

If you build it, they will come: coastal amenities facilitate human engagement in marine protected areas

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Abstract:	1. Calls for using marine protected areas (MPAs) to achieve goals for nature and people are increasing globally. While the conservation and fisheries impacts of MPAs have been comparatively well studied, impacts on other dimensions of human use have received less attention. Understanding how humans engage with MPAs and identifying traits of MPAs that promote engagement is critical to designing MPA networks that achieve multiple goals effectively, equitably, and with minimal environmental impact. 2. In this paper, we characterize human engagement in California's MPA network, the world's largest MPA network scientifically designed to function as a coherent network (124 MPAs spanning 16% of state waters and 1,700 km of coastline), and identify traits associated with higher

	<p>human engagement. We assemble and compare diverse indicators of human engagement that capture recreational, educational, and scientific activities across California's MPAs.</p> <p>3. We find that human engagement is correlated with nearby population density and that site "charisma" can expand human engagement beyond what would be predicted based on population density alone. Charismatic MPAs tend to be located near tourist destinations, have long sandy beaches, and be adjacent to state parks and associated amenities. In contrast, underutilized MPAs were often more remote and lacked both sandy beaches and parking lot access.</p> <p>4. Synthesis and applications: These results suggest that achieving MPA goals associated with human engagement can be promoted by developing land-based amenities that increase access to coastal MPAs or by locating new MPAs near existing amenities during the design phase. Alternatively, human engagement can be limited by locating MPAs in areas far from population centers, coastal amenities, or sandy beaches. Furthermore, managers may want to prioritize monitoring, enforcement, education, and outreach programs in MPAs with traits that predict high human engagement. Understanding the extent to which human engagement impacts the conservation performance of MPAs is a critical next step to designing MPAs that minimize tradeoffs among potentially competing objectives.</p>
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1 If you build it, they will come: coastal amenities 2 facilitate human engagement in marine protected areas

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30 Abstract

- 31 1. Calls for using marine protected areas (MPAs) to achieve goals for nature and people
32 are increasing globally. While the conservation and fisheries impacts of MPAs have
33 been comparatively well studied, impacts on other dimensions of human use have
34 received less attention. Understanding how humans engage with MPAs and identifying
35 traits of MPAs that promote engagement is critical to designing MPA networks that
36 achieve multiple goals effectively, equitably, and with minimal environmental impact.
 - 37 2. In this paper, we characterize human engagement in California's MPA network, the
38 world's largest MPA network scientifically designed to function as a coherent network
39 (124 MPAs spanning 16% of state waters and 1,700 km of coastline), and identify traits
40 associated with higher human engagement. We assemble and compare diverse
41 indicators of human engagement that capture recreational, educational, and scientific
42 activities across California's MPAs.
 - 43 3. We find that human engagement is correlated with nearby population density and that
44 site "charisma" can expand human engagement beyond what would be predicted based
45 on population density alone. Charismatic MPAs tend to be located near tourist
46 destinations, have long sandy beaches, and be adjacent to state parks and associated
47 amenities. In contrast, underutilized MPAs were often more remote and lacked both
48 sandy beaches and parking lot access.
 - 49 4. *Synthesis and applications:* These results suggest that achieving MPA goals associated
50 with human engagement can be promoted by developing land-based amenities that
51 increase access to coastal MPAs or by locating new MPAs near existing amenities
52 during the design phase. Alternatively, human engagement can be limited by locating
53 MPAs in areas far from population centers, coastal amenities, or sandy beaches.
54 Furthermore, managers may want to prioritize monitoring, enforcement, education, and
55 outreach programs in MPAs with traits that predict high human engagement.
56 Understanding the extent to which human engagement impacts the conservation
57 performance of MPAs is a critical next step to designing MPAs that minimize tradeoffs
58 among potentially competing objectives.
- 59
- 60 **Keywords:** California, citizen science, community engagement, human dimensions, human
61 use, marine protected areas, recreation, tourism

62 1. Introduction

63 Marine protected areas (MPAs) — places where human activity, especially extractive
64 practices such as fishing, is prohibited or restricted — are a common ocean management tool
65 used to achieve a mixture of conservation, fisheries, and cultural objectives (Erskine et al.,
66 2021; Grorud-Colvert et al., 2021; Marcos et al., 2021). By restricting extractive and destructive
67 human activities, adequately designed, funded, and regulated MPAs can increase the diversity
68 and abundance of marine fish and invertebrates (Edgar et al., 2014; Gill et al., 2017; Goetze et
69 al., 2021; Zupan et al., 2018) and the function and resilience of marine ecosystems (Cheng et
70 al., 2019; Mellin et al., 2016). In the long term, and with concerted community participation and
71 buy-in, well-designed MPAs can also yield fisheries benefits through increased productivity and
72 spillover resulting from improved biomass and age structure of populations in the MPA (Di
73 Lorenzo et al., 2020; Marshall et al., 2019). Furthermore, MPAs can facilitate and enhance other
74 non-extractive human engagement in ocean ecosystems, such as cultural activities, recreation
75 and tourism, education and outreach, and scientific research (Angulo-Valdés & Hatcher, 2010;
76 Ban et al., 2019; Erskine et al., 2021; Roncin et al., 2008).

77

78 While the ability and prerequisites for MPAs to achieve conservation and fisheries
79 objectives have been comparatively well-studied (e.g., Claudet et al., 2008; Edgar et al., 2014;
80 Giakoumi et al., 2017; Goñi et al., 2010; Lester & Halpern, 2008; Wilson et al., 2020), the
81 enabling conditions for achieving other human use objectives has received less attention (Ban
82 et al., 2019; Erskine et al., 2021; Gerber et al., 2003; Naidoo et al., 2019; Turnbull et al., 2021).
83 This is surprising given the frequency with which human engagement objectives — such as
84 recreation, education, and scientific research — are identified in international, national, and
85 regional MPA planning documents. For example, the Independent World Commission on the
86 Oceans identifies the “*provision of areas for scientific research, education, and recreation*” as a
87 key benefit of MPAs (IWCO, 1998). Similarly, the U.S. Framework for the National System of
88 Marine Protected Areas identifies the benefits of U.S. MPAs as: (1) “*supporting social and*
89 *economic benefits [including] coastal tourism*”, (2) “*providing new educational opportunities*”,
90 and (3) “*enhancing research opportunities*” (NOAA, 2015). In some cases, MPAs may aim to
91 enhance cultural, spiritual, emotional, or intrinsic value benefits derived from the ocean (Allison
92 et al., 2020). Evaluating human engagement in MPAs is needed to track progress towards

93 achieving these objectives and for identifying the design principles that determine human
94 engagement in MPAs. Here, we use California's MPA network, the world's largest MPA network
95 scientifically designed to function as a coherent network (Botsford et al., 2014), as a case study
96 for identifying conditions that promote or limit human engagement in MPAs.

97

98 In 1999, the California state legislature passed the Marine Life Protection Act (MLPA),
99 which directed the state to use the best available science to redesign and greatly expand its
100 system of MPAs to function as a coherent network and to address six goals in service of
101 conservation, fisheries, and other cultural objectives (Marine Life Protection Act, 1999; Gleason
102 et al., 2013). In addition to goals to preserve biodiversity and ecosystem function and to sustain,
103 conserve, protect, and rebuild marine populations, including those of economic value, the MLPA
104 also included a goal to *"improve recreational, educational, and study opportunities provided by
105 marine ecosystems that are subject to minimal human disturbance, and to manage these uses
106 in a manner consistent with protecting biodiversity."* From 2004 to 2012, a community-driven
107 and science-guided design process led to a coordinated network of 124 MPAs, containing 16%
108 of state waters, along California's 1,770 km (1,100 miles) coastline. Following implementation,
109 an extensive monitoring effort began to ensure that the network could undergo adaptive
110 management (Botsford et al., 2014). While some monitoring programs were developed around
111 human engagement in MPAs (e.g., the MPA Watch citizen science program; MPA Watch,
112 2022b), the majority of the monitoring effort was focused on the ecological goals of the MLPA
113 and on elucidating ecological responses to MPA implementation.

114

115 Here, we characterize human engagement in California's MPA network and identify traits
116 associated with high engagement. We assemble and evaluate diverse indicators of engagement
117 that capture a range of recreational, educational, and scientific activities. We then relate levels
118 of human engagement to population density, accessibility, amenities, and other traits likely to
119 influence engagement. This provides a rare quantification of the ways in which people engage
120 with MPAs and the potential pathways for enhancing or limiting engagement based on
121 management goals. These insights are helpful as California (Executive Order N-82-20, 2020),
122 the United States (Executive Order on Tackling the Climate Crisis at Home and Abroad, 2021),
123 and the world (CBD, 2021) aim to protect 30% of the ocean by 2030 (30x30) to meet an array of
124 conservation, fisheries, and other cultural objectives (Sullivan-Stack et al., 2022).

125 2. Methods

126 2.1 Marine protected areas

127 California's coastal waters are protected by a mosaic of spatial management areas that
128 vary in regulatory authority and protection status (**Fig. 1A; Table S1**). State-managed areas
129 include: (1) state marine reserves (SMRs), which prohibit all fishing; (2) state marine
130 conservation areas (SMCAs), which restrict some types of fishing, except for within special no-
131 take SMCAs, which prohibit all fishing; (3) state marine recreational managed areas (SMRMAs),
132 which restrict fishing and allow hunting of waterfowl; (4) state marine parks (SMPs), which
133 prohibit commercial fishing; and (5) special closures, which restrict activity around seabird
134 colonies and marine mammal haulouts and are the only designation not defined as an MPA by
135 the MLPA (**Table S1**). Federal marine reserves and conservation areas (FMRs and FMCAs,
136 respectively) extend certain SMRs and SMCAs around the Channel Islands into federal waters
137 (**Fig. 1A**).

138

139 We focus on the 124 MPAs that the MLPA identifies as being part of California's state-
140 managed coastal MPA network (**Fig. 1A; Table S1**). This excludes federally managed MPAs
141 around the Channel Islands; SMRAs and SMPs in San Francisco Bay, which were established
142 before the MLPA planning process and are not coastal; and special closures, which are not
143 identified as MPAs by the MLPA. We refer to the resulting network of 49 SMRs, 60 SMCAs, 10
144 no-take SMCAs, and 5 SMRMAs as California's state MPA network. While the Channel Islands
145 MPAs were established before the MLPA planning process, they have been legally incorporated
146 into the network. The four MLPA regions (South, Central, North Central, and North Coasts; **Fig.**
147 **1**) encompass a wide range of ecological dynamics, coastal features, oceanographic
148 environments, cultures, and economies.

149 2.2 Surrounding human communities

150 We hypothesized that the number of people living near an MPA and the socioeconomic
151 vulnerability of this population would contribute to engagement levels. In short, we expected that
152 MPAs with larger and less vulnerable nearby human populations (i.e., populations with more
153 disposable income and time for recreation) would experience greater human engagement. We
154 characterized the human population living near MPAs using population demographics data from
155 the 2010 U.S. Decennial Census (USCB, 2010a). The 2010 data is the most recent available

156 data given extended delays in the release of the 2020 U.S. Census data (Schneider, 2023). We
157 downloaded total population estimates by census block, the smallest geographic unit used in
158 the census, using the *tidycensus* R package (Walker et al., 2022) and calculated the density of
159 people living within each block. We rasterized (500x500 m resolution) these data and calculated
160 the number of people living within a 50 km radius (~31 miles) of each MPA (**Fig. 1B**). The
161 number of people living within 50 km is generally ($r^2 > 0.8$) correlated with population densities
162 using buffer distances ranging from 10 to 100 km (~6-60 miles) (**Fig. S1**).

163

164 We estimated the social vulnerability of these populations using twelve indicators
165 identified by (Jepson & Colburn, 2013) and collected by the U.S. Census American Community
166 Survey (USCB, 2010b). These indicators describe various metrics of poverty status, housing
167 characteristics, labor force structure, and population composition (**Table S2; Figures S2-S4**).
168 We downloaded these indicators by census tract, the smallest geographic unit for which all of
169 the indicators were available (one level larger than census block), also using the *tidycensus* R
170 package (Walker et al., 2022). We combined these indicators into a single vulnerability index by
171 averaging the z-scores of each indicator (i.e., indicators were centered on the statewide
172 average and scaled to unit variance). Thus, a value of zero indicates average vulnerability
173 across all of the various indicators, negative values indicate higher than average vulnerability,
174 and positive values indicate lower than average vulnerability. We rasterized the tract-level index
175 to match the population raster and calculated the average vulnerability of the population within
176 50 km of each MPA as the population-weighted average of the social vulnerability index.

177 2.3 Human engagement in protected areas

178 We developed indicators of human engagement in recreational, educational, and
179 scientific activities in California's state MPA network using a mixture of citizen science,
180 naturalist, and state agency datasets (**Table S3**). We focused on recreational, educational, and
181 scientific engagement given that they are specific objectives of the network (Marine Life
182 Protection Act, 1999) and given the lack of data on other cultural, spiritual, or emotional types of
183 human engagement. We used data from two citizen science programs (MPA Watch and Reef
184 Environmental Education Foundation) and two naturalist social networks (iNaturalist and eBird),
185 which provide spatially referenced records of activities (e.g., surfing, swimming, boating,
186 tidepooling, diving, etc.) or observations of wildlife submitted by individual users, as indicators of
187 recreational and educational engagement in MPAs. While popular social media platforms such

188 as Instagram, Facebook, Flickr, and Twitter may provide a better indicator of visitation rates
189 than specialist platforms such as iNaturalist and eBird (Tenkanen et al., 2017), the volume of
190 data generated by these platforms requires careful subsampling to be manageable (e.g.,
191 Hausmann et al., 2017). Although analysis of these social media indicators of engagement was
192 outside the scope of this study, we encourage their use in future research. We used data from
193 the California Department of Fish and Wildlife (CDFW) on the annual numbers of permits issued
194 for scientific research in California's MPAs as an indicator of scientific engagement. Finally, we
195 used CDFW data on regulatory citations as an indicator of regulatory compliance within the
196 network.

197

198 We used MPA Watch survey data to measure consumptive and non-consumptive
199 human activities in California's MPA network. MPA Watch is a citizen science program that
200 trains volunteers to observe and collect data on human engagement in protected areas (MPA
201 Watch, 2022b). Volunteers use a standardized survey protocol (MPA Watch, 2022a) to record
202 consumptive (e.g., fishing) and non-consumptive (e.g., surfing, boating, tidepooling, running,
203 etc.) activities occurring both on- and off-shore of coastal sampling sites (**Table S4**).
204 Consumptive activities are classified as either active (e.g., fishing line in water) or inactive (e.g.,
205 fishing pole on boat but not being used); we focus on active consumptive activities. We caution
206 that SMRMA and some SMCA allow some forms of harvest and that MPA Watch volunteers,
207 while well trained, are not legal authorities on MPA boundaries and regulations. Thus, our ability
208 to infer the legality of consumptive activities documented by MPA Watch volunteers is limited.
209 MPA Watch has been in operation since 2011 and, as of writing, has conducted over 33,000
210 surveys in 49 MPAs (47 of which meet our inclusion criteria) and 60 control (non-MPA) locations
211 (**Fig. S5**). While some MPAs have been surveyed consistently since 2011, others did not
212 receive consistent visits until 2015 or later (**Fig. S5A**). To allow comparison between sites with
213 variable temporal coverage, we limited analysis to surveys that took place from January 1, 2015
214 to December 31, 2021. To eliminate spurious results from surveys that were conducted either
215 early in the morning or late at night or were either shorter or longer than the official protocol
216 (MPA Watch, 2022a), we also limited analysis to surveys that occurred between 6AM and 8PM
217 and lasted between 10 and 60 minutes (**Fig. S5BC**). We quantified human engagement by MPA
218 in terms of (1) the percent of surveys in which an activity was observed and (2) the median
219 number of activities observed per hour for surveys in which activities were observed (zeroes
220 excluded because of high zero-inflation) (**Figs. S6 & S7**).
221

222 We used iNaturalist submission records to measure engagement in wildlife observation
223 within and adjacent to MPAs. iNaturalist is a web- and app-based platform that allows observers
224 to submit wildlife photos for identification by amateur and professional naturalists (iNaturalist,
225 2022). iNaturalist was launched in 2008 and as of writing, has more than 100 million
226 observations, 2 million observers, and 380,000 observed species globally. We used the *rnat* R
227 package (Barve et al., 2021) to download all iNaturalist observations submitted by users in a
228 bounding box spanning the California coastline from January 1, 2000 to December 31, 2021
229 (iNaturalist allows back submissions, hence the availability of pre-2008 observations). We
230 defined MPA-associated observations as observations occurring within 100 meters of an MPA
231 and quantified human engagement from 2012 through 2021 by MPA in terms of the number of
232 (1) unique observers (number of iNaturalist users who submitted wildlife observations) and (2)
233 observations (number of entries submitted). More than 5,800 observers have submitted >72,000
234 observations associated with 121 of California's state MPAs (**Figs. S8 & S9**).
235

236 We used eBird submission records to measure engagement in birding within and
237 adjacent to MPAs. eBird is a global program that collates observations of birds submitted by
238 birdwatchers (eBird, 2022). It was launched in 2002 by the Cornell University Lab of Ornithology
239 and the National Audubon Society but allows back submissions from birding diaries. As a result,
240 eBird contains observations dating back centuries in many locations. As of writing, the eBird
241 includes over 69.7 million submissions from nearly 800,000 birders. We downloaded eBird
242 observations from California and, as with the iNaturalist data, identified observations occurring
243 within 100 meters of an MPA from 2012 through 2021. We quantified human engagement by
244 MPA in terms of the number of (1) unique observers and (2) observations. More than 19,000
245 birders have conducted >193,000 surveys and made >3.8 million submissions to eBird
246 associated with 114 of California's state MPAs (**Figs. S10 & S11**).
247

248 We used Reef Environmental Education Foundation (REEF) diver surveys as an
249 indicator of engagement in diving and snorkeling in California's MPAs. REEF is an international
250 marine conservation organization that trains volunteer SCUBA divers and snorkelers to collect
251 and report information on marine fish and select invertebrate and algae species during
252 recreational SCUBA dives and snorkels (REEF, 2022). The diver survey program was launched
253 in 1993 and, as of writing, has >250,000 surveys by 16,000 volunteers at 15,000 sites
254 worldwide. We received records of >14,700 surveys conducted in California and identified 4,085
255 surveys occurring within 41 of California's state MPAs from 2012 through 2021 (**Figs. S12 &**

256 **S13).** We quantified human engagement by MPA in terms of the (1) number of surveys
257 conducted and (2) number of years in which a survey was conducted.

