
84 Models selected are those with two trophic levels (phyto- and zooplankton) of output available at
85 time of download. All data were downloaded from the CMIP5 data portal ([http://cmip-
86 pcmdi.llnl.gov/cmip5/data_portal.html](http://cmip-pcmdi.llnl.gov/cmip5/data_portal.html)). Spherical interpolation (for curvilinear grids) and
87 nearest coordinate regridding (for rectilinear grids) were used to regrid output to a common 1° ×
88 1° rectilinear grid spanning 0 – 66°N and 120°E – 70°W, with the Bering Sea and Sea of
89 Okhotsk excluded. We note that output from two additional models, HadGEM2-CC and

90 HadGEM2-ES (Collins *et al.*, 2011), was available but not used in our analysis due to unrealistic
91 negative plankton densities across much of the central North Pacific.

92

93 Data used

94 Both historical and projected data are examined. All projections are from the representative
95 concentration pathway (RCP) 8.5 scenario, “a relatively conservative business as usual case...
96 with no explicit climate policy” (Riahi *et al.*, 2011). We focus on two 20-year time periods
97 representing the beginning and end of the 21st century: 1986 – 2005 and 2081 – 2100. The
98 beginning of the 21st century is captured by the last 20 years of the historical runs and the end of
99 the 21st century by the last 20 years of the 21st century in the RCP8.5 projection.

100

101 Data are vertically integrated across the epipelagic zone, represented as the upper 200 m of the
102 water column. Vertical resolution varies by model and we integrated across all depths of 200 m
103 or less. We examine potential temperature, phytoplankton carbon density, and zooplankton
104 carbon density output by each model. Data are examined as a model ensemble to address the
105 possible influence of individual model drift (Sen Gupta *et al.*, 2013).

106

107 Vertically integrated (upper 200 m) ocean temperatures from the World Ocean Atlas 2013
108 (WOA13; Locarnini *et al.*, 2013) were used as baseline temperatures when calculating projected
109 ecosystem impacts (see below). The temperature increase projected by each model was then
110 added to the WOA13 data to determine projected ecosystem change.

111

112 Pelagic habitat

113 We examine changes in thermal habitat by comparing probability frequency distributions of
114 pooled 20-year, monthly epipelagic temperatures. Because so much of the literature is focused
115 on SST, we also present projected changes in SST though these changes are not the focus of our
116 analysis. In both cases, monthly temperatures were used in an effort to fully capture seasonal
117 extremes and distributions are binned in 0.5 °C bins. Change in zooplankton densities are
118 similarly compared, though annual densities are used as this is the only temporal scale available
119 for three-dimensional biogeochemical data through the CMIP5 data portal. Distributions are
120 binned in 0.05 g C m⁻² bins. Twenty-year means from the beginning and end of the 21st century

121 (see above) are used to evaluate the absolute change in epipelagic temperature and percent
122 change in both phyto- and zooplankton densities.

123

124 Ecosystem impacts

125 We assess two measures of ecosystem impact: tuna and billfish species richness and carrying
126 capacity. Species richness captures the total number of tuna and billfish species present and
127 carrying capacity the total number of fish the ecosystem can support. Species richness (SR) is a
128 function of epipelagic temperature, following equation 1 as determined by Boyce *et al.* (2008),

$$129 \quad SR = -0.0033T^3 + 0.1156T^2 - 0.4675T \quad (1)$$

130 where T is epipelagic temperature in °C. Carrying capacity (K) is determined from ecological
131 theory, following equation 2,

$$132 \quad K \propto [R]M^{-3/4}e^{E/kT} \quad (2)$$

133 where $[R]$ is limiting resource supply, which we take as zooplankton density, M is target fish
134 mass, E is activation energy (0.63 eV; Brown *et al.*, 2004), k is Boltzmann's constant (8.62×10^{-5}
135 eV K⁻¹; Brown *et al.*, 2004; Jennings *et al.*, 2008), and T is epipelagic temperature in Kelvin
136 (Brown *et al.*, 2004). Given that Eq. 2 is a proportional relationship, we evaluate relative
137 changes in the right-hand-side of the equation and refer to these as changes in potential carrying
138 capacity (K_p). Twenty-year means from the beginning and end of the 21st century are used to
139 evaluate changes in SR and K_p . We hold M constant over both periods so the resulting change in
140 K_p is independent of M . To assess whether R or T has a greater influence on K_p we examine the
141 difference between the absolute percent change in both R and $e^{E/kT}$ following equation 3,

$$142 \quad |\% \Delta R| - |\% \Delta e^{E/kT}| \quad (3)$$

143 with positive results indicating that changes in zooplankton density have the greatest influence
144 on K_p and negative results indicating that changes in T have the greatest influence on K_p .

145

146 **Results**

147 Pelagic habitat

148 *Thermal habitat*

149 Across all models, sea surface and epipelagic temperatures are projected to increase (Table 1,
150 Figs. 1a and 2a). Additionally, all model scenarios project the emergence of new, warmer
151 temperatures by the end of the 21st century. Emerging SSTs (*i.e.*, temperatures not present at the
152 beginning of the 21st century that are present at the end of the 21st century) range from 31.5 – 38
153 °C and epipelagic temperatures from 29 – 35 °C. Change in thermal habitat is also captured
154 through the difference between frequency distributions over time. Epipelagic temperatures that
155 comprise the majority of the North Pacific at the beginning of the 21st century (15.6 – 23.7 °C on
156 average) decline in frequency and warmer temperatures come to dominate by the end of the
157 century (24.5 – 32.9 °C on average).

158

159 Our results focus on the warmest temperatures in the North Pacific as these temperatures cover
160 the largest area. However, it is important to note that there is a similar distributional shift in the
161 coolest temperatures. Here, too, there is model consensus on a shift towards warmer
162 temperatures, as well as a loss of the coolest temperatures by the end of the 21st century (Fig. 1a).
163 Across all models, disappearing epipelagic temperatures range from -1.5 – 2 °C. Three models
164 (IPSL-CM5B-LR, MPI-ESM-LR, and MPI-ESM-MR) project a loss of the coolest SSTs, ranging
165 from -2.0 – -1.0 °C.

166

167 *Food available to fish*

168 We take zooplankton density to be a proxy for food available to fish. Across all models, the
169 distribution of zooplankton densities is projected to shift towards lower values (Table 1, Figs. 1b
170 and 2b). Densities that comprise the majority of the North Pacific at the beginning of the 21st
171 century (0.50 – 1.10 g C m⁻² on average) decline in frequency and lower densities come to
172 dominate by the end of the century (0.18 – 0.49 g C m⁻² on average).

173

174 Not only do the models used in our study project zooplankton densities to decline across much of
175 the North Pacific, but they also project these declines to be amplified relative to declines in
176 phytoplankton densities (Fig. 2c, warm colors represent waters where zooplankton declines are

177 projected to be greater than phytoplankton declines). When declining zooplankton densities are
178 examined in relation to projected phytoplankton changes, we find that zooplankton declines
179 exceed phytoplankton declines to a large degree. All models but three (CanESM2, GISS-E2-H-
180 CC, and GISS-E2-R-CC) place such waters across much of the North Pacific excluding only
181 subpolar waters, and in some cases equatorial and California Current upwelling waters.
182 Projected declines in zooplankton exceed those of phytoplankton by 10 – 30% on average, with
183 individual model maxima of 25 – 50% found along the periphery of the North Pacific subtropical
184 gyre (NPSG).

185

186 Ecosystem impacts

187 Changes in predicted tuna and billfish species richness (*SR*) follow projected changes in
188 epipelagic temperature. Across all models, the area of maximum *SR* shifts northward and
189 eastward. Species richness declines across much of the central and western subtropics and
190 increases in temperate and subpolar waters, with the magnitude of change increasing with
191 distance toward the western tropical Pacific and temperate latitudes, peaking at approximately
192 four species lost or gained (Fig. 2d). Most models project potential carrying capacity (K_p) for
193 commercially valuable fish to decline by 20 – 50% across the North Pacific, or by roughly 2 –
194 5% per decade over the 21st century (Fig. 2e). As with trophic amplification, the areas projected
195 to see the greatest declines in K_p are found along the periphery of the NPSG. Declining K_p is a
196 result of both increasing epipelagic temperature and declining zooplankton density, with the
197 primary driver varying across the North Pacific. In the western equatorial Pacific and NPSG
198 declining zooplankton density has a stronger impact on K_p , while in the eastern equatorial
199 Pacific and at temperate latitudes increasing epipelagic temperature is the stronger driver (Fig.
200 2f).

201

202 **Discussion**

203 The CMIP5 projections presented in this study suggest a number of changes to North Pacific
204 pelagic habitat. Broadly, thermal habitat is projected to warm and be spatially redistributed.
205 Zooplankton densities are projected to decline and to an amplified degree relative to
206 phytoplankton declines. When these projections are examined more finely and in relation to one

207 another, they suggest that commercial fisheries in the central North Pacific may see catch decline
208 by 20 – 50% and be comprised of three to four fewer tuna and billfish species.

209

210 Changing pelagic habitat

211 While warming epipelagic temperatures might be expected to unfold as a straightforward
212 poleward creep of present-day conditions, we find that this is not the case (Fig. 2a, d). Rather,
213 warmer temperatures appear to emerge from the western equatorial Pacific and expand eastward
214 and northward as moderate temperatures retreat in kind. Over time, this results in a reshaping of
215 pelagic thermal habitat. For example, thermal habitat associated with adult tuna foraging is
216 displaced by thermal habitat more commonly associated with tuna spawning grounds and
217 spawning habitat is replaced by temperatures that exceed even the warmest temperatures
218 associated with commercially valuable fish (Boyce *et al.*, 2008; Lehodey *et al.*, 2011, 2013).

219

220 Evidence suggests that fish and other pelagic organisms will relocate to maintain residence in
221 preferred thermal habitat in both freshwater (Grenouillet and Comte, 2014) and marine (Pinsky
222 *et al.*, 2013; Montero-Serra *et al.*, 2015) environments, and with relocations varying over
223 different life history stages (Walsh *et al.*, 2015). Some fish may simply be able to spend more
224 time in deeper, cooler waters. However, such an adaptation comes at a cost. For example, fish
225 may forage less successfully at the lower light levels found below the epipelagic realm.

226 Organisms that are unable to exploit deeper habitat will be forced to relocate geographically.

227 Such vertical and geographic relocations could ultimately alter predator – prey dynamics.

228

229 The emergence of new thermal habitat also raises questions, as it is projected to exceed current
230 maximum temperatures. It remains unknown how or whether pelagic organisms will adapt to
231 these temperatures. Storch *et al.* (2014) suggest there are firm limits on temperatures to which
232 animals can adapt. They find that due to constraints posed by cellular complexity, the highest
233 SST that allowed multicellular Eukaryota to grow was 40 °C, close to temperatures projected to
234 occur over the North Pacific in our study. The unprecedented rate at which climate is changing
235 (Doney *et al.*, 2014) adds further uncertainty to questions surrounding adaptation.