258

259 We used records of scientific permits issued by CDFW for research conducted within
260 California's MPA network as an indicator of the contributions of MPAs to scientific knowledge.
261 While permits are required for any extractive or manipulative research in California's coastal
262 waters, purely observational research (i.e., research without capturing, handling, etc.) does not
263 require permits; thus, the permit data may underestimate the amount of research occurring in
264 the network. From 2012-2021, 5,329 scientific permits were issued for research in all 124 of
265 California's state MPAs (**Figs. S14 & S15**). We quantified human engagement by MPA in terms
266 of the (1) number of permits issued and (2) number of years in which permits were issued.

267

268 We used records of citations issued by the CDFW Law Enforcement Division for
269 regulatory violations occurring within California's MPA network as an indicator of compliance.
270 From 2016-2021, 2,812 citations were issued for violations occurring within 85 of California's
271 state MPAs (**Figs. S16 & S17**). We quantified non-compliance by MPA in terms of the (1)
272 number of citations issued and (2) number of years in which citations were issued. We used
273 generalized linear models assuming a Poisson distribution to evaluate the correlation between
274 the total number of citations issued within an MPA and human population density, human
275 engagement (defined using the iNaturalist observer data), and observations of active fishing
276 (defined using the MPA Watch survey data). We caution that the lack of patrol effort information
277 limits our ability to infer non-compliance rates (i.e., whether more citations corresponds to more
278 effort or more illegal activity) and advise that, going forward, CDFW record information on effort
279 (e.g., number of patrol hours) to improve ability to document patterns of non-compliance and
280 target patrol strategies.

281

282 To compare human engagement across indicators (**Fig. 2**), we selected key metrics for
283 each indicator (**Table S3**) to display in an engagement scorecard (**Fig. 3**). We centered each
284 metric on its mean and scaled it to unit variance to facilitate comparisons across indicators. We
285 also measured and compared the degree to which engagement is concentrated within specific
286 MPAs, a metric of the selectivity of users, by developing the engagement accumulation curves
287 shown in **Figure 4**. We developed these curves by first calculating the percent contribution of
288 each MPA to network-wide engagement for each of the metrics selected for the scorecard. We
289 then plotted the accumulation of these contributions beginning with the MPA with the highest

290 engagement and ending with the MPA with the lowest engagement. The steeper the resulting
291 curve, the more network-wide engagement is dominated by a few MPAs.

292 2.4 Drivers of human engagement

293 We hypothesized that human engagement in MPAs would be correlated with nearby
294 population density (Cinner et al., 2018; Ravenstein, 1885) except for (1) “charismatic” MPAs
295 that draw participation from afar and thus generate more engagement than would be predicted
296 based on nearby population density, and (2) “underutilized” MPAs that are difficult to access
297 (e.g., located offshore, limited road access, etc.) and thus generate less engagement than
298 would be predicted based on nearby population density. To distinguish charismatic and
299 underutilized MPAs, we regressed human engagement (as measured by the number of
300 iNaturalist observers) against population density and extracted the MPAs that fell above
301 (charismatic) or below (underutilized) 75% of the fitted values (**Fig. 5**). For this model, we used
302 the number of iNaturalist observers as our measure of human engagement because it was the
303 most spatially comprehensive indicator (i.e., describes engagement in the greatest number of
304 MPAs) and it correlates with all of the indicators of non-extractive engagement (i.e., it is not
305 correlated with citations or consumptive activities; **Fig. S18**).

306

307 We used logistic regression to identify traits associated with charismatic and
308 underutilized MPAs (**Fig. 6**). We considered 13 traits describing a range of MPA design features
309 (age, size, protection level), habitats (sandy beach, rocky intertidal, kelp, estuary), accessibility
310 and amenities (distance to port; number of parks, parking lots, campgrounds, and picnic areas
311 within 1 km), and the social vulnerability index. See **Table S5** for the source of each explanatory
312 variable. We then used a series of logistic regressions to evaluate the association between
313 engagement (charismatic vs. typical and underutilized vs. typical) and these traits. We defined
314 the logistic target level for each model based on “typical” MPAs (response of 0) versus
315 charismatic or underutilized (response of 1). Logistic models were constructed stepwise after a
316 *priori* identifying relevant drivers of engagement. The best fitting models were selected using
317 Akaike Information Criterion (AIC) (Akaike, 1974) to identify the most parsimonious model of the
318 relationship between engagement and the evaluated traits.

319 2.5 Comparison to non-MPA areas

320 The methods described above were used to determine which MPAs within California's
321 MPA network generate the most human engagement and to identify the factors that drive
322 differences in the levels of engagement; however, they are unable to reveal whether MPAs
323 generate more, less, or equivalent human engagement as similar non-MPA areas. To
324 understand the degree to which MPA designations impact human engagement in coastal areas,
325 we compared engagement in MPA areas to similar counterfactual non-MPA areas. We identified
326 similar counterfactual areas through statistical matching (Ferraro, 2009), which is being
327 increasingly used to elucidate the ecological impacts of MPAs (Ahmadi et al., 2015; Gill et al.,
328 2017). In short, we rasterized California's state waters into 200 m raster cells and paired each
329 MPA cell with a non-MPA counterfactual cell with otherwise similar properties. We identified
330 non-MPA counterfactual cells that were similar to their MPA reference cells in their depth (m),
331 distance from shore (km), nearby population density, proximity to parks, and proximity to public
332 beaches. These matching variables were selected based on their association with engagement
333 based on theory (Cinner et al., 2018; Ravenstein, 1885) and as revealed through the regression
334 analysis (**Fig. 6**). Ideally, we would also match based on pre-MPA visitation rates (Devillers et
335 al., 2015), but the lack of sufficient pre-MPA visitation data (see limited pre-2007 data in **Figure**
336 **S8**) precluded this gold standard. However, by controlling for these known and quantifiable
337 drivers of MPA site selection and human engagement, we can isolate, to the greatest extent
338 practicable, the impact of MPA designation on human engagement. We derived these values for
339 both MPA and counterfactual cells using the sources listed in **Table S6**. We identified suitable
340 counterfactuals through statistical matching using the *MatchIt* package (Ho et al., 2011), using
341 one-to-one Mahalanobis distance matching with replacement and propensity score calipers of
342 0.20 standard deviations (Ho et al., 2007). After an appropriate counterfactual was identified for
343 each MPA cell (**Fig. S19**), we calculated the log-response ratio of the sum of activities within
344 each MPA's cells and its paired counterfactuals cells for the three engagement indicators with
345 activities reported inside and outside MPAs using GPS coordinates (i.e., the iNaturalist, eBird,
346 and REEF indicators). We tested whether the mean log-ratio of these sums differed from zero
347 using t-tests (i.e., whether MPAs and non-MPAs generate different levels of human
348 engagement). Log-response ratios were calculated after adding 1 to the engagement values
349 occurring in both the numerator and denominator to avoid non-finite ratio values.

350

351 All data analysis and visualization was done in R (R Core Team, 2021) and all data and
352 code are available on GitHub here: <https://github.com/NCEAS/ca-mpa>

353 3. Results

354 3.1 Human engagement in protected areas

355 MPA Watch volunteers observed non-consumptive activities in the vast majority of
356 surveys conducted coastwide and within all of the 47 surveyed MPAs (**Fig. 2A**). MPA visitors
357 were most commonly observed walking and recreating on the beach, often with their pets.
358 Offshore recreation included boating, surfing, bodyboarding, and swimming. MPA visitors were
359 also often observed viewing wildlife and exploring tidepools (**Fig. S6BC**). MPAs in the South
360 Coast region were most popular, especially those near the metropolitan areas of San Diego and
361 Los Angeles (**Fig. 2A**).

362

363 MPA Watch volunteers observed active consumptive activities (i.e., fishing and hand
364 collection of organisms) in all but four of the 47 surveyed MPAs (**Fig. 2B**) but at rates
365 substantially lower than those observed for non-consumptive activities (**Fig. S7BC**). Hook and
366 line fishing was the most commonly observed consumptive activity and was observed in ~6% of
367 surveys within SMCAs (MPAs in which certain types of fishing are often allowed). However,
368 active hook and line fishing was also reported by volunteers in surveys in no-take SMCAs
369 (~1.8% of surveys) and SMRs (~2% of surveys) (**Fig. S7B**). Hand collection of organisms, trap
370 fishing, and spear fishing were the next most frequently reported consumptive activities. Net
371 fishing, dive fishing, commercial passenger fishing vessel (CPFV) fishing, and kelp harvest were
372 more rarely reported (**Fig. S7BC**). Observations of consumptive activities were more frequent in
373 South Coast MPAs and within SMCAs, which allow some types of harvest.

374

375 The number of people submitting wildlife observations to iNaturalist from within
376 California's MPA network increased through time (**Fig. S8BC**). The majority of observers submit
377 observations from only one MPA per year, but some observers make submissions from up to 21
378 MPAs per year (**Fig. S8C**). Observers are especially interested in plants (often land-based),
379 shells (mollusks), and seabirds (**Fig. S8B**). iNaturalist participation is especially high in the
380 touristic Monterey Bay area and secondarily high in the densely populated San Diego, Los
381 Angeles, and San Francisco areas (**Fig. 2C**). MPA engagement was less selective than
382 predicted by human population density for this form of human engagement (**Fig. 4**). On
383 average, California's MPAs have not generated more iNaturalist engagement than

384 counterfactual sites ($p=0.12$), indicating that non-MPA areas with similar features generate just
385 as much engagement as MPAs for this type of activity (**Fig. 7**).
386

387 Birders have been visiting California's MPAs since before they were designated as
388 protected areas (**Fig. S11BC**). The participation of birders in the eBird citizen science program
389 increased linearly from the 1960-2005 and exponentially since 2005 (**Fig. S11B**). Participation
390 has been greatest, in terms of number of birders submitting eBird observations, at popular
391 birding hotspots such as Bolsa Chica Basin SMCA, Elkhorn Slough SMR, Matlahuayl SMR,
392 Morro Bay SMRMA, and Point Reyes SMR (**Figs. 2D & 3**). MPAs within estuaries -- including
393 Bolsa Chica Basin, Elkhorn Slough, and Morro Bay -- generate a disproportionate amount of
394 eBird activity: despite representing only 2% of California's state MPA network by area (17% by
395 count), around 40% of recent annual visits to the network logged by eBirders have been within
396 estuarine MPAs (**Fig. S11C**). Despite the tendency for eBirders to visit estuarine MPAs, the
397 selectivity of birders was generally proportional to that predicted by population density (**Fig. 4**),
398 suggesting that estuarine MPAs are located in areas with high population density. On average,
399 California's MPAs have generated slightly more eBird engagement than counterfactual sites
400 ($p=0.02$), indicating that MPA status attracts engagement for this type of activity (**Fig. 7**).
401

402 The number of recreational divers and snorkelers contributing to the REEF citizen
403 science survey program from within California's MPA network increased from the program's
404 inception in 1994 to a peak in 2011, then decreased until a resurgence during the COVID-19
405 pandemic (2020-2021) (**Fig. S13BC**). Participants visited a range of habitats and depths but
406 generally favored kelp forests and rocky reefs (**Fig. S13BC**). The majority of participation has
407 come from MPAs with high profile dive sites including, in decreasing order of prevalence,
408 Matlahuayl SMR, Edward F. Ricketts SMCA, Point Lobos SMR, Pacific Grove Marine Gardens
409 SMCA, and Carmel Bay SMCA (**Figs. 2E & 3**). REEF divers have been more selective in their
410 MPA visitation than any of the other evaluated user groups (**Fig. 4**). California's MPAs have, on
411 average, generated much more REEF survey engagement than counterfactual sites ($p<0.0001$),
412 indicating that MPA status attracts engagement for this type of activity (**Fig. 7**).
413

414 The number of scientific permits issued for research within California's MPA network has
415 been variable through time and decreased during the COVID-19 pandemic (2020-2021) (**Fig.**
416 **S15B**). The distribution of scientific research throughout the MPA network has been more even
417 than other types of human engagement (**Fig. 4**). In general, fewer permits have been issued for

418 research in the North and North Central Coast regions and more permits have been issued for
419 research in the Central (especially Monterey Bay) and South (especially Los Angeles and San
420 Diego) Coast regions (**Figs. 2F & 3**), where academic institutions and marine science non-
421 profits are more highly concentrated. Scientific research in MPAs of different designations has
422 generally occurred in proportion to the representation of the different MPA designations within
423 the network (i.e., no bias towards no-take areas) (**Fig. S15C**).

424

425 The number of citations issued for regulatory violations was highest in MPAs in the
426 South Coast region, especially in the MPAs around Catalina Island, a major tourist destination
427 off the coast of Los Angeles (**Fig. S17A**). In general, the number of citations is positively
428 correlated with nearby human population size ($p<0.001$; **Fig. S17B**) and human engagement
429 ($p<0.001$; **Fig. S17C**) in MPAs, where engagement is defined as the total number of people
430 contributing iNaturalist observations from within an MPA from 2012-2021. Interestingly, the
431 number of citations was negatively correlated with the observation of active consumptive activity
432 by MPA Watch observers (**Fig. S17D**), which could indicate that the active consumptive activity
433 reported by MPA Watch observers is sanctioned or that active consumptive activity is more
434 prominent in areas with less active enforcement. Citations were more highly concentrated in
435 certain MPAs than would be predicted by human population density alone (**Fig. 4**).

436 3.2 Drivers of human engagement

437 Across all indicators, human engagement in MPAs was highest in the populous South
438 Coast region and the touristic Monterey Bay area in the Central Coast region, and lowest in the
439 remote North Coast region (**Figs. 2 & 3**). We found that human engagement in MPAs was
440 correlated to nearby population density ($r^2=0.14$; $p<0.001$) but that MPA traits can enhance or
441 reduce engagement beyond what would be predicted based on population density alone (**Fig.**
442 **5**). Elevated engagement in 20 “charismatic” MPAs (MPAs whose engagement is greater than
443 would be expected based on population density) was associated with older MPAs with long
444 sandy beaches and many adjacent land-based parks (**Fig. 6; Table S7**). Reduced engagement
445 in 42 “underutilized” MPAs (MPAs whose engagement is lower than would be expected based
446 on population density) was associated with remoteness (i.e., far from the nearest port), lack of
447 sandy beaches, and lack of parking lot access (**Fig. 6; Table S7**). Counter to our hypothesis,
448 social vulnerability was not a significant driver of human engagement in MPAs (**Table S7**).

449 4. Discussion

450 Understanding the ability and prerequisites for MPAs to achieve human use objectives is
451 central to designing MPA networks that provide multiple benefits to people and nature.
452 California's MPA network supports a diverse array of recreational, educational, and scientific
453 activities. MPAs are commonly used for recreational activities such as walking, playing, or
454 relaxing on the beach or boating, surfing, swimming, or SCUBA diving in the ocean.
455 Engagement in these activities makes important contributions to local economies (Pendleton &
456 Kildow, 2006) and to cultural, emotional, and physical health (Hipp & Ogunseitan, 2011;
457 Jacobson, 2020). Wildlife viewing is also common within California's MPAs and provides a
458 platform for education and research. Many visitors engage in MPAs through citizen science
459 programs that provide opportunities both to learn about the natural world and to contribute to
460 meaningful scientific datasets (Freiwald et al., 2018; Rapacciulo et al., 2021). Finally, scientific
461 researchers have utilized the MPA network as a "large-scale ecological experiment" (*sensu*
462 Jensen et al., 2012) to derive globally-relevant insights into MPA performance, marine ecology,
463 and fisheries and conservation science (e.g., Starr et al., 2015; White et al., 2021; Ziegler et al.,
464 2022).

465

466 However, not all MPAs generate equal levels of human engagement. In general,
467 engagement is positively correlated with surrounding human population density: the more
468 people living near an MPA, the more engagement an MPA generates. Charismatic MPAs,
469 MPAs that receive more engagement than would be expected based on nearby population
470 density, likely draw additional users because they have adjacent land-based attractions (i.e.,
471 parks) and associated amenities (e.g., parking lots, restrooms, campgrounds). These MPAs
472 also have higher amounts of sandy beaches, which based on the MPA Watch surveys, tend to
473 generate higher engagement than rocky beaches. Furthermore, many of the charismatic MPAs
474 are located in areas spanning the Monterey Bay and Big Sur coastlines and the city of San
475 Diego, which attract high numbers of tourists. These results are consistent with studies of land-
476 based protected areas that find that visitation rates are driven primarily by the availability of
477 amenities such as parking lots, walking paths, and campgrounds and the accessibility of parks
478 to human populations (see (Heagney et al., 2018) and references within). Finally, engagement
479 is moderated by the selectivity of different user groups. For example, whereas divers are highly
480 selective in their choice of MPAs to visit, scientists have conducted research much more evenly

481 across the statewide MPA network. Birders disproportionately visit estuarine MPAs, which tend
482 to harbor large bird populations due to their high productivity (Paracuellos & Tellería, 2004).