236

237 In addition to changes in thermal habitat, we also project a shift towards lower zooplankton
238 densities over the 21st century. Spatially, the lowest zooplankton densities are associated with
239 the oligotrophic NPSG. Declining densities are manifested as both an expansion of the NPSG,
240 as well as lower densities in NPSG waters (Fig. 2b). While we examine the oligotrophic NPSG
241 from the perspective of zooplankton densities, our results are similar to those from other studies
242 focused on phytoplankton that project the gyre's expansion (Sarmiento *et al.*, 2004; Steinacher *et*
243 *al.*, 2010; Polovina *et al.*, 2011; Cabré *et al.* 2015).

244
245 Not only do the models used in our study project zooplankton densities to decline across much of
246 the North Pacific, but they also project these declines to be amplified relative to declines in
247 phytoplankton densities. Stock *et al.* (2014) link trophic amplification to declining zooplankton
248 growth efficiency as food resources (net primary production) decline, while Chust *et al.* (2014)
249 link trophic amplification to nonlinear coupling of phytoplankton and zooplankton biomass. It
250 remains unclear whether this amplification in the plankton community will propagate further up
251 through the food web, however modeling work suggests that it will be amplified by some
252 micronekton (Bell *et al.*, 2013) and possibly throughout the size spectrum (Lefort *et al.*, 2015).
253 If trophic amplification does indeed carry through the food web, an amplification of roughly
254 20% at each trophic linkage could result in apex predator density (trophic level 4 – 5) declining
255 by up to 50 – 60% by the end of the century, or by 5 – 6% per decade.

256 257 Ecosystem impacts of changing pelagic habitat

258 The projected impacts of climate change in the North Pacific extend beyond the immediate
259 changes to temperature and food availability. Increasing epipelagic temperature is projected to
260 lead to a redistribution of tuna and billfish *SR* (Fig. 2d). There is strong model agreement of a
261 decline of up to 3 – 4 species across much of the subtropics with an increase of similar
262 magnitude projected for temperate latitudes. These projected changes in *SR* largely, and not
263 surprisingly given Eq. 1, mirror the changing footprint of thermal habitat in the North Pacific.
264 Based solely on thermal tolerance, much of the subtropical North Pacific is projected to become
265 less hospitable to adult commercially valuable tuna and billfish. While a decline of only a few
266 species may not seem very substantial, the longline fisheries in these waters target only a small
267 number of species, primarily bigeye tuna and swordfish (*Xiphias gladius*), and also catch several

268 commercially-valuable, non-target species such as skipjack tuna, yellowfin tuna, shortbill
269 spearfish (*Tetrapturus angustirostris*), and striped marlin (*Tetrapturus audax*). Thus, even a
270 small decline in SR could significantly impact catch composition, magnitude, and value.
271 Likewise, at the northern limits of the fishery, the small increase in species diversity could
272 potentially benefit fishermen. Whether this potential benefit would be offset by the increased
273 expense of traveling further from port to fish is unknown. Fishermen may also shift their
274 homeport based on target catch relocation, as some in the Hawaii-based longline fishery have
275 already done.

276

277 Increasing epipelagic temperatures combined with largely declining zooplankton densities are
278 projected to act together to lower North Pacific K_p over the 21st century. We find strong model
279 agreement that K_p is projected to decline by roughly 20 – 50% across the North Pacific (Fig. 2e).
280 Despite our measure of K_p being a simple relationship based on ecological theory, this projection
281 is in line with previous studies that have projected similar declines in exploitable high-trophic-
282 level biomass as the result of climate change (Lefort *et al.*, 2015; Woodworth-Jefcoats *et al.*,
283 2015). We also find that declines in K_p exceed those of zooplankton densities, further
284 suggesting that trophic amplification in the plankton community may propagate up through the
285 food web. Additionally, K_p is projected to decline even in regions where plankton densities are
286 projected to increase (Fig. 2b, e). This suggests that potential increases in biomass at the base of
287 the food web won't be enough to compensate for the metabolic costs of increasing temperatures.
288 Further examination of the impact of temperature versus zooplankton on K_p shows that the
289 dominant driver of change varies spatially (Fig. 2f). In subtropical regions where zooplankton
290 declines are projected to be greatest, these declines seem to have the greater impact on K_p . In the
291 eastern North Pacific and at temperate latitudes, waters seeing the greatest increase in epipelagic
292 temperature, temperature increases drive K_p declines.

293

294 Potential carrying capacity is projected to decline most in and around the central North Pacific.
295 This has the potential to particularly impact longline fisheries operating in this area. Potential
296 fisheries yields could decline by up to 50% over a time when the Food and Agriculture
297 Organization of the United Nations projects that food resources will need to increase by roughly
298 70% to meet the demands of a growing human population (UN, 2011). Such an increase in

299 demand could further strain the ecosystem, as the heavy removal of large fish has the potential to
300 drive down exploitable biomass independent of any bottom-up impacts (Blanchard *et al.*, 2005;
301 Ward and Myers, 2005; Polovina and Woodworth-Jefcoats, 2013).

302
303 The areas of greatest trophic amplification and declining K_p occur around the boundaries of the
304 NPSG (Fig. 2c, e). To the north of the NPSG lies the North Pacific transition zone, a narrow
305 area used as a migration and foraging corridor by a number of pelagic species (Polovina *et al.*,
306 2001; Hazen *et al.*, 2013). To the south of the NPSG are spawning grounds for a number of
307 tropical tuna species (Lehodey *et al.*, 2011, 2013). Thus, the areas likely to see the greatest
308 declines in food availability are areas crucial to specific life history stages of pelagic species.
309 Such a mismatch in resource demand and supply could amplify climate impacts on species
310 exploiting these regions. Furthermore, given that organisms from around the North Pacific target
311 these areas, changes here have the potential to impact the entire basin. These maxima of
312 declining phyto- and zooplankton densities are not flanked by corresponding areas of increasing
313 densities, suggesting that productive regions around the NPSG are not simply relocating. Or, if
314 productive regions are relocating, they are still experiencing overall declines in phytoplankton
315 densities. The importance of these regions bordering the NPSG, along with their relatively small
316 size, makes them ideal areas for monitoring climate change as it unfolds. Survey (Howell *et al.*,
317 2015; Polovina *et al.*, 2015) and tagging (Block *et al.*, 2011) efforts already in place in these
318 regions may provide insight into how organisms across the food web are responding to climate
319 change.

320
321 One question we are unable to address in this study is how regions bordering the NPSG may be
322 impacted by changes in phenology. The transition zone in particular moves meridionally with
323 the seasons. The phenology of both the seasonal migration of the transition zone (Hazen *et al.*,
324 2013) and its associated productivity (Polovina *et al.*, 2011) may change as a result of climate
325 change. Thus, organisms targeting the region at specific times of the year may have to migrate
326 farther or to different locations. Both finer temporal resolution projections and tagging data may
327 help address such phenology questions.

328
329 Caveats

330 Our study focuses on the two primary influences on ecosystem capacity, temperature and food
331 availability. These are far from the only influences, though. Other variables such as oxygen
332 concentration, pH, and exploitation can influence pelagic carrying capacity. Given that changes
333 in many of these variables are projected to have negative impacts in the North Pacific (Koslow *et*
334 *al.*, 2011; Bopp *et al.*, 2013), they are likely to exacerbate the impacts of warming temperatures
335 and declining food availability.

336
337 We also assume that physical climate influences will be the primary determinants of ecosystem
338 capacity. However, species and trophic interactions are also influential. In some cases, these
339 interactions can have a larger impact than physical climate drivers (Grenouillet and Comte, 2014;
340 Ockendon *et al.*, 2014). Additionally, changes in temperature and food availability can alter
341 foraging range and create new competition (Bond and Lavers, 2014). Such changes in predator –
342 prey interactions could have large impacts on commercial fisheries and could potentially be
343 examined through species-based ecosystem modeling approaches and network theory.

344
345 In this study we examine only the epipelagic realm, though many commercially valuable fish
346 also inhabit mesopelagic depths (Howell *et al.*, 2010; Abecassis *et al.*, 2012). Future impact
347 studies could examine a broader vertical habitat range. For example, Lefort *et al.* (2015) suggest
348 that fishes able to migrate between epi- and mesopelagic depths may fare better in the face of
349 climate change than fishes restricted to either realm. Finally, we examine only one climate
350 change scenario. By examining RCP8.5, we hopefully project the upper limits of potential
351 climate change impacts. Future work could examine more optimistic RCPs, potentially
352 providing motivation to take mitigating actions by presenting goals for limited impacts.

353 354 Commercial fishery impacts of changing pelagic habitat

355 Through examining a suite of CMIP5 earth system models we find that climate change may
356 significantly alter North Pacific epipelagic habitat over the 21st century. Warming thermal
357 habitat and declining zooplankton densities are projected to lower potential carrying capacity,
358 and in turn fishery yield, by approximately 2 – 5% per decade. Additionally, based on changing
359 thermal habitat alone, species richness across much of the subtropics is projected to decline by
360 up to four tuna and billfish species by the end of the century. Together, these changes have the

361 potential to significantly impact commercial fish catch in the North Pacific. Fishery managers
362 can use these projections to place current yields and management actions in a broader climate-
363 based context. For example, early warning thresholds for changing catch composition or yield
364 could be based on projected climate impacts. Such strategic management plans would ensure
365 that the ecosystem is not further stressed by unsustainable removals.

366

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378

379 **References**

380 Abecassis M, Dewar H, Hawn D, Polovina J (2012) Modeling swordfish daytime vertical habitat
381 in the North Pacific Ocean from pop-up archival tags. *Marine Ecology Progress Series*, **452**, 219-
382 236. DOI: 10.3354/meps09583

383

384 Bell JD, Ganachaud A, Gehrke PC, *et al.* (2013) Mixed responses of tropical Pacific fisheries
385 and aquaculture to climate change. *Nature Climate Change*, **3**, 591-599. DOI:
386 10.1038/NCLIMATE1838

387

388 Blanchard JL, Dulvy NK, Jennings S, Ellis JR, Pinnegar JK, Tid A, Kell LT (2005) Do climate
389 and fishing influence size-based indicators of Celtic Sea fish community structure? *ICES Journal*
390 *of Marine Science*, **62**, 405-411. DOI: 10.1016/j.icesjms.2005.01.006