483

484 It is also critical to understand patterns of unsanctioned use within California's MPA
485 network. Overall, consumptive use was observed in a higher proportion of surveys conducted in
486 MPAs that allow some types of harvest (i.e., SMCAs and SMRMs) than in fully no-take MPAs
487 that prohibit all fishing (i.e., SMRs and no-take SMCAs). However, MPA Watch surveys, which
488 we caution are conducted by citizen scientists and not by law enforcement officers, document
489 fishing inside many of California's no-take MPAs. While observed much less frequently than
490 non-consumptive activities, fishing was still reported in 10% of all MPA Watch surveys
491 conducted in no-take MPAs. The vast majority of reported fishing in no-take areas was by
492 recreational anglers using hook-and-line fishing gear. In most cases, we suspect this was due to
493 a lack of education on the location of MPA boundaries by recreational anglers, as opposed to
494 deliberate poaching activities. The rare observation of commercial fishing in MPAs suggests
495 high compliance by the commercial fleet, which is highly informed about the location and
496 regulations of MPAs. This is consistent with official summaries showing that, in 2011 (the most
497 recent year with publicly available data), 271 citations were issued to commercial fishers while
498 10,052 citations were issued to recreational fishers (~4 times larger than the number issued to
499 recreational hunters) (CDFW, 2011). This suggests that outreach within the recreational fishing
500 community could be especially effective at increasing compliance with MPA regulations.

501

502 Our findings have several key management implications. If promoting human
503 engagement in MPAs is a management objective, our results suggest that MPA planners could
504 improve access and promote engagement either by (1) locating new MPAs in areas with
505 adjacent land-based parks and amenities or (2) investing in the development of new land-based
506 parks and/or amenities adjacent to existing MPAs. Furthermore, aligning protections on land
507 and sea could improve MPA performance by preventing pollution, sedimentation, or
508 eutrophication resulting from run-off from land-based activities (Cicin-Sain & Belfiore, 2005).
509 Alternatively, if reducing human engagement is desired — for example, to enhance protection of
510 biodiversity or other ecosystem or cultural services sensitive to human visitation or to limit
511 cumulative stressors to promote climate resilience — then planners could locate MPAs far from
512 people or land-based parks and amenities (Campbell et al., 2020). Our results could also help
513 guide decisions about where to invest in the monitoring, enforcement, and outreach programs
514 required to ensure compliance (Murray & Hee, 2019). We found that the citation frequency for

515 MPA rule violations increased with engagement and adjacent population size. These programs
516 may want to prioritize MPAs in areas of high population density and with adjacent land-based
517 amenities and sandy beaches. However, remote MPAs can also be areas of elevated non-
518 compliance due to lower levels of perceived risk of detection (Crawford et al., 2004; Rojo et al.,
519 2019), and enforcement should not entirely abandon these areas. In addition to monitoring and
520 enforcement, expanded education and outreach is needed to prevent non-compliance before it
521 happens, especially amongst recreational anglers (Bergseth & Roscher, 2018).

522

523 Equitable human engagement in California's MPA network is also an important
524 socioeconomic objective. Unfortunately, the indicators of engagement evaluated here do not
525 include demographic information on the identity of human users, limiting our ability to evaluate
526 the equity of engagement among different user groups. The collection of information in the
527 identity of MPA users is thus a vital first step towards considering equity in future MPA planning
528 and outreach. Knowledge of the representativeness of current users is necessary to design and
529 implement programs that promote access and engagement among underrepresented groups.
530 This knowledge could be gained by interviewing MPA visitors in intercept surveys and
531 assessing the composition of these users relative to that of surrounding communities (e.g.,
532 Scully-Engelmeyer et al., 2021). It could also be gained through focus groups with the various
533 community organizations that engage with MPAs, such as fishing, diving, and/or birding clubs,
534 or direct interaction with communities (e.g., Diedrich et al., 2017). The equity of access and
535 engagement should be considered at the outset of any additional MPA planning, including the
536 identification of methods for tracking and benchmarking progress towards these objectives. As
537 California prepares to expand its MPA network to meet 30x30 goals, it will be important to build
538 on the successes and lessons of the original participatory planning process (Gleason et al.,
539 2013) to further enhance the ability for ocean users, especially indigenous people, to ensure
540 that their values are reflected in the objectives, regulations, and design of the expanded network
541 (Barclay et al., 2017; Voyer et al., 2015; Voyer & Gladstone, 2018).

542

543 MPAs with low human engagement can still provide valuable contributions to the human
544 engagement, conservation, and fisheries goals of the MPA network. While total engagement at
545 some MPAs is low, these MPAs could be more important to small but underserved human
546 populations in the neighboring area. This is a key benefit of the MLPA's spacing requirements,
547 which mandated that California's MPAs be placed within 50-100 km of each other (Saarman &
548 Carr, 2013). This spacing ensures that coastal populations have relatively similar access to

549 MPAs along the entire California coast. Thus, while MPAs in low population areas have lower
550 engagement, the people living in these areas have opportunities for access similar to people
551 living in higher population areas. Furthermore, MPAs also aim to achieve conservation and
552 fisheries benefits and MPAs with low human engagement can be critical contributors to these
553 goals. This is especially true given that human engagement with MPAs has the potential to
554 negatively impact ecosystem function and MPA performance (Milazzo et al., 2002). Limiting
555 human engagement can also reduce the cumulative impacts of multiple stressors on MPAs,
556 including climate change, eutrophication, and pollution (Mach et al., 2017). MPAs with low
557 human engagement are thus key in the design of effective MPA networks, as they can buffer or
558 offset the impacts of human activities in MPAs with greater engagement and limit cumulative
559 impacts in a multi-stressor environment. A network of MPAs, like that in California, provides the
560 opportunity to design individual MPAs that meet differing criteria and perspectives regarding
561 human-nature relationships (Pereira et al., 2020) while contributing to overall network
562 performance across a range of axes.

563

564 The methodological framework developed here presents a useful starting point for
565 assessing human engagement in any MPA network. To start, the iNaturalist and eBird citizen
566 science programs already have wide global coverage and REEF has high participation in many
567 regions. Other social media platforms, such as Instagram, Twitter, and Flickr, may also be used
568 to assess how, when, and where people engage in MPAs (Retka et al., 2019; Tenkanen et al.,
569 2017). However, these indicators do not capture all types of human engagement or all of the
570 information needed to understand the ecological impacts of human engagement or the equity of
571 engagement amongst different human populations. Notably, our indicators do not capture
572 information on: (1) user demographics, which are key for understanding equality in access
573 (Nicholls & Shafer, 2001); (2) activities that have negative ecological impacts, such as
574 anchoring (Creed & Amado Filho, 1999); or (3) money spent on licenses, entry fees, food, gas,
575 and lodging, among other expenses associated with human engagement in MPAs, which are
576 helpful in quantifying the broader impact of MPAs to local economies (Sala et al., 2013).
577 Furthermore, the types of engagement evaluated here, especially engagement in science and
578 tourism, likely undercount underserved and disadvantaged communities, as the geoscientific
579 community remains largely white (Dutt, 2020) and the expense of tourism and even coastal
580 parking can be a barrier to engagement. Notably, our analysis does not explicitly account for
581 tribal and indigenous engagement with MPAs, which is an important consideration for

582 California's MPA network. In addition, some of our datasets have known biases. For example,
583 iNaturalist observations require the use of a smartphone, which may exclude some user groups.

584

585 Understanding the ability and enabling conditions for MPAs to achieve human
586 engagement objectives is important as entities around the world aim to protect 30% of the
587 ocean by 2030 to meet objectives for people and nature (CBD, 2021). This paper presents a
588 transferable framework for evaluating human engagement with MPA networks and our analyses
589 indicate that human engagement can potentially be increased by placing or developing MPAs
590 near people in concert with existing land-based attractions or amenities. Critical next steps in
591 MPA and human engagement research are to identify strategies for designing MPA networks to
592 promote equitable human engagement, capturing the full extent and value of MPAs in promoting
593 recreation and tourism, education and outreach, and scientific research, and minimizing
594 negative impacts of engagement on the conservation and fisheries objectives.

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610 endorse any particular analytical methods, interpretations, or conclusions based upon the data it
611 provides. Unless otherwise stated, use of CDFW's data does not constitute CDFW's
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614 Author Contributions Statement

615 CMF, JGE, JEC, JGS, KJN, and TBF conceived the ideas and designed methodology; CMF,
616 JGE, JGS, JB, KJN, and TBF collected the data; CMF, JGS, and JB analyzed the data; CMF led
617 the writing of the manuscript. All authors contributed critically to the drafts and gave final
618 approval for publication.

619 Conflict of Interests Statement

620 The authors have no conflicts of interest to declare.

621 Data Availability Statement

622 All code and data are available on GitHub here: <https://github.com/NCEAS/ca-mpa>

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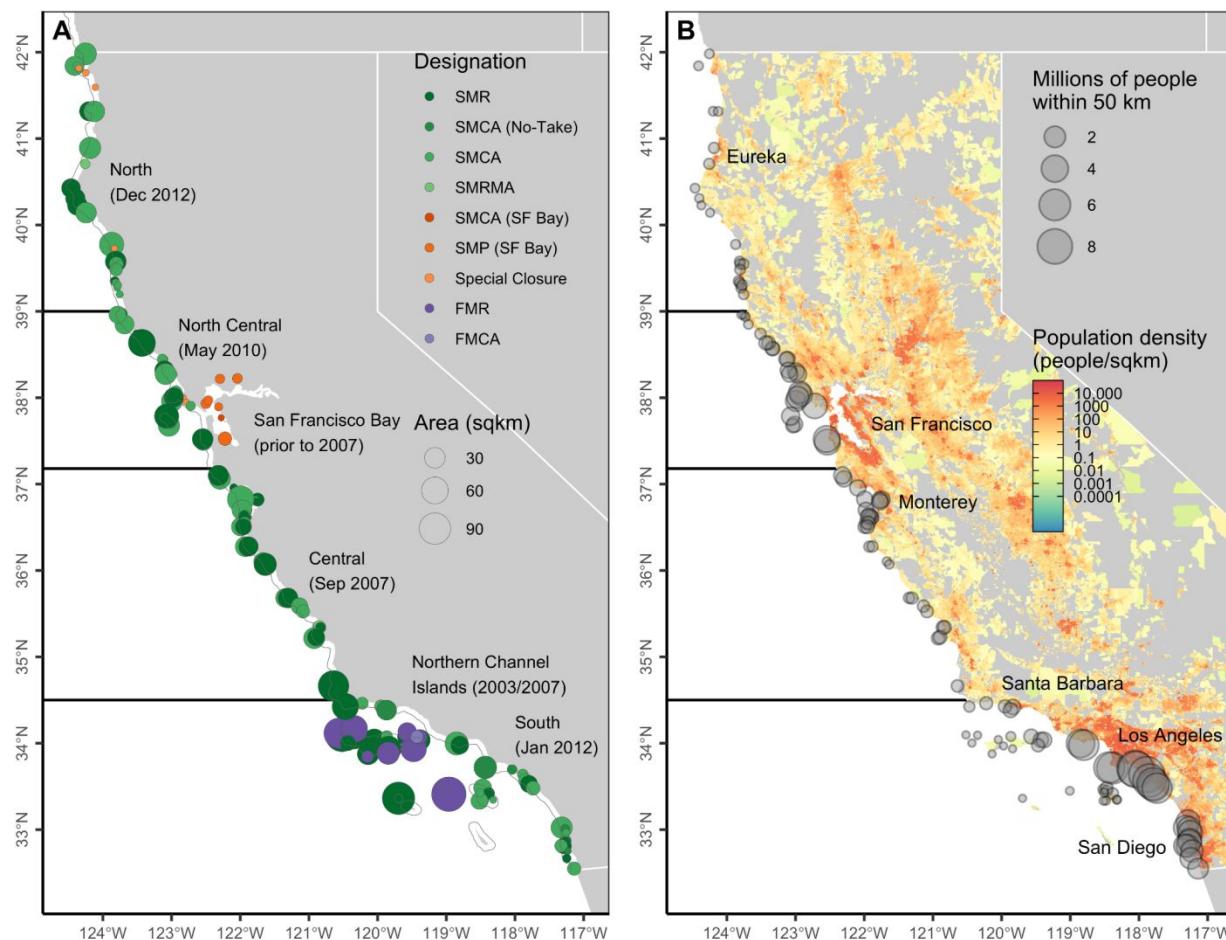
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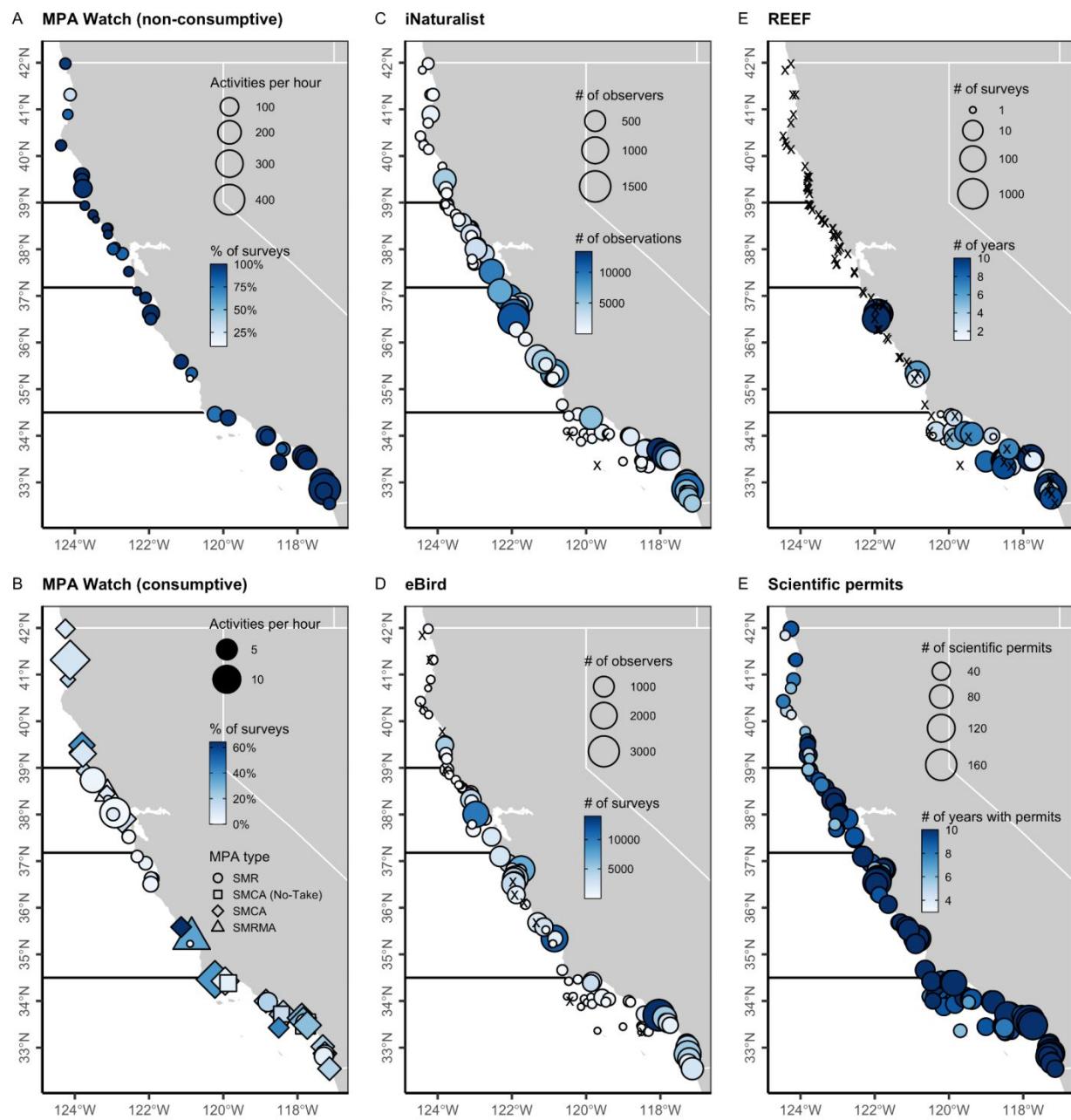
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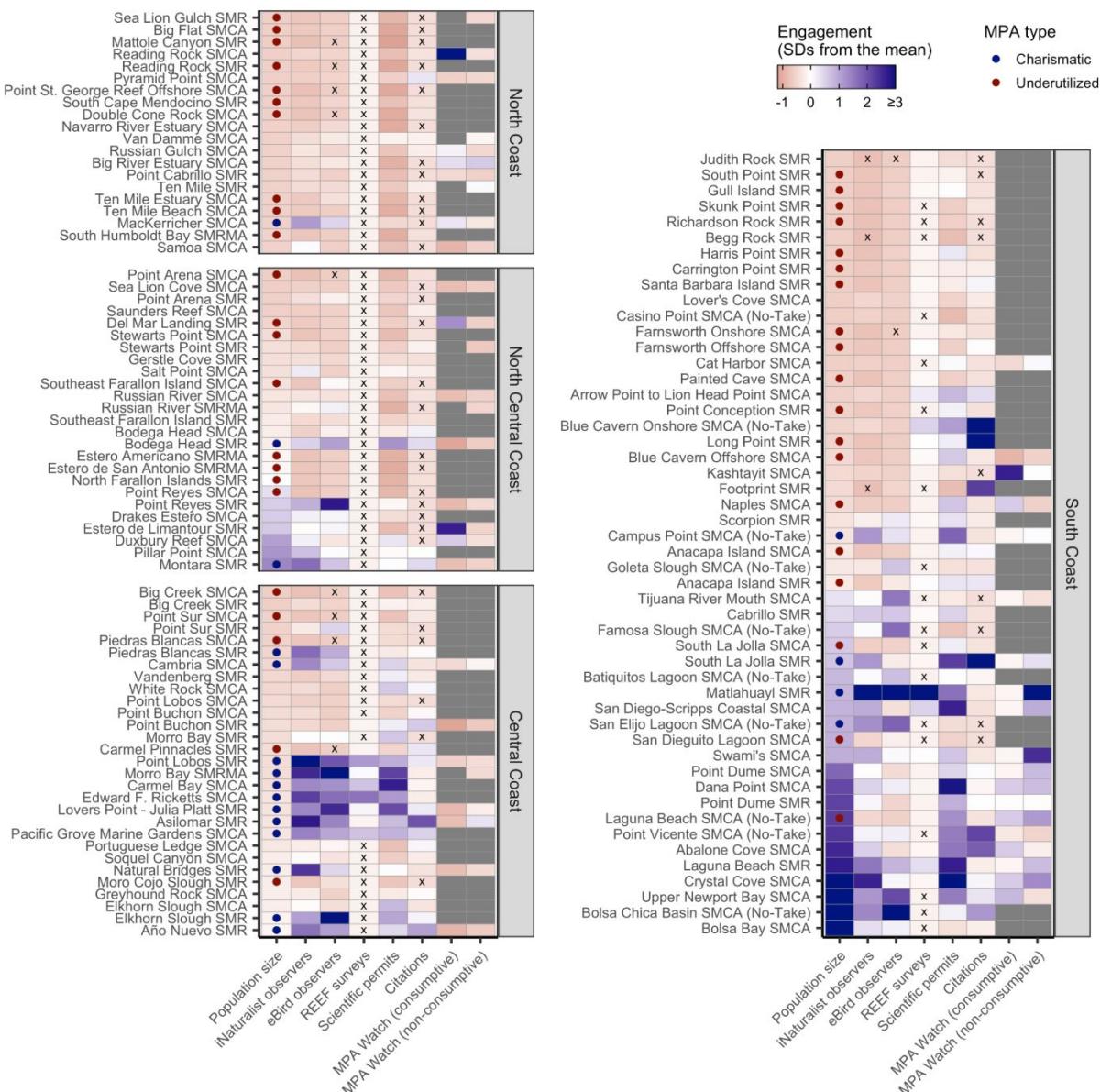
918 **Tables & Figures**



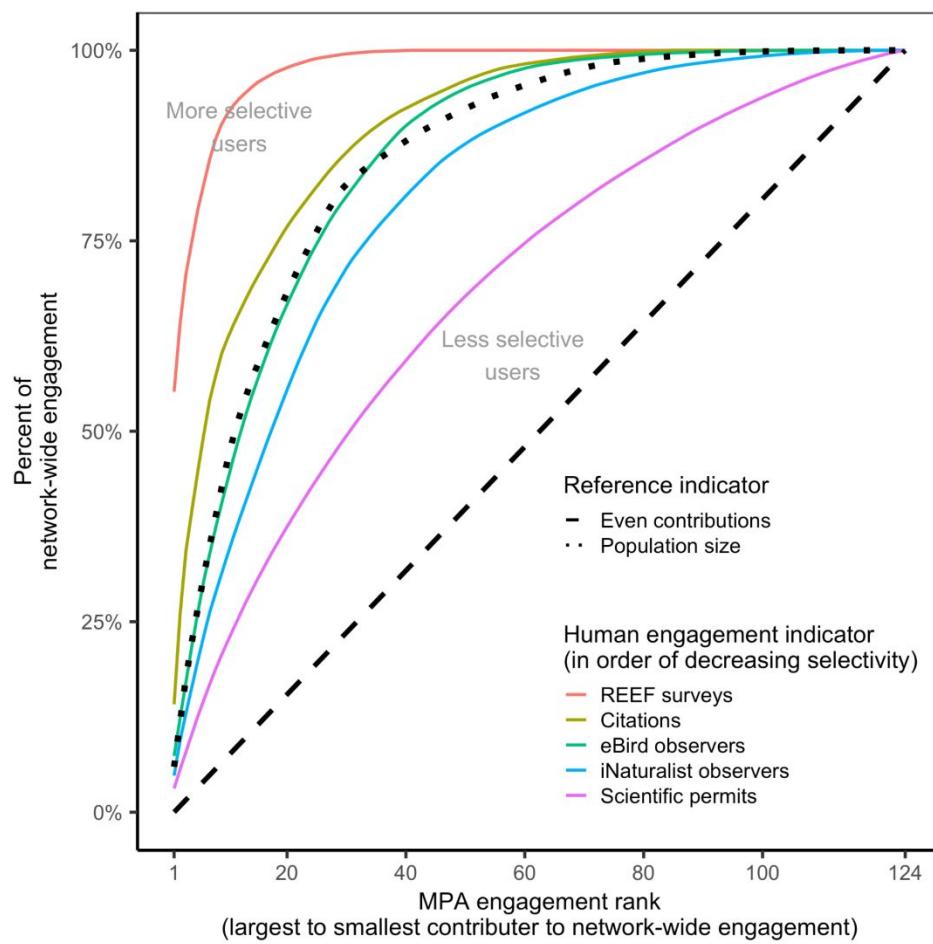
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 920 **Figure 1.** Maps illustrating **(A)** California's marine protected area (MPA) network and **(B)** nearby
 921 human population density. In **(A)**, greens indicate state MPAs established by the Marine Life
 922 Protection Act (MLPA), oranges indicate state MPA designations excluded from the analysis,
 923 and purples indicate federal MPAs excluded from the analysis. See **Section 2.1** and **Table S1**
 924 for the definition of each MPA designation. Point size indicates MPA area (km^2). Dark horizontal
 925 lines delineate the four primary MLPA regions (labeled with month of implementation). MPAs in
 926 the San Francisco Bay region were established before 2007 and were not part of the MLPA
 927 planning effort. MPAs in the Northern Channel Islands were also established before MLPA
 928 (2003 and 2007 in state and federal waters, respectively) but have been officially incorporated
 929 into the network. The thin gray line indicates state waters (3 nautical miles offshore). In **(B)**,
 930 point size indicates the number of people living within 50 km of each MPA. Colors indicate
 931 population density by census block in the 2010 U.S. Census. A few key coastal cities are
 932 labeled for reference.



933
934 **Figure 2.** Maps illustrating six indicators of human engagement in California's state marine
935 protected area (MPA) network. Multiple metrics are used to measure engagement for each
936 indicator; see **Table S3** for definitions of these metrics. Across indicators, larger symbols and
937 deeper colors indicate higher engagement. In C-F, black x's mark MPAs without any reported
938 engagement. Dark horizontal lines delineate the four MLPA regions. See **Figure S17** for a map
939 of the regulatory citations indicator.

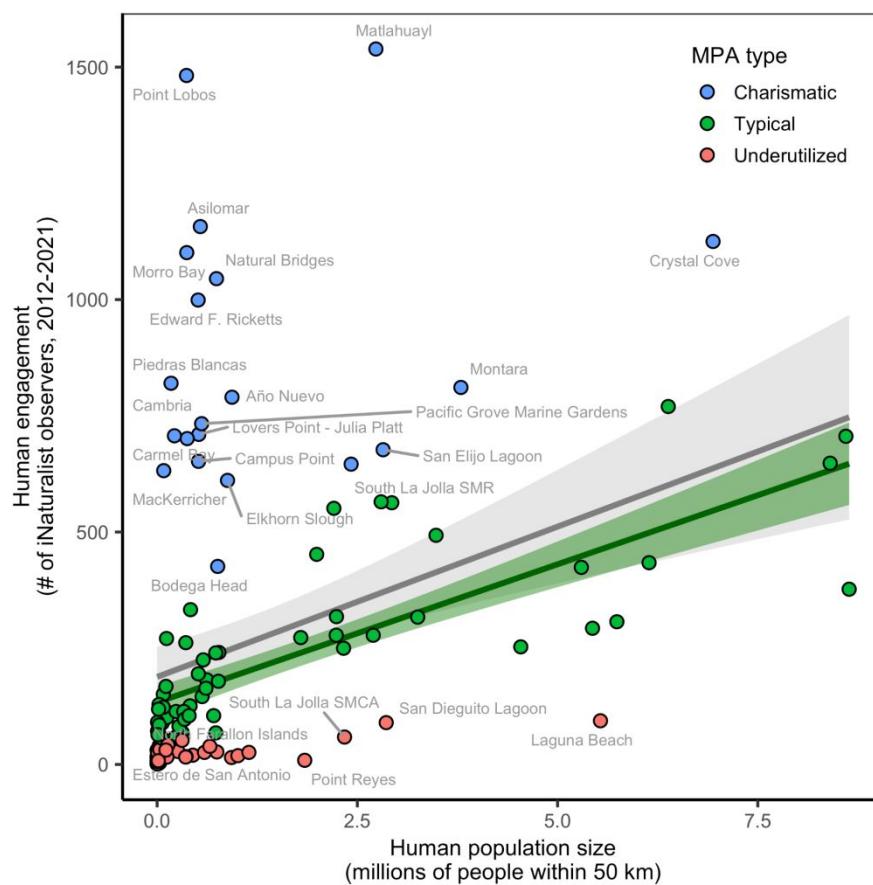


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941 **Figure 3.** A synthesis of human engagement indicators within California's state marine
942 protected areas (MPAs). MPAs are sorted by population density within 50 km (first column of
943 each plot) within each region. Engagement indicators are centered on the average of each
944 indicator and scaled to unit variance to ease comparison across indicators; thus, color indicates
945 the number of standard deviations (SDs) from the mean where blue shades indicate MPAs with
946 above average engagement and red shades indicate MPAs with below average engagement.
947 Gray indicates MPAs without data and x's indicate MPAs with true zeros. MPAs with greater
948 ("charismatic") and less ("underutilized") engagement than expected based on surrounding
949 population density are marked in the population size column. See **Table S3** for definitions and
950 metrics of the displayed indicators.



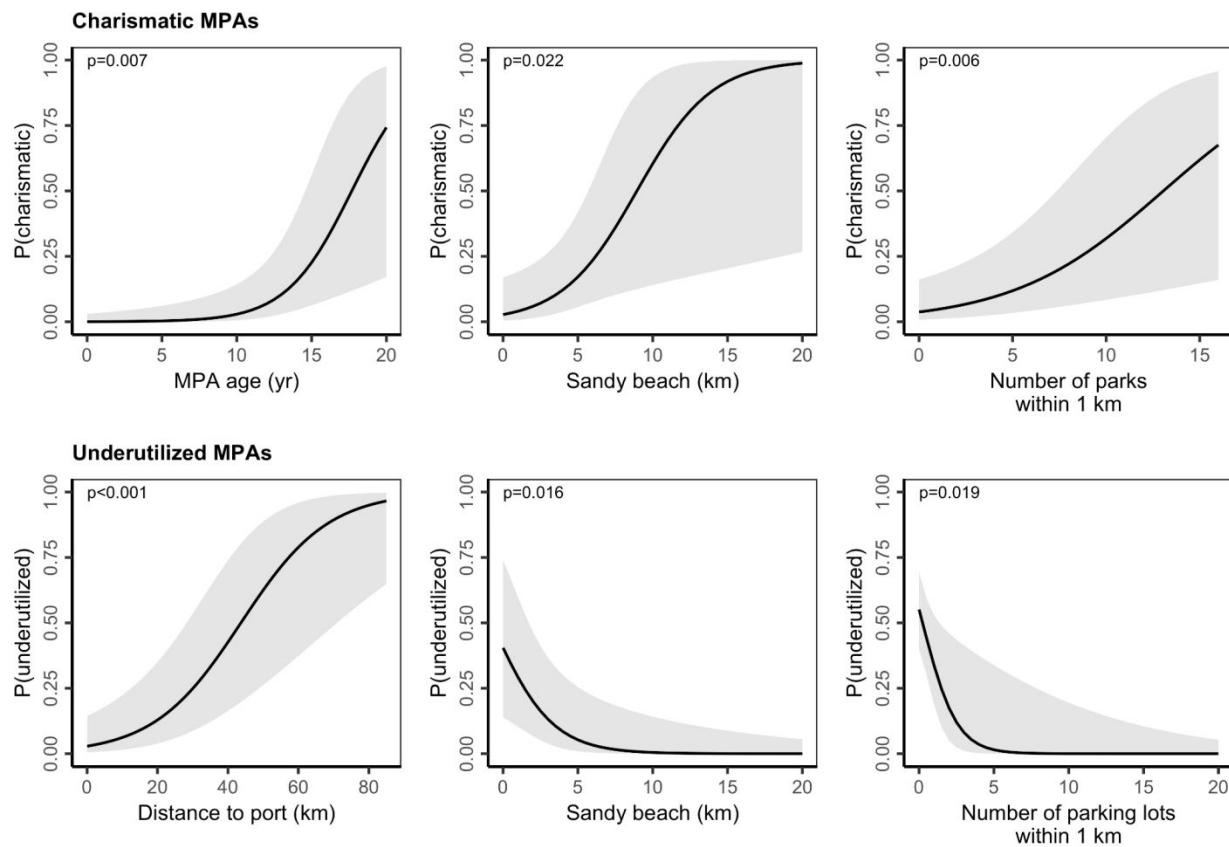
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Figure 4. Cumulative contributions of individual marine protected areas (MPAs) to network-wide engagement based on several indicators of human engagement. The diagonal dashed line indicates a theoretical accumulation curve in which individual protected areas contribute equally to engagement within the overall network. Curved lines above this reference line indicate accumulation curves in which some protected areas make larger contributions (higher performers) to network-wide engagement than others (lower performers); the steeper the curve, the more network-wide engagement is dominated by a few protected areas. The accumulation curve for population size (dotted black line) provides an additional frame of reference: if human engagement were proportional to population size, engagement would accumulate according to this curve. Thus, curves steeper than this line indicate that benefits are more concentrated than would be predicted by population density (i.e., engagement is more selective) whereas curves shallower than this line indicate a more even distribution of benefits than would be predicted by population density (i.e., engagement is less selective). The MPA Watch indicators are excluded because they are not available for all MPAs within the network.



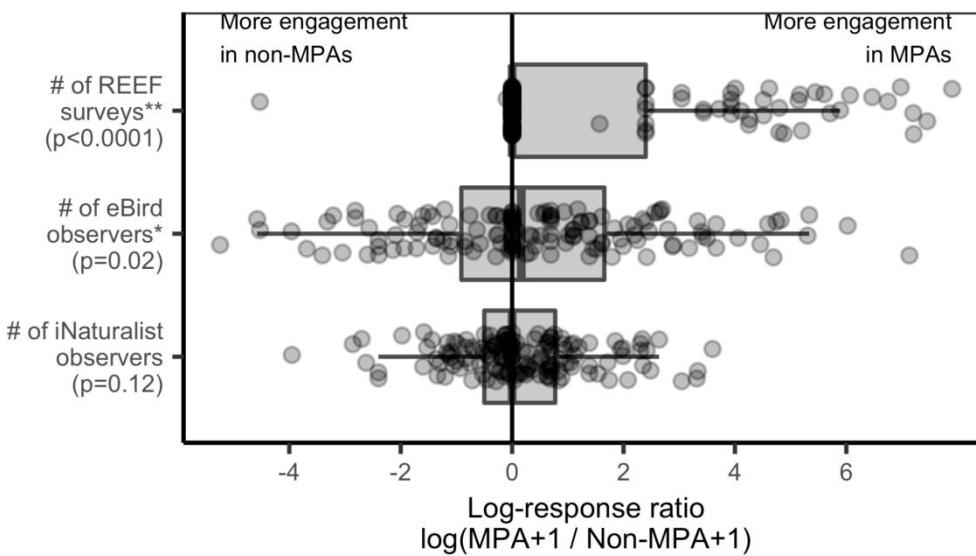
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967 **Figure 5.** Correlation between human engagement in an MPA and the number of people living
 968 within 50 km of the area. Human engagement is measured as the number of iNaturalist
 969 observers submitting observations within 100 m of an MPA from 2012 through 2021. The gray
 970 line and 95% confidence interval illustrate a linear regression ($r^2=0.14$; $p<0.001$) fit to all points.
 971 Blue points with residuals greater than 75% of the fitted values were classified as “charismatic”
 972 MPAs, whose engagement is higher than would be expected based on population density. Red
 973 points with residuals less than 75% of the fitted values were classified as “underutilized” MPAs,
 974 whose engagement is lower than would be expected based on population density. The
 975 charismatic and selected underutilized MPAs are labeled with their abbreviated names. The
 976 green line and 95% confidence interval illustrate a linear regression ($r^2=0.62$; $p<0.001$) fit to the
 977 “typical” protected areas (green points), whose engagement is largely determined by population
 978 density.