391

392 Block BA, Jonsen ID, Jorgensen SJ, *et al.* (2011) Tracking apex marine predator movements in a
393 dynamic ocean. *Nature*, **475**, 86-90. DOI: 10.1038/nature10082
394
395 Bond AL, Lavers JL (2014) Climate change alters the trophic niche of a declining
396 apex marine predator. *Global Change Biology*, **20**, 2100–2107. DOI: 10.1111/gcb.12554
397
398 Bopp L, Resplandy L, Orr JC, *et al.* (2013) Multiple stressors of ocean ecosystems in the 21st
399 century: projections with CMIP5 models. *Biogeosciences*, **10**, 6225–6245. DOI: 10.5194/bg-10-
400 6225-2013
401
402 Boyce DG, Tittensor DP, Worm B (2008) Effects of temperature on global patterns of tuna and
403 billfish richness. *Marine Ecology Progress Series*, **355**, 267-297. DOI: 10.3354/meps07237
404
405 Brown JH, Gillooly JF, Allen AP, Savage VM, West GB (2004) Toward a metabolic theory of
406 ecology. *Ecology*, **85**(7), 1771-1789.
407
408 Cabré A, Marinov I, Leung S (2015) Consistent global responses of marine ecosystems to future
409 climate change across the IPCC AR5 earth system models. *Climate Dynamics*, **45**, 1253-1280.
410 doi: 10.1007/s00382-0147-2374-3
411
412 Chang EKM, Guo Y, Xia X (2012) CMIP5 multimodel ensemble projection of storm track
413 change under global warming. *Journal of Geophysical Research*, **117**, D23118. DOI:
414 10.1029/2012JD018578
415
416 Cheung WWL, Lam VWY, Sarmiento JL, Kearney K, Watson R, Zeller D, Pauly D (2010)
417 Large-scale redistribution of maximum fisheries catch potential in the global ocean under climate
418 change. *Global Change Biology*, **16**, 24-35. DOI: 10.1111/j.1365-2486.2009.01995.x
419
420 Christian JR, Arora VK, Boer GJ, *et al.* (2010) The global carbon cycle in the Canadian Earth
421 system model (CanESM1): Preindustrial control simulation. *Journal of Geophysical Research*,
422 **115**, G03014. DOI: 10.1029/2008JG000920

423

424 Chust W, Allen JI, Bopp L, *et al.* (2014) Biomass changes and trophic amplification of plankton
425 in a warmer ocean, *Global Change Biology*, **20**, 2124–2139. DOI: 10.1111/gcb.12562

426

427 Collins WJ, Bellouin N, Doutriaux-Boucher M, *et al.* (2011) Development and evaluation of an
428 Earth-System model – HadGEM2. *Geoscientific Model Development*, **4**, 1051-1075. DOI:
429 10.5194/gmd-4-1051-2011

430

431 Doney SC, Bopp M, Long MC (2014) Historical and future trends in ocean climate and
432 biogeochemistry. *Oceanography*, **27(1)**, 108–119. DOI: 10.5670/oceanog.2014.14.

433

434 Dufresne J-L, Foujols M-A, Denvil S, *et al.* (2013) Climate change projections using the IPSL-
435 CM5 Earth System Model: from CMIP3 to CMIP5. *Climate Dynamics*, **40**, 2123-2165. DOI:
436 10.1007/s00382-012-1636-1

437

438 Dunne JP, John JG, Shevliakova E, *et al.* (2013) GFDL’s ESM2 Global Coupled Climate-
439 Carbon Earth System Models. Part II: Carbon System Formulation and Baseline Simulation
440 Characteristics. *Journal of Climate*, **26**, 2247-2267. DOI: 10.1175/JCLI-D-12-00150.s1.

441

442 Food and Agriculture Organization of the United Nations (FAO), 2012: *FAO Yearbook, Fishery
443 and Aquaculture Statistics, Capture Production*.

444

445 Giorgetta MA, Jungclaus J, Reick CH, *et al.* (2013) Climate and carbon cycle changes from 1850
446 to 2100 in MPI-ESM simulations for the Coupled Model Intercomparison Project phase 5.
447 *Journal of Advances in Modeling Earth Systems*, **5**, 572-597. DOI: 10.1002/jame.20038

448

449 Grenouillet G, Comte L (2014) Illuminating geographical patterns in species’ range
450 Shifts. *Global Change Biology*, **20**, 3080–3091. DOI: 10.1111/gcb.12570

451

452 Hazen EL, Jorgensen S, Rykaczewski RR, *et al.* (2013) Predicted habitat shifts of Pacific top
453 predators in a changing climate. *Nature Climate Change*, **3**: 234-238. DOI:

454 10.1038/NCLIMATE1686

455

456 Howell EA, Bograd SJ, Hoover AL, Seki MP, Polovina JJ (2015) Variation in phytoplankton
457 composition between two North Pacific frontal zones along 158°W during winter–spring 2008–
458 2011. *Progress in Oceanography*, In Press. DOI: 10.1016/j.pocean.2015.06.003

459

460 Howell EA, Hawn DR, Polovina JJ (2010) Spatiotemporal variability in bigeye tuna (*Thunnus*
461 *obesus*) dive behavior in the central North Pacific Ocean. *Progress in Oceanography*, **86**, 81-93.
462 DOI: 10.1016/j.pocean.2010.04.013.

463

464 Howell EA, Wabnitz CCC, Dunne JP, Polovina JP (2013) Climate-induced primary productivity
465 change and fishing impacts on the Central North Pacific ecosystem and Hawaii-based pelagic
466 longline fishery. *Climatic Change*, **119**, 79-93. DOI: 10.1007/s10584-012-0597-z.