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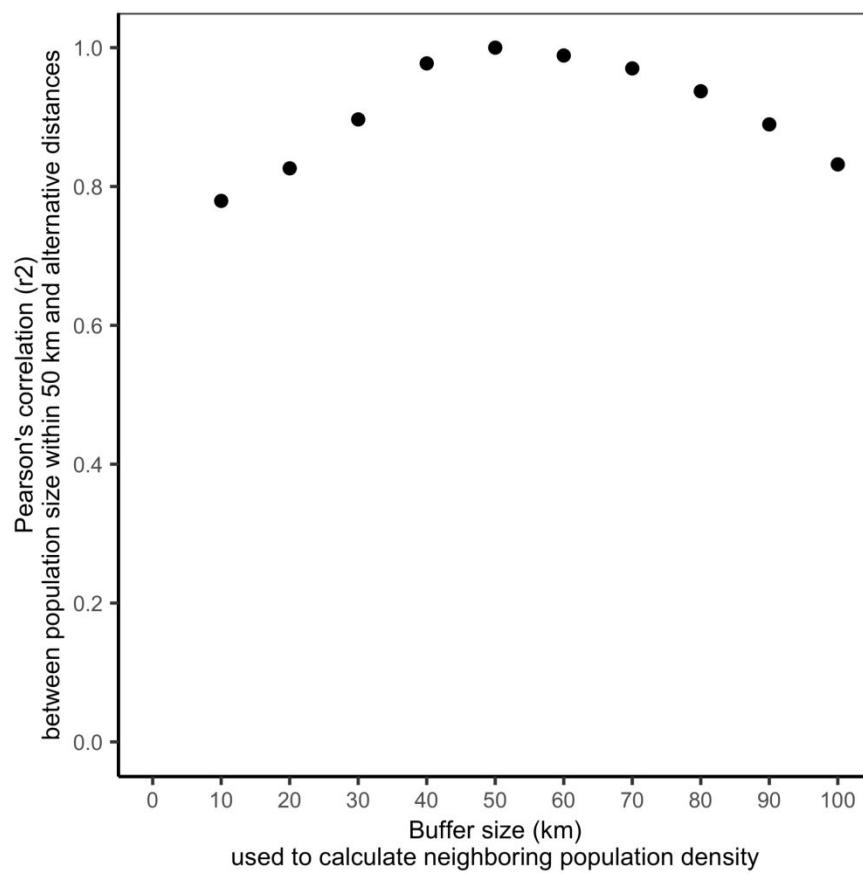
Figure 6. Marginal effects of significant predictors of “charismatic” (top row) and “underutilized” (bottom row) MPAs as identified through stepwise logistic regression. Marginal effects represent the predicted probability when varying the variable of interest while fixing the other variables at their means. Shading depicts 95% confidence intervals. See **Table S5** for the list of predictors included in each model and **Table S7** for the results of each model fit.



985

986 **Figure 7.** The level of human engagement in marine protected areas (MPAs) compared to non-
 987 MPA counterfactuals for indicators with the required data. Log-response ratios were calculated
 988 after adding 1 to the engagement values occurring in both the numerator and denominator to
 989 avoid non-finite values. Log-response ratios greater than zero indicate MPAs where the MPA
 990 designation is associated with higher engagement relative to the counterfactual whereas ratios
 991 less than zero indicate MPAs where the MPA designation is associated with lower engagement
 992 relative to the counterfactual. Asterisks indicate indicators whose mean response ratio is
 993 significantly different from zero (* p<0.01, ** p<0.05). P-values are shown parenthetically. In the
 994 boxplots, the solid line indicates the median, the box indicates the interquartile range (IQR; 25th
 995 to 75th percentiles), the whiskers indicate 1.5 times the IQR, and the points beyond the
 996 whiskers indicate outliers. Points represent log-response ratios for each MPA and
 997 counterfactual pair.

998 Supplemental Tables & Figures

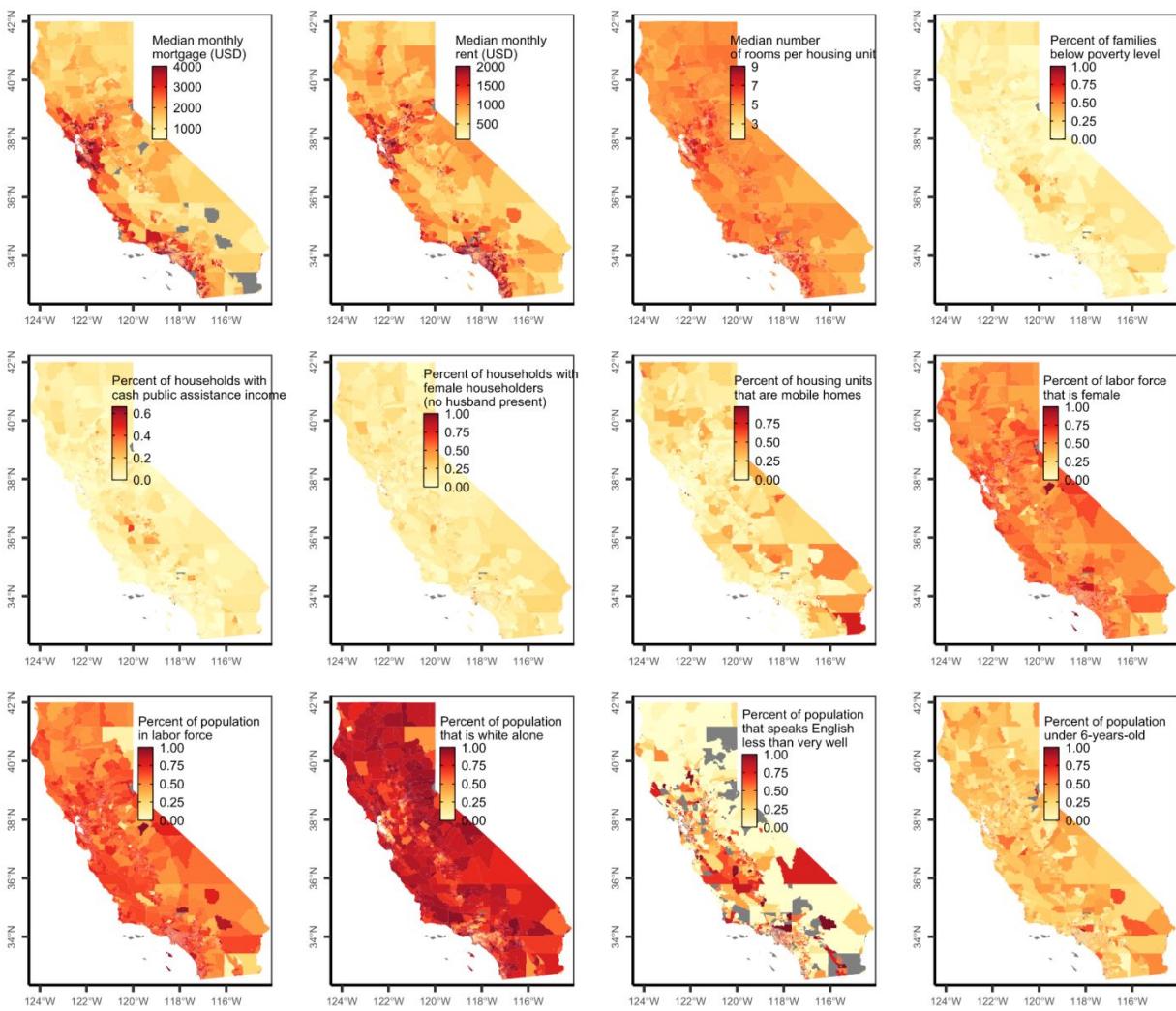


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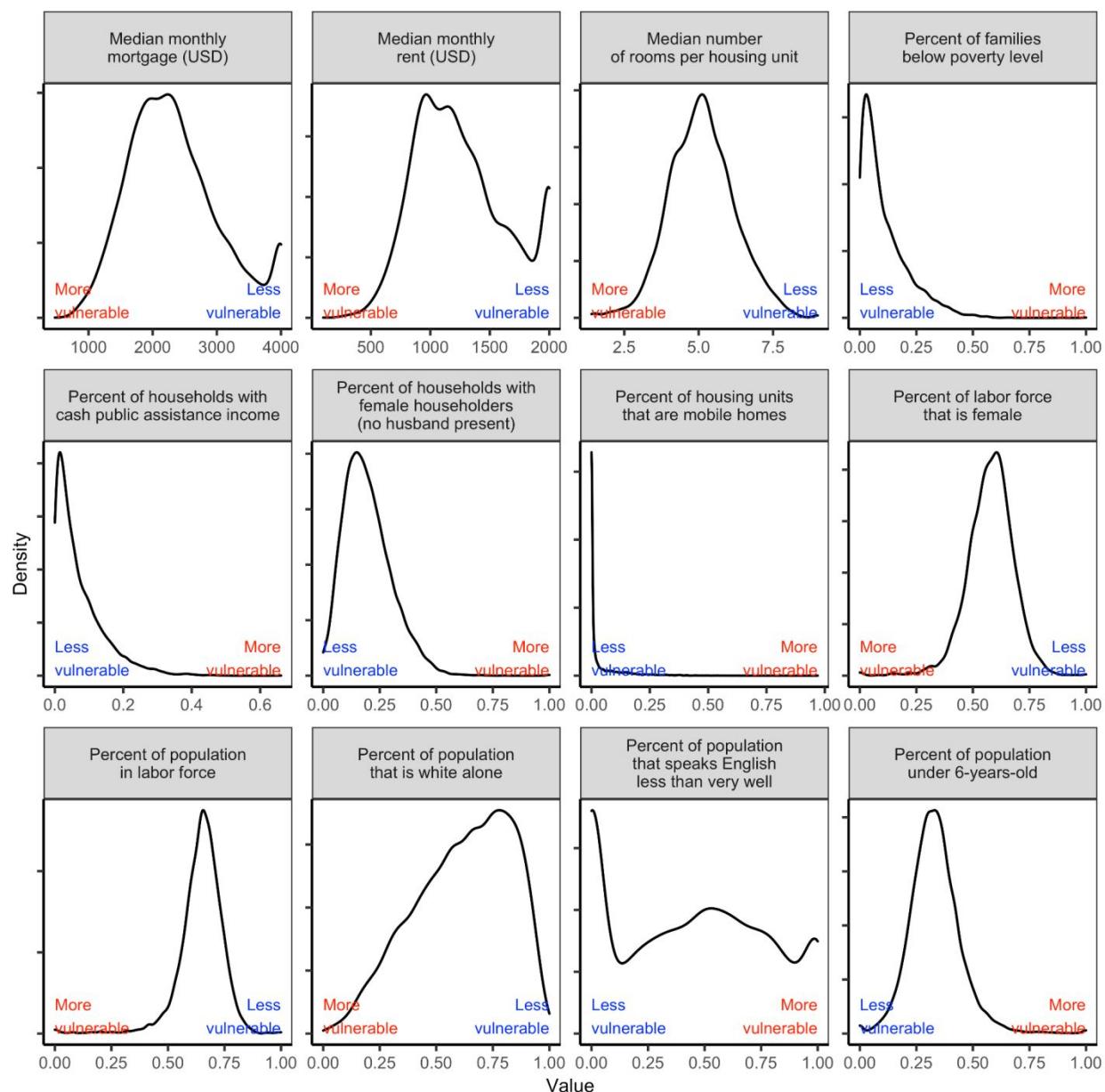
1001 **Figure S1.** The correlation between population density calculated using the selected 50 km

buffer and population densities calculated using alternative buffer distances.



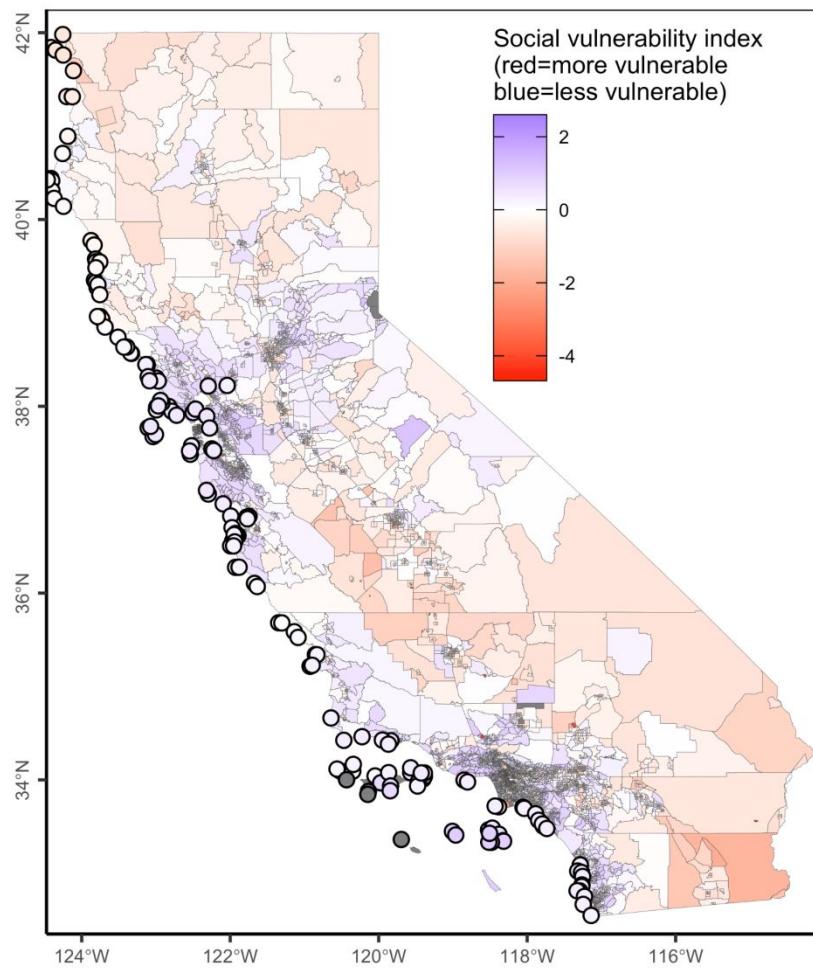
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Figure S2. Maps of the social vulnerability indicator used to calculate the social vulnerability index by California US Census tract.



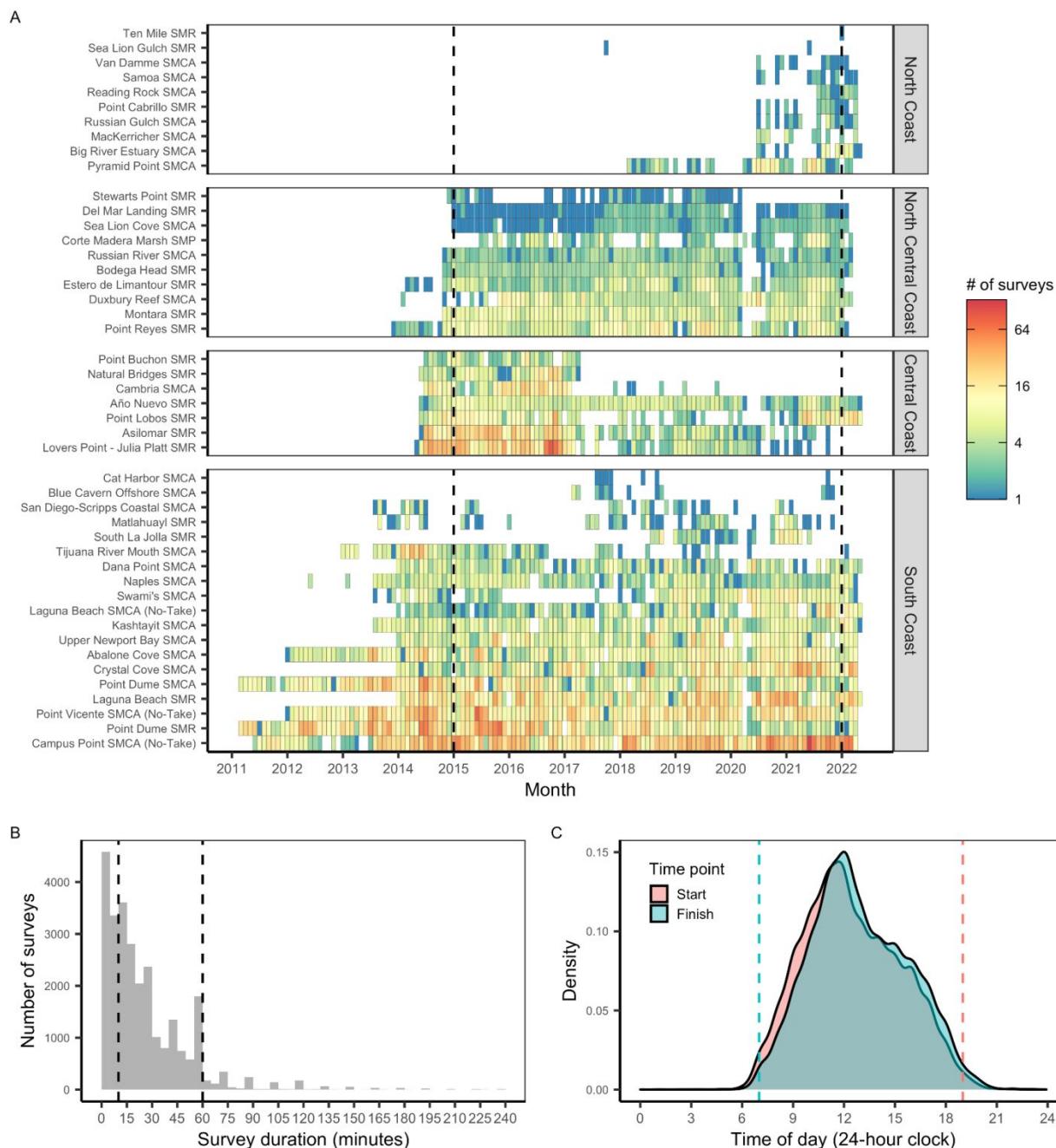
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Figure S3. Distribution of the social vulnerability indicator values by California US Census tract used to calculate the social vulnerability index. Values were centered on the statewide average and scaled to unit variance. Indicators in which higher vulnerability is indicated by higher values (e.g., percent of families below poverty level, percent of households with cash public assistance income) were multiplied by -1 so that higher vulnerability is represented as low values for all indicators. The social vulnerability index, mapped in **Figure S4**, was calculated as the average of the centered, scaled, and standardized indicators.



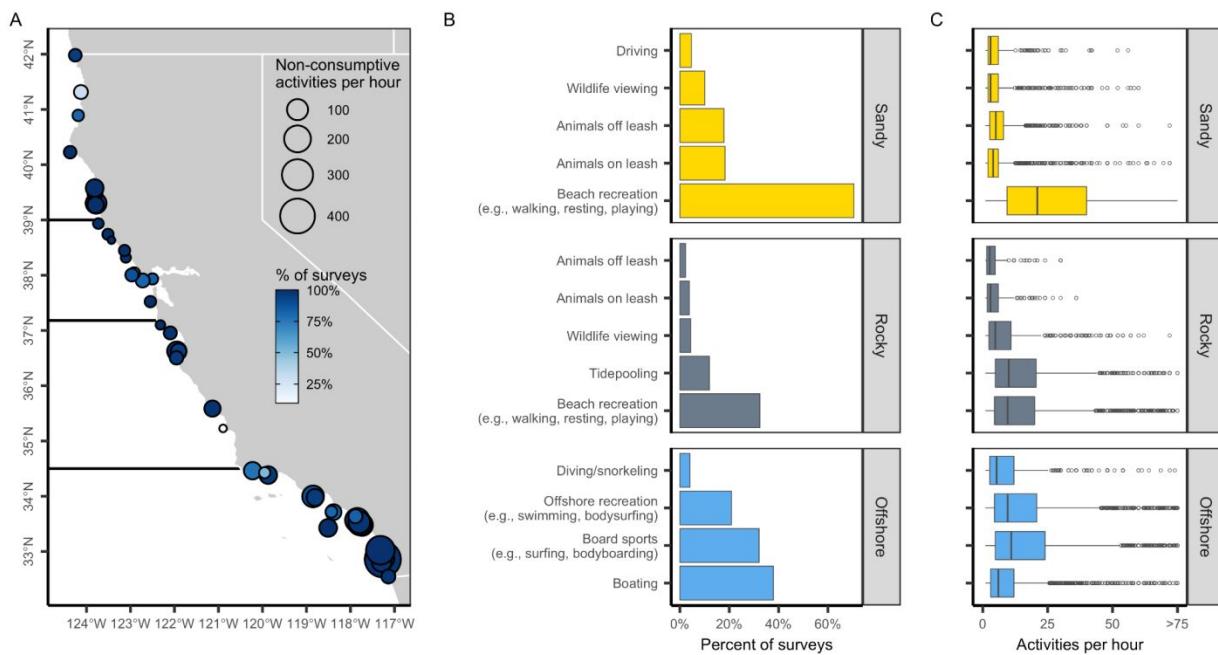
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Figure S4. Social vulnerability index by US Census tract (polygons on land) and average social vulnerability index within 50 km of each MPA (points at sea). The social vulnerability index is calculated as the average of the 12 indicators of social vulnerability described in **Table S2**. Indicators were centered on their statewide average and scaled to unit variance before the index was calculated. Negative (red) values indicate higher social vulnerability and positive (blue) values indicate lower social vulnerability.



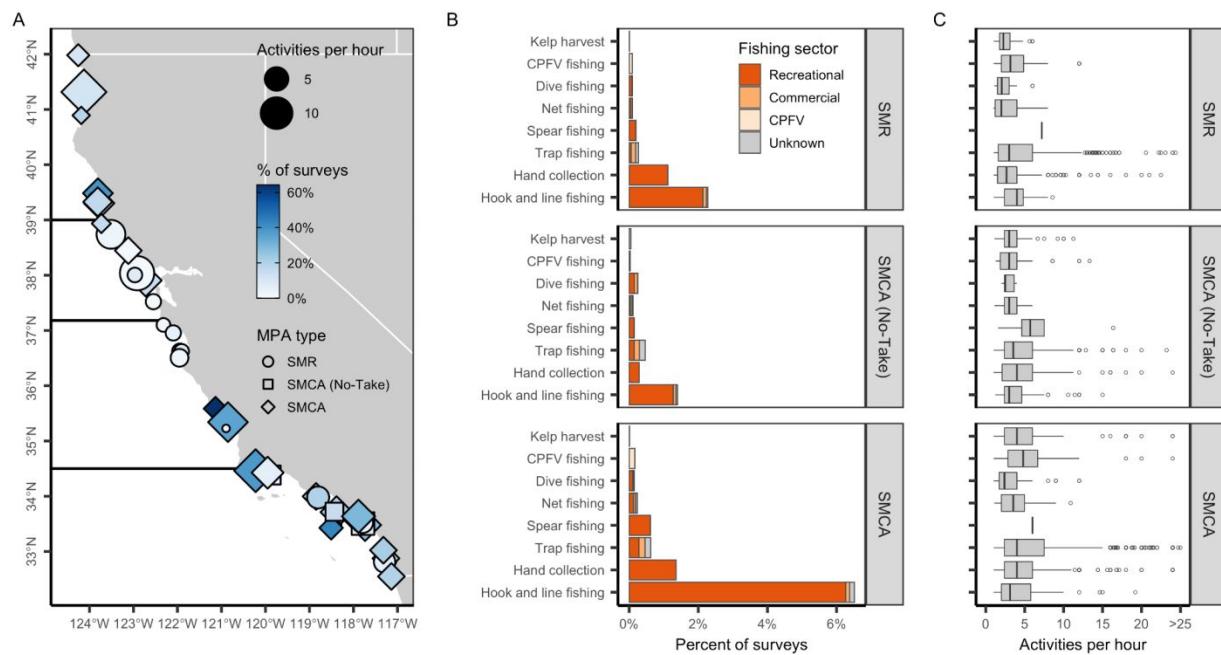
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Figure S5. The (A) coverage of usable MPA Watch surveys over time by marine protected area (MPA). A usable survey is a survey in which the duration was accurately recorded (i.e., end time occurs after start time). Note log-scale for fill color. San Francisco Bay MPAs are plotted in the North Central Coast region for simplicity. Only surveys occurring between January 1, 2015 and December 31, 2022 were considered in the analysis. We also excluded (B) surveys shorter than 10 minutes or longer than 60 minutes and (C) surveys ending before 7AM or starting after 7PM.

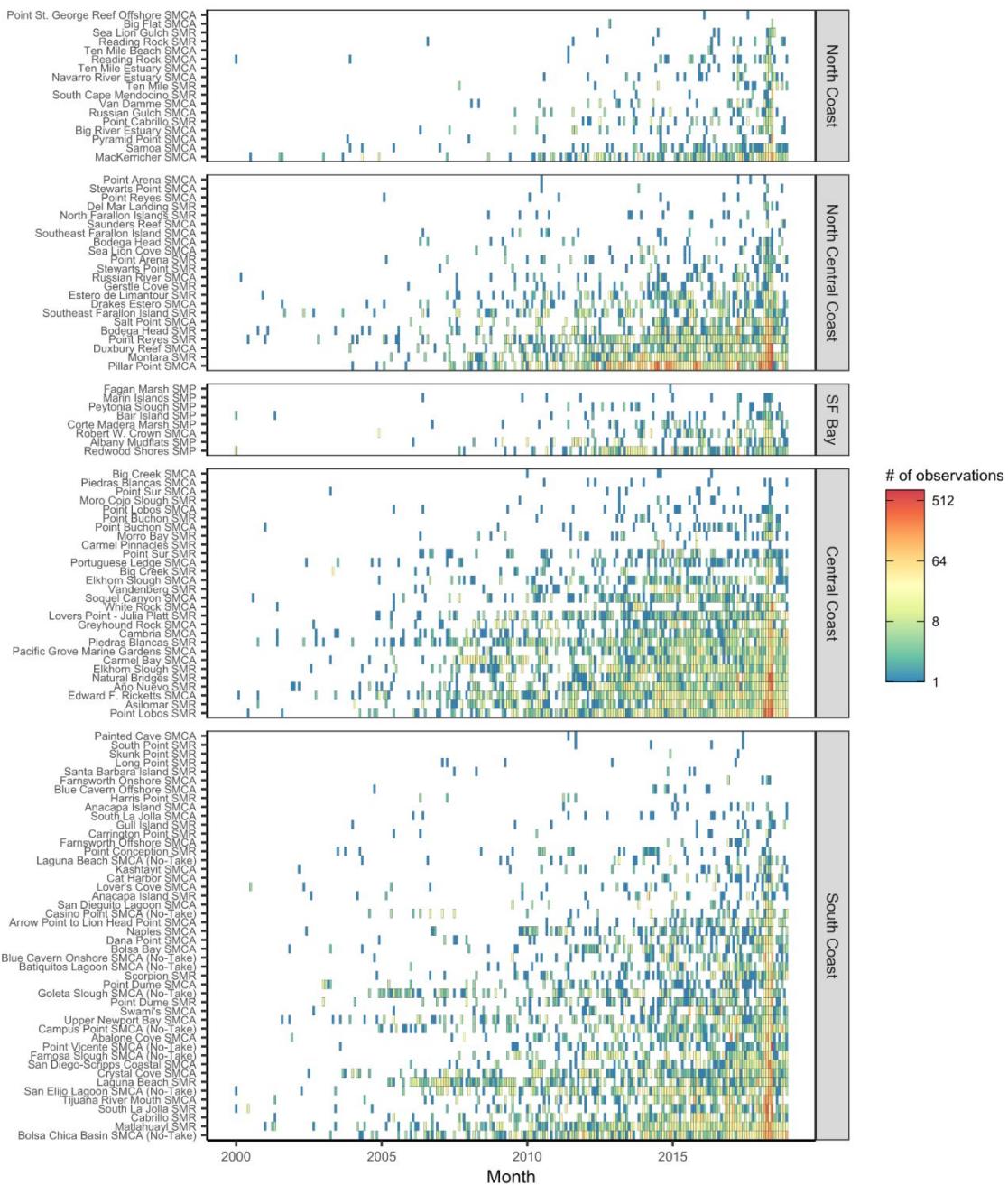


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Figure S6. Non-consumptive activities in California's state marine protected areas (MPAs) based on surveys conducted by MPA Watch. Panel **A** shows the percent of surveys within an MPA in which non-consumptive activities were observed (color ramp) and the median number of non-consumptive activities observed per hour (point size) for surveys in which such activities were observed (i.e., zeroes excluded). Dark horizontal lines delineate the four MLPA regions. Panel **B** shows the percent of surveys in which non-consumptive activities were observed by habitat area. Panel **C** shows the number of non-consumptive activities observed per hour for surveys in which such activities were observed (i.e., zeroes excluded). In the boxplots, the solid line indicates the median, the box indicates the interquartile range (IQR; 25th to 75th percentiles), the whiskers indicate 1.5 times the IQR, and the points beyond the whiskers indicate outliers.



1039
1040 **Figure S7.** Active consumptive activities in California's state marine protected areas (MPAs)
1041 based on surveys conducted by MPA Watch. Two SMRMA are categorized as SMCAs to
1042 increase visibility. Panel **A** shows the percent of surveys within MPAs of varying levels of
1043 protection (point shape) in which active consumptive activities were observed (color ramp) and
1044 the median number of active consumptive activities observed per hour (point size) for surveys in
1045 which such activities were observed (i.e., zeroes excluded). Dark horizontal lines delineate the
1046 four MLPA regions. Panel **B** shows the percent of surveys in which active consumptive activities
1047 were observed by fishing sector (CPFV=commercial passenger fishing vessel). Panel **C** shows
1048 the number of active consumptive activities observed per hour for surveys in which such
1049 activities were observed (i.e., zeroes excluded). In the boxplots, the solid line indicates the
1050 median, the box indicates the interquartile range (IQR; 25th to 75th percentiles), the whiskers
1051 indicate 1.5 times the IQR, and the points beyond the whiskers indicate outliers.



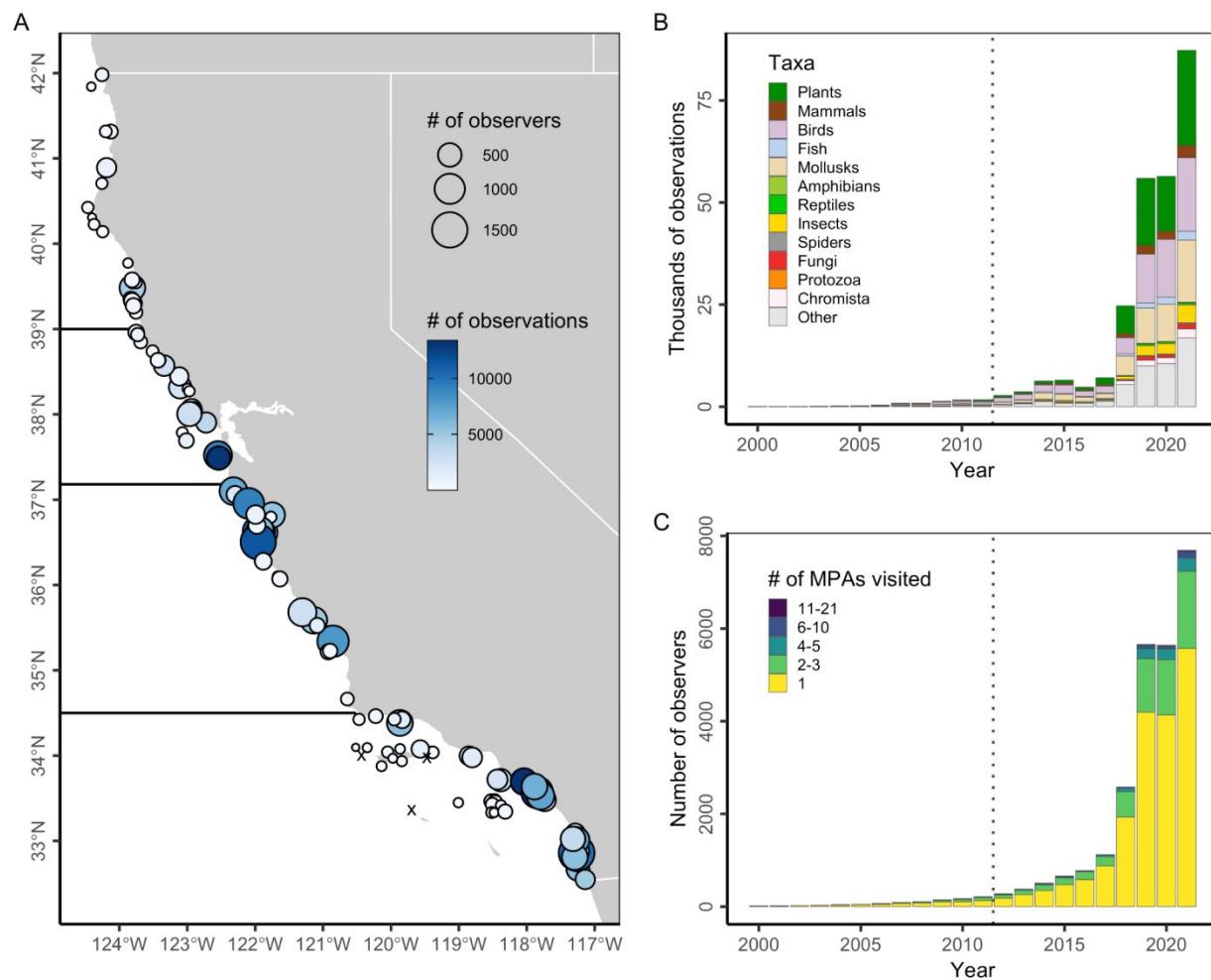
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Figure S8. Coverage of iNaturalist observation data over time by marine protected area (MPA).

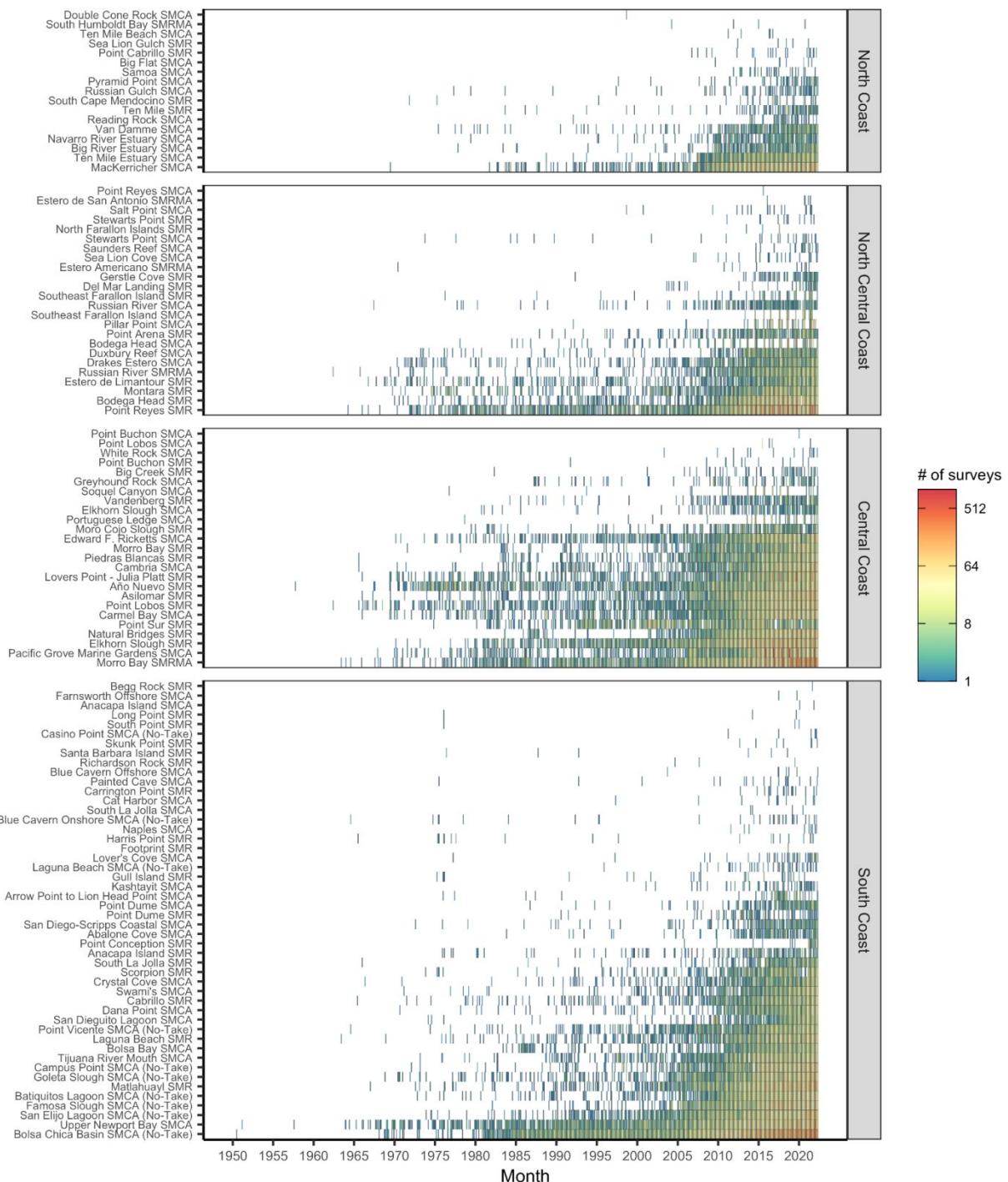
Note log-scale for fill color. MPAs are listed in order of overall sample size within each region.