467

468 Jennings S, Mélin F, Blanchard JL, Forster RM, Dulvy NK, Wilson RW (2008) Global-scale
469 predictions of community and ecosystem properties from simple ecological theory. *Proceedings*
470 *of the Royal Society B*, **275**, 1375-1383. DOI: 10.1098/rspb.2008.0192

471

472 Koslow JA, Goericke R, Lara-Lopez A, Watson W (2011) Impact of declining intermediate-
473 water oxygen on deepwater fishes in the California Current. *Marine Ecology Progress Series*,
474 **436**, 207-218. DOI: 10.3354/meps09270

475

476 Lefort S, Aumont O, Bopp L, Arsouze T, Gehlen M, Maury O (2015) Spatial and body-size
477 dependent response of marine pelagic communities to projected global climate change. *Global*
478 *Change Biology*, **21**, 154–164. DOI: 10.1111/gcb.12679

479

480 Lehodey P, Hampton J, Brill RW, *et al.* (2011) Vulnerability of oceanic fisheries in the tropical
481 Pacific to climate change. In Bell JD, Johnson JE and Hobday AJ (eds) (2011) *Vulnerability of*
482 *Tropical Pacific Fisheries and Aquaculture to Climate Change*. Secretariat of the Pacific
483 Community, Noumea, New Caledonia.

484

485 Lehodey P, Semoma I, Calmettes B, Hampton J, Nicol S (2013) Modeling the impact of climate
486 change on Pacific skipjack tuna population and fisheries. *Climatic Change*, **119**, 95-109. doi:
487 10.1007/s10584-012-0595-1
488
489 Locarnini RA, Mishonov AV, Antonov JJ, *et al.* (2013) World Ocean Atlas 2013, Volume 1:
490 Temperature. Levitus S, Ed., Mishonov A, Technical Ed., NOAA Atlas NESDIS 73, 40 pp.
491
492 Montero-Serra I, Edwards M, Genner MJ (2015) Warming shelf seas drive the subtropicalization
493 of European pelagic fish communities. *Global Change Biology*, **21**, 144–153. DOI:
494 10.1111/gcb.12747
495
496 Ockendon N, Baker DJ, Carr JA, *et al.* (2014) Mechanisms underpinning climatic impacts on
497 natural populations: altered species interactions are more important than direct effects. *Global*
498 *Change Biology*, **20**, 2221–2229, DOI: 10.1111/gcb.12559
499
500 Pinsky ML, Worm B, Fogarty MJ, Sarmiento JL, Levin SA (2013) Marine taxa track local
501 climate velocities. *Science*, **341**, 1239-1242. DOI: 10.1126/science.1239352
502
503 Polovina JJ, Dunne JP, Woodworth PA, Howell EA (2011) Projected expansion of the
504 subtropical biome and contraction of the temperature and equatorial upwelling biomes in the
505 North Pacific under global warming. *ICES Journal of Marine Science*, **68(6)**, 986-995. doi:
506 10.1093/icesjms/fsq198
507
508 Polovina JJ, Howell E, Kobayashi DR, Seki MP (2001) The transition zone chlorophyll front, a
509 dynamic global feature defining migration and forage habitat for marine resources. *Progress in*
510 *Oceanography*, **49**, 469-483.
511
512 Polovina JJ, Howell E, Kobayashi DR, Seki MP (2015) The Transition Zone Chlorophyll Front
513 updated: Advances from a decade of research. *Progress in Oceanography*, In press. DOI:
514 10.1016/j.pocean.20105.01.006
515

516 Polovina JJ, Woodworth-Jefcoats PA (2013) Fishery-induced changes in the subtropical Pacific
517 pelagic ecosystem size structure: observations and theory. PLoS ONE, **8(4)**, e62341. DOI:
518 10.1371/journal.pone.0062341
519
520 Riahi K, Rao S, Krey V, *et al.* (2011) RCP 8.5 – A scenario of comparatively high greenhouse
521 gas emissions. Climatic Change, **109**, 33-57. DOI: 10.1007/s10584-011-0149-y
522
523 Romanou A, Romanski J, Gregg WW (2014) Natural ocean carbon cycle sensitivity to the
524 parameterizations of the recycling in a climate model. Biogeosciences, **11**, 1137-1154. DOI:
525 10.5194/bg-11-1137-2014
526
527 Sarmiento JL, Slater R, Barber R, *et al.* (2004) Response of ocean ecosystems to climate
528 warming. Global Biogeochemical Cycles, **18**, GB3003. doi: 10.1029/2003GB002134
529
530 Scheff J, Frierson DMW (2012) Robust future precipitation declines in CMIP5 largely reflect the
531 poleward expansion of model subtropical dry zones. Geophysical Research Letters, **39**, L18704.
532 DOI: 10.1029/2012GL052910.
533
534 Schmidt GA, Kelley M, Nazarenko L, *et al.* (2014) Configuration and assessment of GISS
535 ModelE2 contributions to the CMIP5 archive. Journal of Advances in Modeling Earth Systems,
536 **6**, 141-184. DOI: 10.1002/2013MS000265
537
538 Sen Gupta A, Jourdain NC, Brown JN, Monselesan D (2013) Climate drift in the CMIP5
539 models. Journal of Climate, **26**, 8597-8615. DOI: 10.1175/JCLIM-D-12-00521.1
540
541 Steinacher M, Joos F, Frölicher TL, *et al.* (2010) Projected 21st century decrease in marine
542 productivity: a multi-model analysis. Biogeosciences, **7**, 979-1005.
543
544 Stock CA, Dunne JP, John JG (2014) Drivers of trophic amplification of ocean productivity
545 trends in a changing climate. Biogeosciences, **11**, 7125-7135. DOI: 10.5194/bg-11-7125-2014
546

547 Storch D, Menzel L, Frickenhaus S, Pörtner H-O (2014) Climate sensitivity across marine
548 domains of life: limits to evolutionary adaptation shape species interactions. *Global Change*
549 *Biology*, **20**, 3059–3067. DOI: 10.1111/gcb.12645
550
551 Taylor KE, Stouffer RJ, Meehl GA (2012) An overview of CMIP5 and the experiment design.
552 *Bulletin of the American Meteorological Society*, **93(4)**, 485-498. DOI: 10.1175/BAMS-D-11-
553 00094.1
554
555 United Nations (UN), 2011: Seven billion and growing: The role of population policy in
556 achieving sustainability. Technical Paper No. 2011/13. 36pp.
557
558 Walsh HJ, Richardson DE, Marancik KE, Hare JA (2015) Long-term changes in the distributions
559 of larval and adult fish in the Northeastern U.S. shelf ecosystem. *PLoS ONE*, **10(9)**, e0137382.
560 DOI: 10.1371/journal.pone.0137382.
561
562 Ward P, Myers RA (2005) Shifts in open-ocean fish communities coinciding with the
563 commencement of commercial fishing. *Ecology*, **86(4)**, 835-847.
564
565 Woodworth-Jefcoats PA, Polovina JJ, Howell EA, Blanchard JL (2015) Two takes on the
566 ecosystem impacts of climate change and fishing: Comparing a size-based and a species-based
567 ecosystem model in the central North Pacific. *Progress in Oceanography*, **138**, 533-545. DOI:
568 10.1016/j.pocean.2015.04.004
569
570 Xu L, Xie S-P, Lui Q (2012) Mode water ventilation and subtropical countercurrent over the
571 North Pacific in CMIP5 simulations and future projections. *Journal of Geophysical Research*,
572 **117**, C12009. DOI: 10.1029/2012JC008377.
573
574 Yongyun H, Lijun T, Jiping L (2012) Poleward expansion of the Hadley circulation in CMIP5
575 simulations. *Advances in Atmospheric Sciences*, **30**, 790-795. DOI: 10.1007/s00376-012-2189-
576 4.
577

578 Yukimoto S, Yoshimura H, Hosaka M, *et al.* (2011) Meteorological Research Institute-Earth
579 System Model Version 1 (MRI-ESM1) Model Description. Technical Report of the
580 Meteorological Research Institute No. 64.
581

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582 **Tables**

583 Table 1: For each model, the SST and epipelagic temperature ranges that decrease in frequency, increase in frequency, and emerge by
 584 the end of the 21st century followed by the zooplankton density ranges that increase and decrease in frequency by the end of the
 585 century.

Model	Sea Surface Temperature (°C)			Epipelagic Temperature (°C)			Zooplankton Density (g C m ⁻²)	
	Decreases	Increases	Emerges	Decreases	Increases	Emerges	Increases	Decreases
Canadian Centre for Climate Modelling and Analysis Earth system model ¹ (CanESM2)	20.0 – 30.0	30.0 – 38.0	34.0 – 38.0	14.5 – 21.5	21.5 – 32.5	30.5 – 32.5	0.05 – 0.20	0.20 – 0.40
NOAA Geophysical Fluid Dynamics Laboratory Earth System Model Generalized ocean layer dynamics ² (GFDL-ESM2G)	20.0 – 29.5	29.5 – 35.0	32.5 – 35.0	13.0 – 18.5	24.5 – 32.0	30.0 – 32.0	0.50 – 0.90	0.90 – 1.85
NOAA Geophysical Fluid Dynamics Laboratory Earth System Model Modular Ocean Model 4 ² (GFDL-ESM2M)	20.0 – 30.0	30.0 – 35.0	33.0 – 35.0	15.0 – 25.5	25.5 – 32.5	30.0 – 32.5	0.40 – 0.95	0.95 – 1.75
NASA Goddard Institute for Space Studies ModelE2 Earth System Model with carbon cycle coupled to the HYCOM ocean model ^{3,4} (GISS-E2-H-CC)	22.0 – 30.0	30.0 – 34.5	32.0 – 34.5	17.5 – 23.0	23.0 – 32.5	31.0 – 32.5	0.00 – 0.10	0.10 – 1.00
NASA Goddard Institute for Space Studies ModelE2 Earth System Model with carbon cycle coupled to the Russell ocean model ^{3,4} (GISS-E2-R-CC)	20.0 – 30.5	30.5 – 34.5	32.5 – 34.5	16.5 – 26.5	26.5 – 33.0	31.5 – 33.0	0.00 – 0.15	0.15 – 0.85
Institut Pierre-Simon Laplace Low Resolution CM5A ⁵ (IPSL-CM5A-LR)	21.0 – 30.0	30.0 – 36.0	32.5 – 36.0	15.5 – 26.0	26.0 – 34.0	30.0 – 34.0	0.30 – 0.45	0.60 – 1.10
Institut Pierre-Simon Laplace Medium resolution CM5A ⁵ (IPSL-CM5A-MR)	21.5 – 31.0	31.0 – 36.5	33.0 – 36.5	15.5 – 26.0	26.0 – 34.0	31.0 – 34.0	0.30 – 0.65	0.65 – 0.95
Institut Pierre-Simon Laplace Low resolution CM5B ⁵	21.5 –	30.0 –	32.0 –	17.5 –	26.5 –	31.0 –	0.25 –	0.35 –

(IPSL-CM5B-LR)	30.0	35.5	35.5	24.0	32.5	32.5	0.35	0.70
Max-Planck-Institute für Meteorologie Earth System Model low resolution ⁶ (MPI-ESM-LR)	20.0 – 30.0	30.0 – 37.0	34.0 – 37.0	14.5 – 22.0	22.0 – 33.5	31.5 – 33.5	0.10 – 0.75	0.75 – 1.50
Max-Planck-Institute für Meteorologie Earth System Model medium resolution ⁶ (MPI-ESM-MR)	20.0 – 30.0	30.0 – 36.5	34.0 – 36.5	16.0 – 21.0	21.0 – 35.0	32.5 – 35.0	0.10 – 0.70	0.70 – 1.55
Meteorological Research Institute Earth System Model Version 1 ⁷ (MRI-ESM1)	21.0 – 29.5	21.0 – 34.5	31.5 – 34.5	16.5 – 27.0	27.0 – 30.5	29.0 – 30.5	0.00 – 0.20	0.20 – 0.40