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 1056 **Figure S9.** Human interest in wildlife within California's state marine protected areas (MPAs)
 1057 based on usage of the iNaturalist web- and app-based application. Panel **A** shows the number
 1058 of observers (point size) and observations (color ramp) within 100 m of California's MPAs from
 1059 2012 through 2021. Note log-scale in fill color. Black x's mark the 4 MPAs without any
 1060 iNaturalist submissions. Dark horizontal lines delineate the four MLPA regions. Panel **B** shows
 1061 the number of observations made within 100 m of California MPAs from 2000-2021 by
 1062 taxonomic group. Panel **C** shows the number of observers making observations within 100 m of
 1063 California MPAs from 2000-2021 grouped by the number of MPAs that they visited.

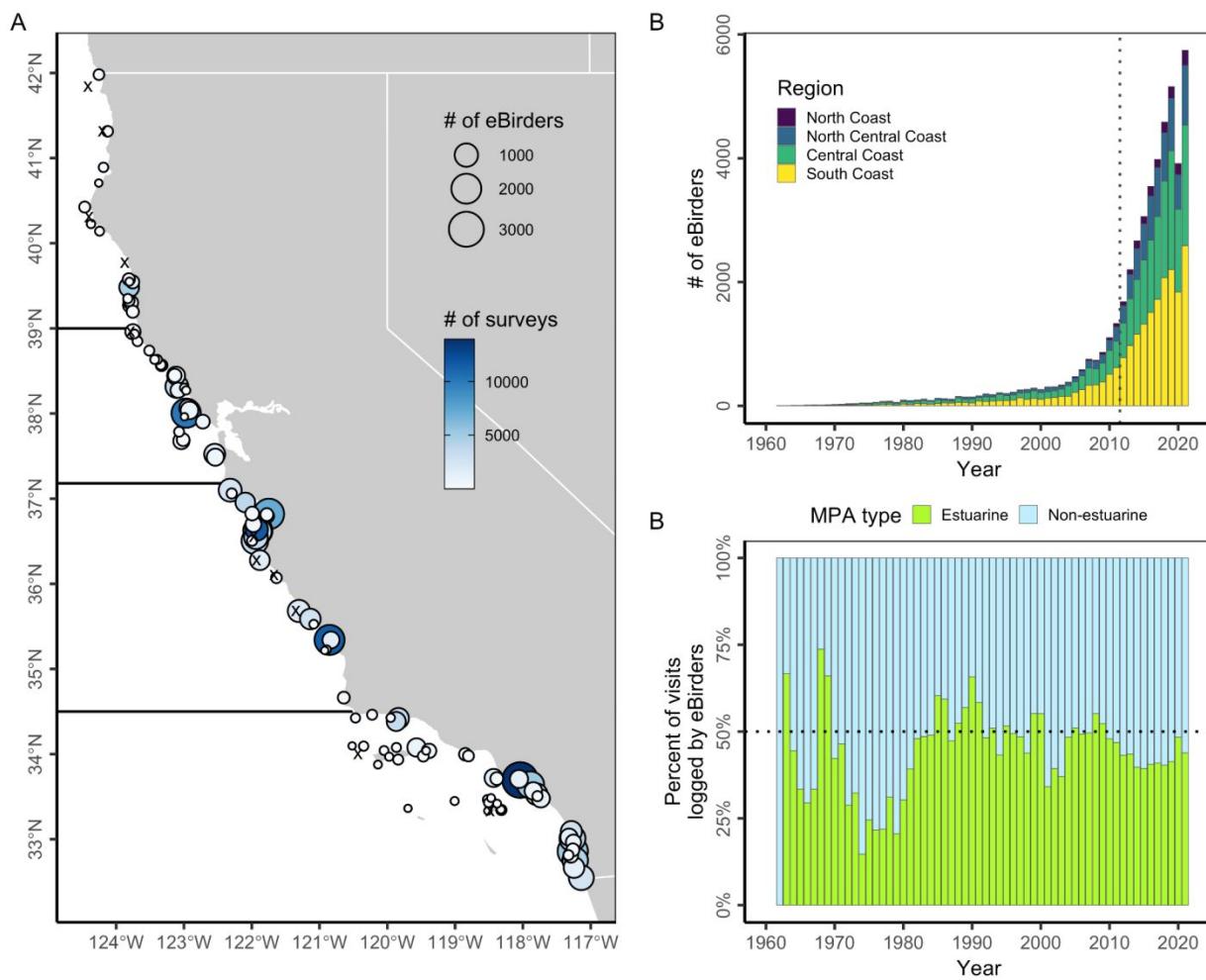


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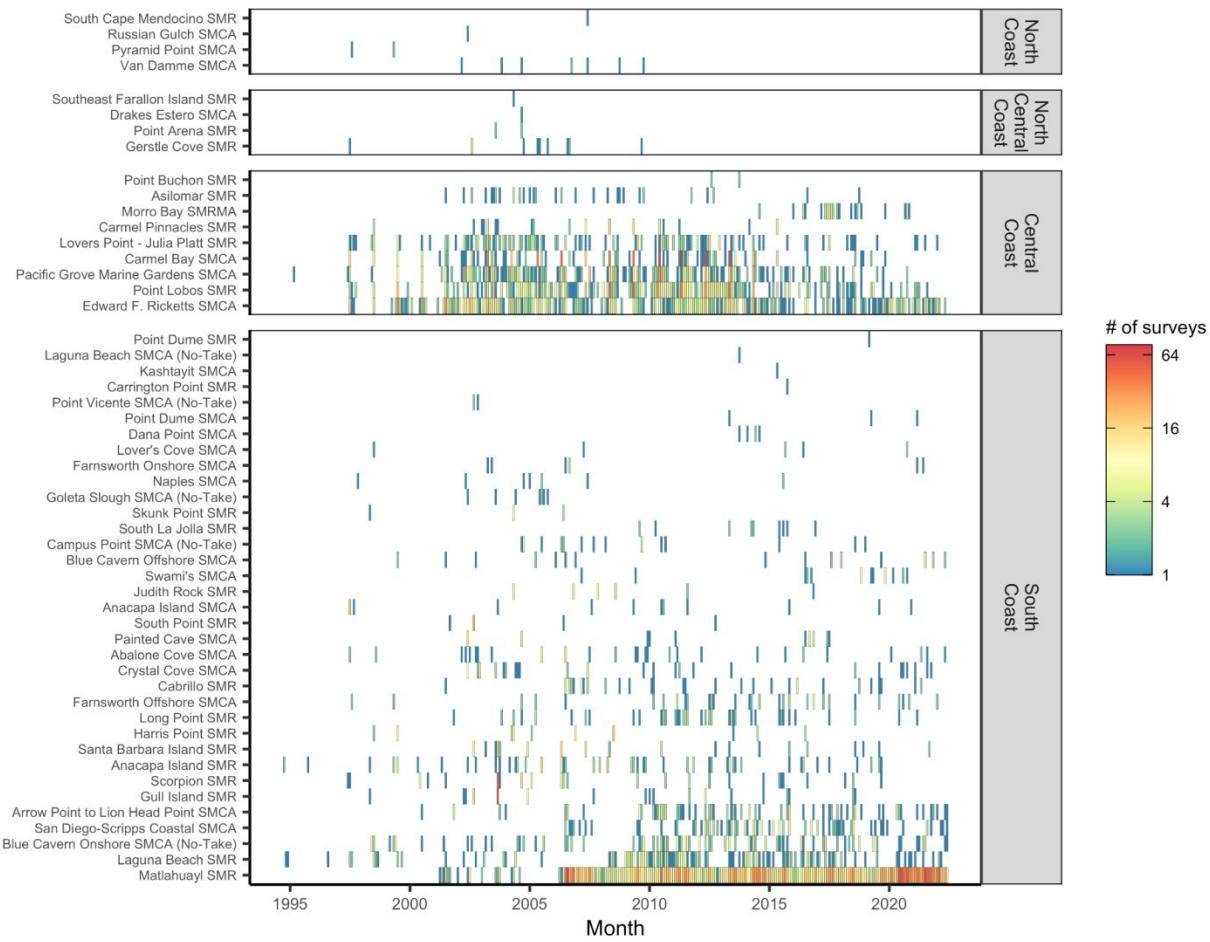
Figure S10. Coverage of eBird observation data over time by marine protected area (MPA). Note log-scale for fill color. MPAs are listed in order of overall sample size within each region.

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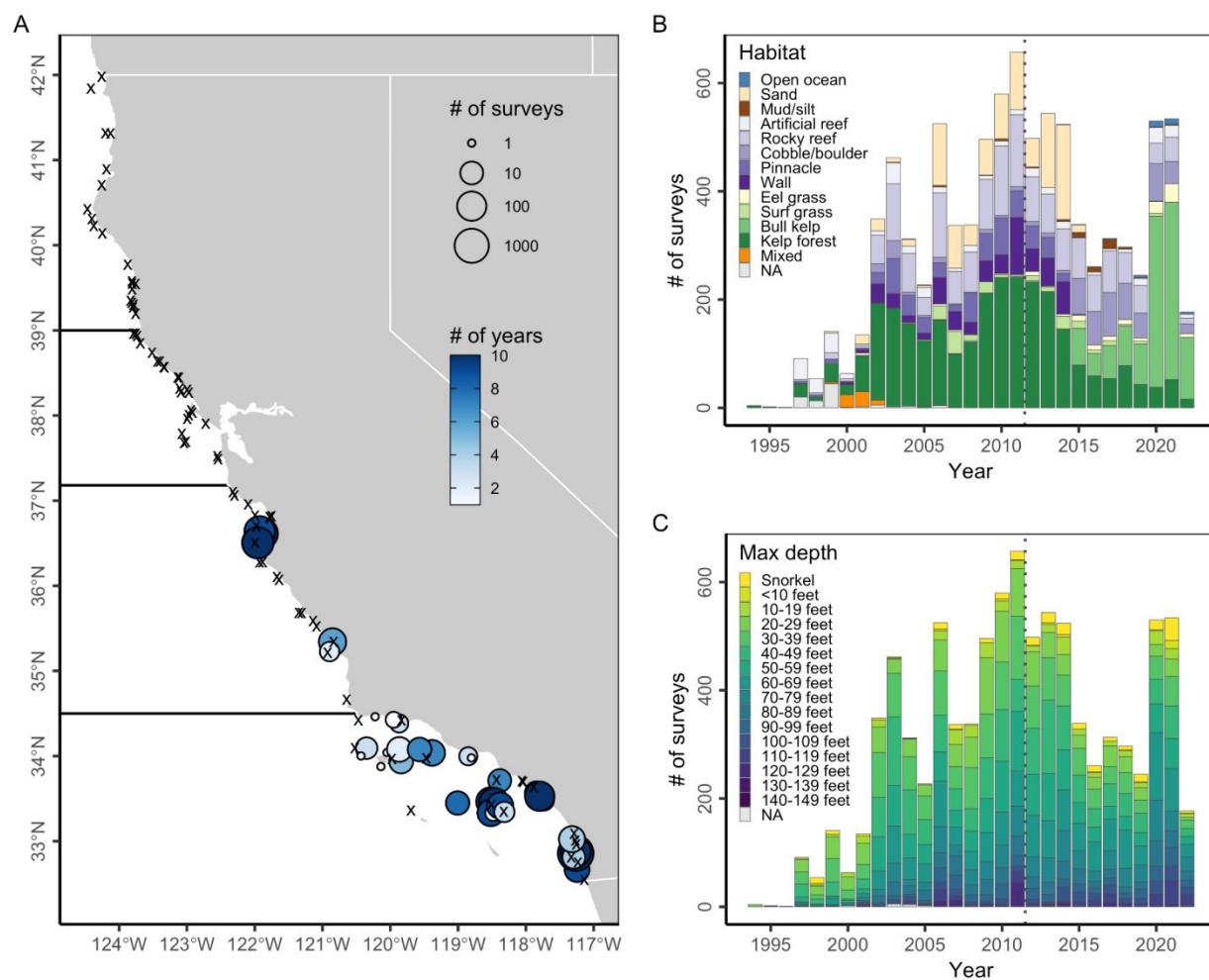


1067
 1068 **Figure S11.** Human engagement in birding within California's state marine protected areas
 1069 (MPAs) based on submissions to the eBird citizen science program. Panel **A** shows the total
 1070 number of eBirders (point size) and surveys (color ramp) submitted by eBirders from within 100
 1071 m of California's MPAs from 2012 through 2021. Black x's mark the 11 MPAs without any eBird
 1072 submissions. Dark horizontal lines delineate the four MLPA regions. Panel **B** shows the number
 1073 of eBirders making observations from within 100 m of California MPAs from 1960-2021. Panel **C**
 1074 shows the percent of visits to MPAs logged by eBirders occurring from within estuarine and non-
 1075 estuarine MPAs from 1960-2021. Estuarine MPAs represent 2% of the network by area and
 1076 17% by count.