586 ¹Christian *et al.*, 2010 ²Dunne *et al.*, 2013 ³Romanou *et al.*, 2014 ⁴Schmidt *et al.*, 2014 ⁵Dufresne *et al.*, 2013 ⁶Giorgetta *et al.*,
587 2013 ⁷Yukimoto *et al.*, 2011

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589 **Table Caption**

590 **Table 1** For each model, the SST and epipelagic temperature ranges that decrease in frequency,
591 increase in frequency, and emerge by the end of the 21st century followed by the zooplankton
592 density ranges that increase and decrease in frequency by the end of the century.

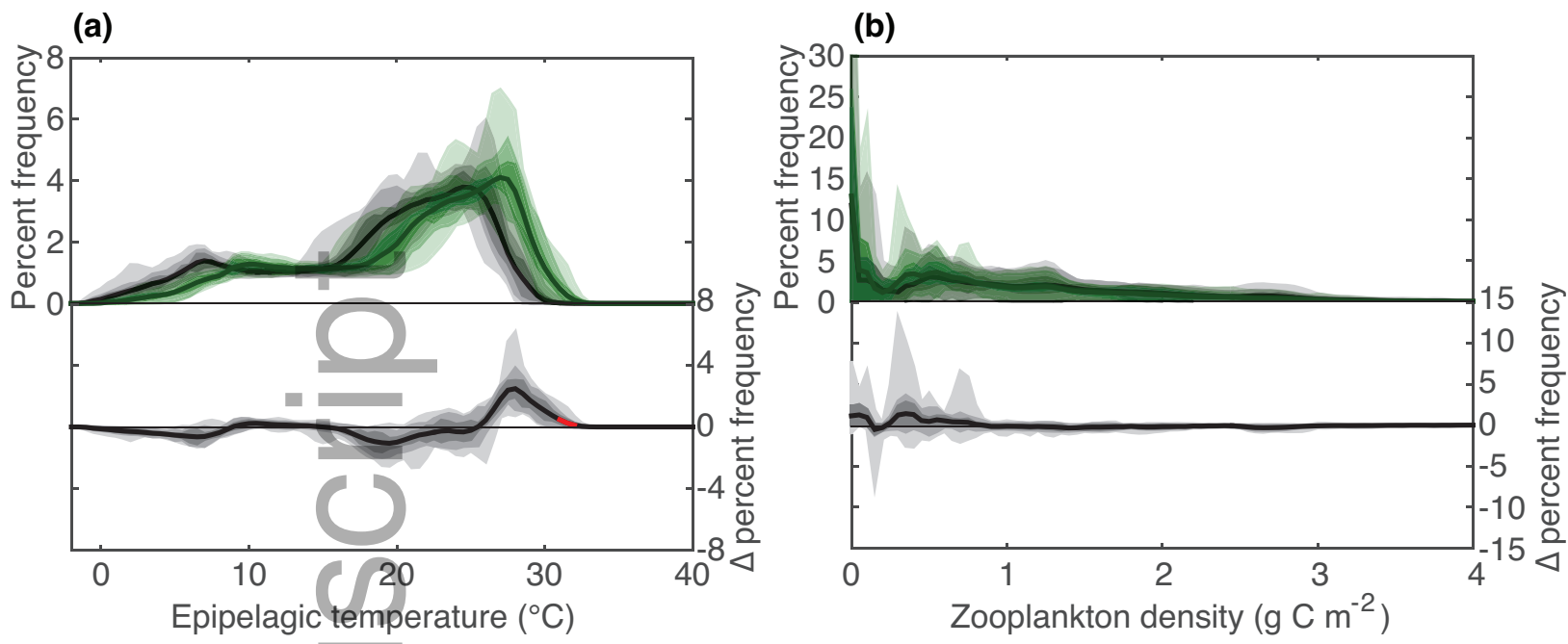
593

594 **Figure Captions**

595 **Fig. 1** Percent frequency distributions of 20-year mean pooled epipelagic temperatures (a) and
596 zooplankton densities (b) from the beginning (grey) and end (green) of the 21st century are
597 plotted above the difference between the two distributions. Solid lines show multi-model means.
598 Terciles encompassing 33%, 67%, and 100% of the models are shaded progressively lighter.
599 The red line in the lower panel of (a) indicates the temperature range over which at least half the
600 models project the emergence of new thermal habitat.

601

602 **Fig. 2** Multi-model median projected change in epipelagic habitat (a – b) and resulting degrees of
603 ecosystem impact (c – f) over the 21st century: projected change in epipelagic temperature (a)
604 and zooplankton density (b), degree of trophic amplification (indicated by warm colors) or the
605 difference between projected phytoplankton and zooplankton percent declines (c), projected
606 change in tuna and billfish species richness over the 21st century for waters within the bounds of
607 a positive solution to equation 1 (5 – 30 °C) (d), projected percent change in potential carrying
608 capacity (e), and the difference in the strength of changing zooplankton density (warm colors)
609 versus changing epipelagic temperature (cool colors) as drivers of change in potential carrying
610 capacity (f). In (a – e) stippling indicates areas where at least 80% of the models used project a
611 change of the same sign. In (f) stippling indicates areas where at least 80% of the models used
612 indicate the same dominant driver.



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