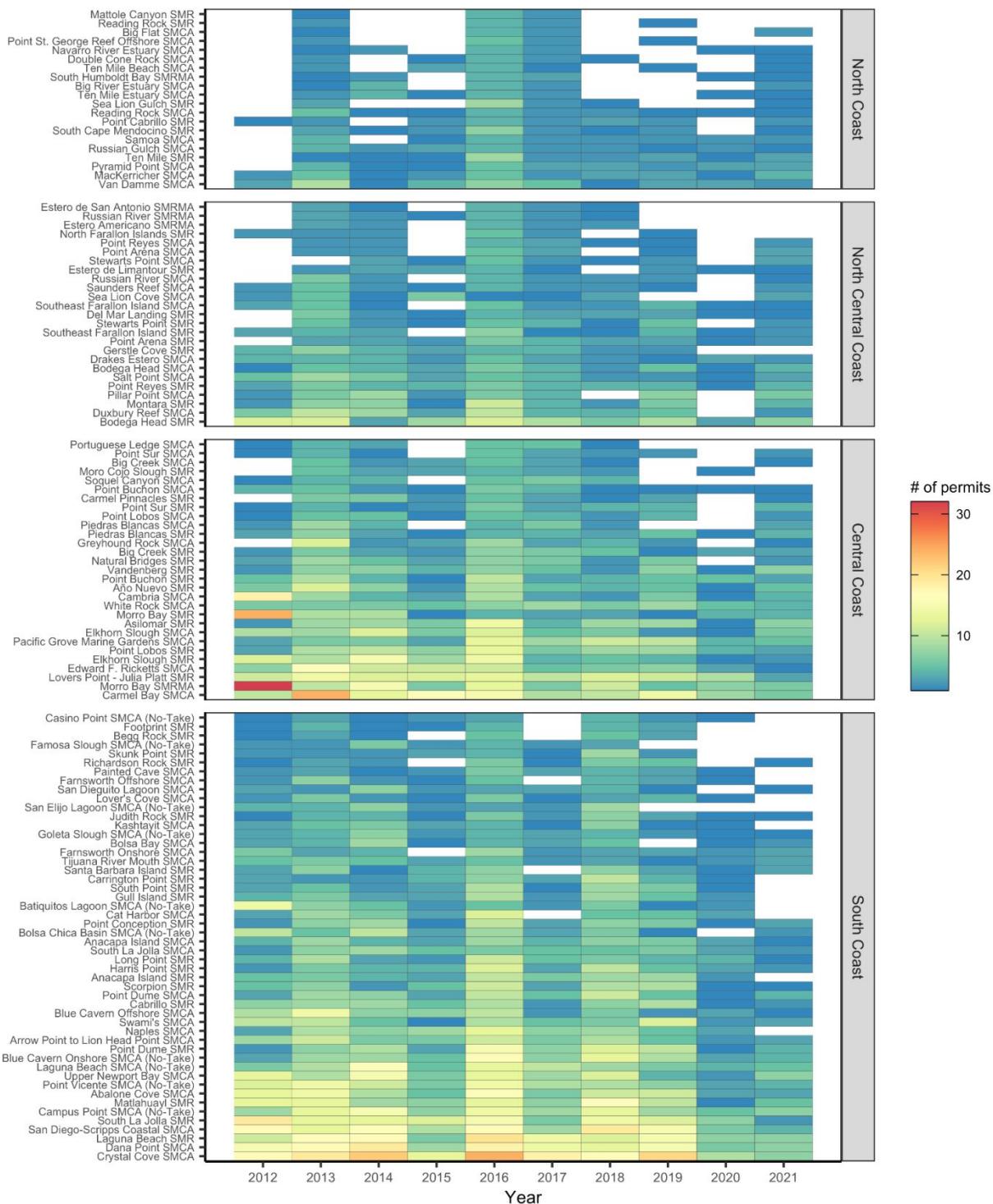


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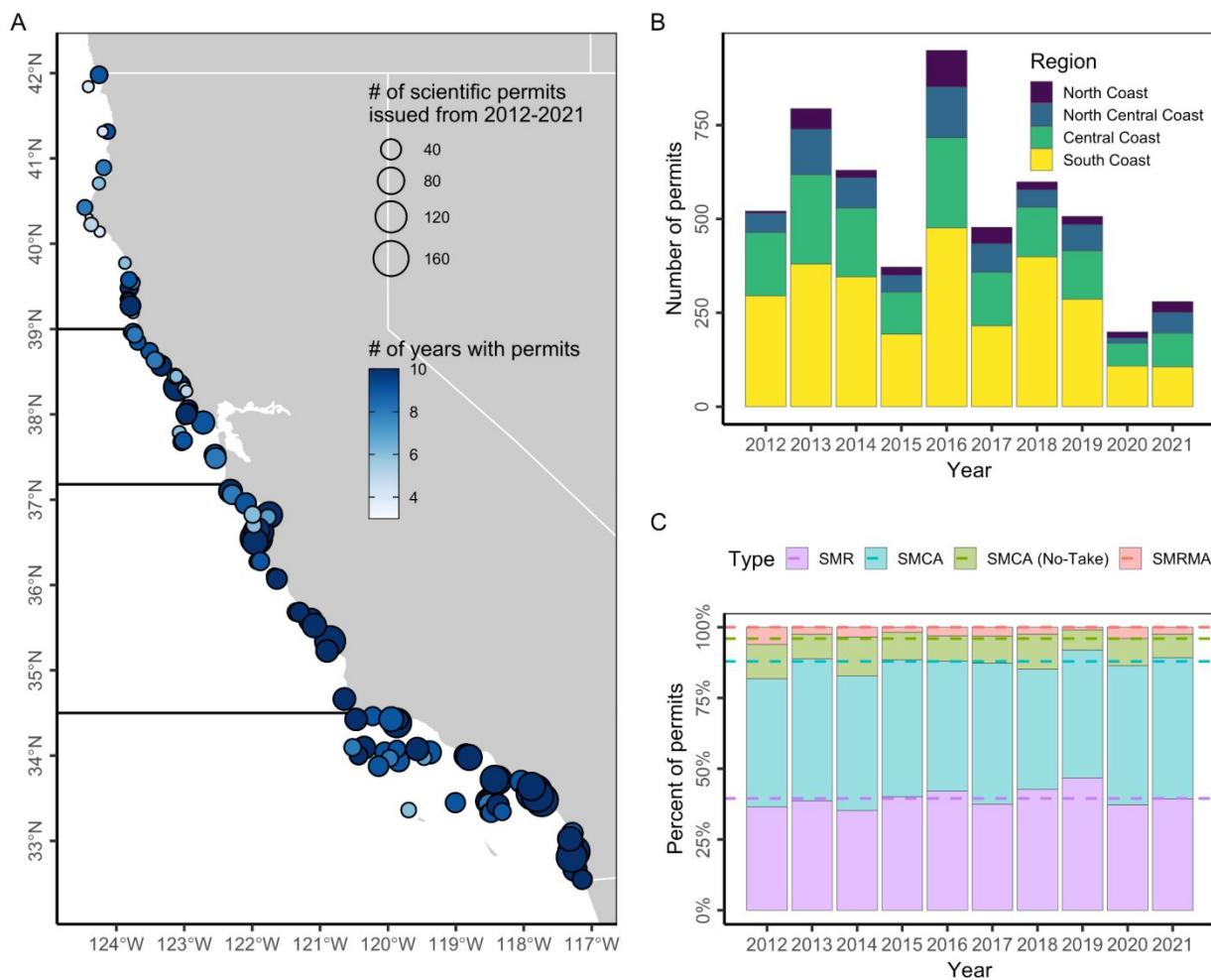
Figure S12. Coverage of REEF survey data over time by marine protected area (MPA). Note log-scale for fill color. MPAs are listed in order of overall sample size within each region. One San Francisco Bay MPA (Redwood Shores SMP) is plotted in the North Central Region for simplicity.



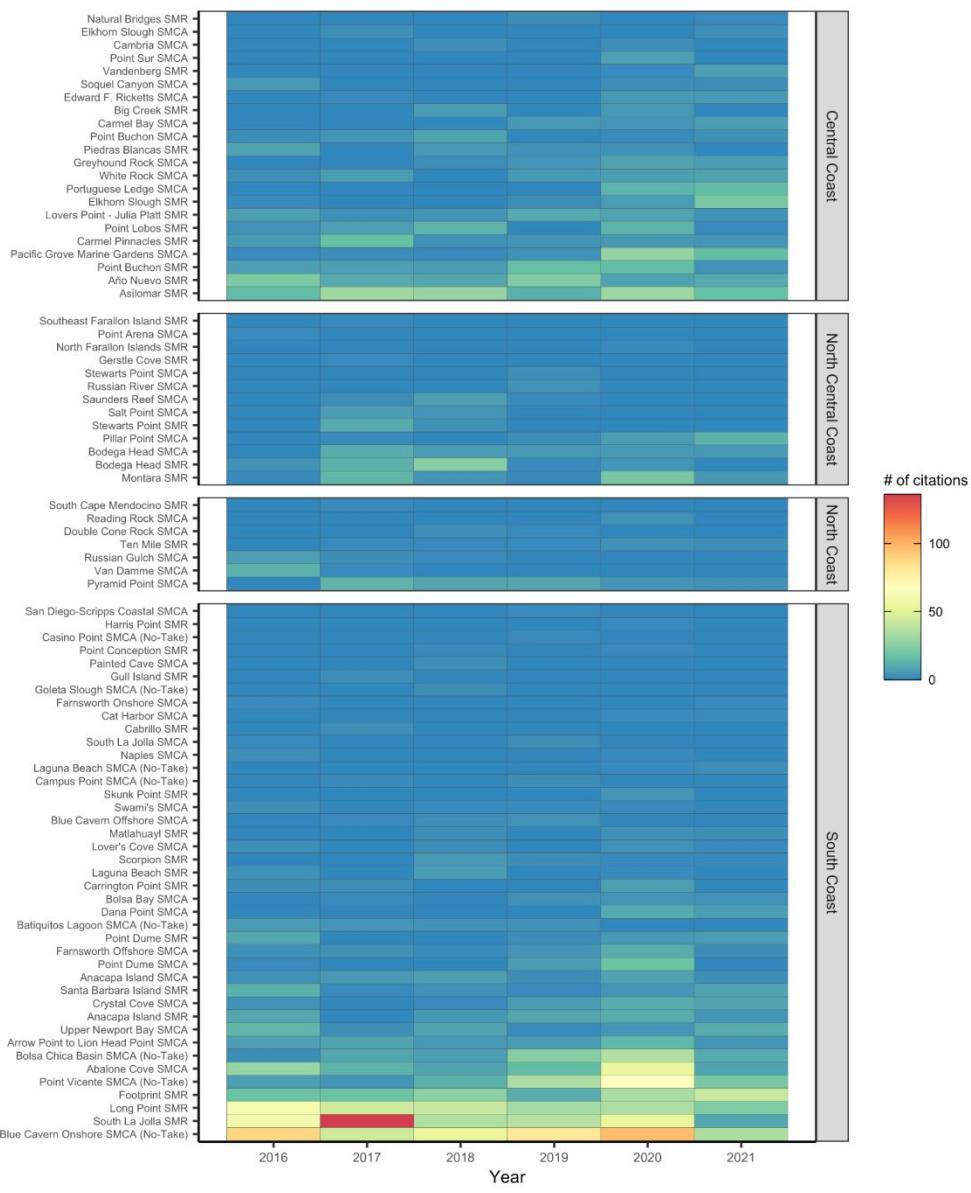
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1083 **Figure S13.** Engagement of recreational divers and snorkelers in the REEF citizen science
1084 survey program within California's state marine protected areas (MPAs). Panel **A** shows the
1085 number of surveys (point size) conducted in California's MPAs from 2012 through 2021 and the
1086 number of years with survey data (color ramp) for each MPA. Note log-scale in point size. Black
1087 x's mark the 83 MPAs without any REEF surveys. Dark horizontal lines delineate the four MLPA
1088 regions. Panel **B** shows the number of surveys within California's MPA network from 1994-2022
1089 by habitat type. Panel **C** shows the number of surveys within California's MPA network from
1090 1994-2022 by depth zone.



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1092 **Figure S14.** Number of scientific permits issued annually from 2012 to 2021 by marine
1093 protected area (MPA). MPAs are listed in order of overall sample size within each region.

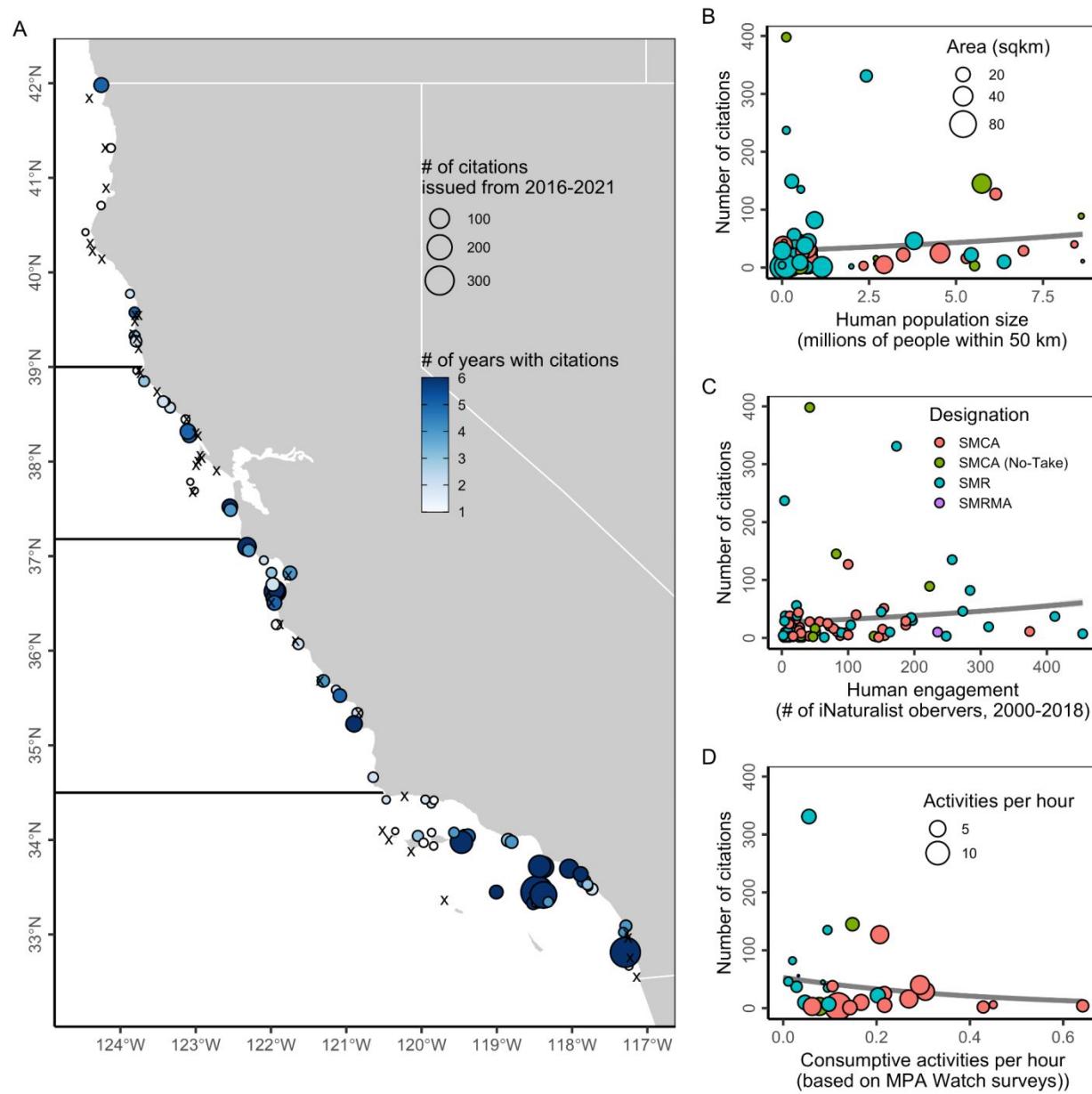


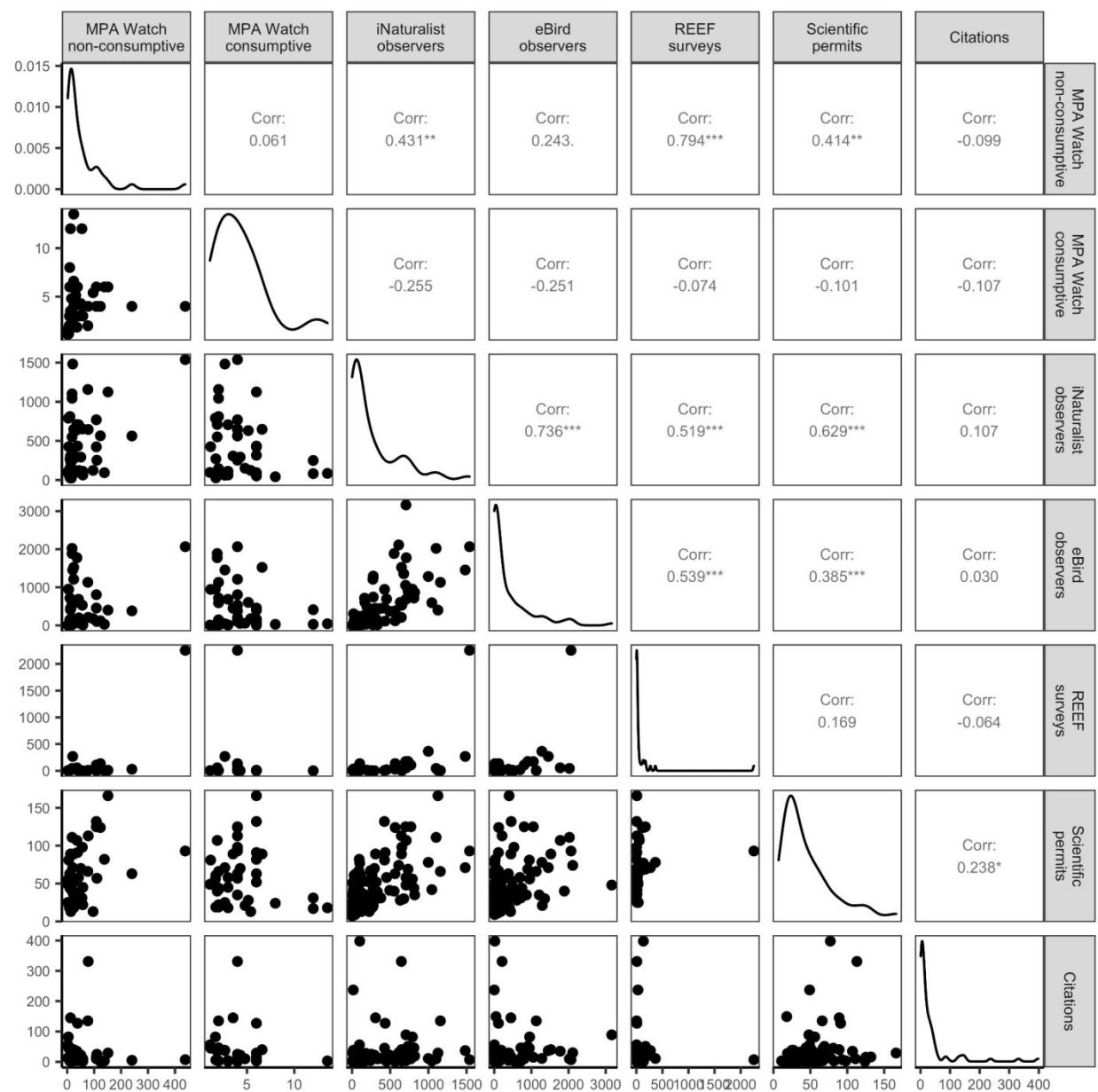
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1095 **Figure S15.** Number of scientific permits issued for research within California's state marine
1096 protected areas (MPAs) from 2012 through 2021. Panel (A) shows the number of scientific
1097 permits issued (point size) and number of years in which permits were issued (color ramp) for
1098 each MPA. Dark horizontal lines delineate the four MLPA regions. In (B), bars indicate the
1099 percentage of annual permits issued to MPAs of different designations and lines indicate the
1100 representation of MPAs of those designations in the network.



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1102 **Figure S16.** Number of citations issued by CDFW Law Enforcement for regulatory violations
 1103 occurring within California's MPAs from 2016 to 2021. MPAs are listed in order of overall
 1104 sample size within each region.

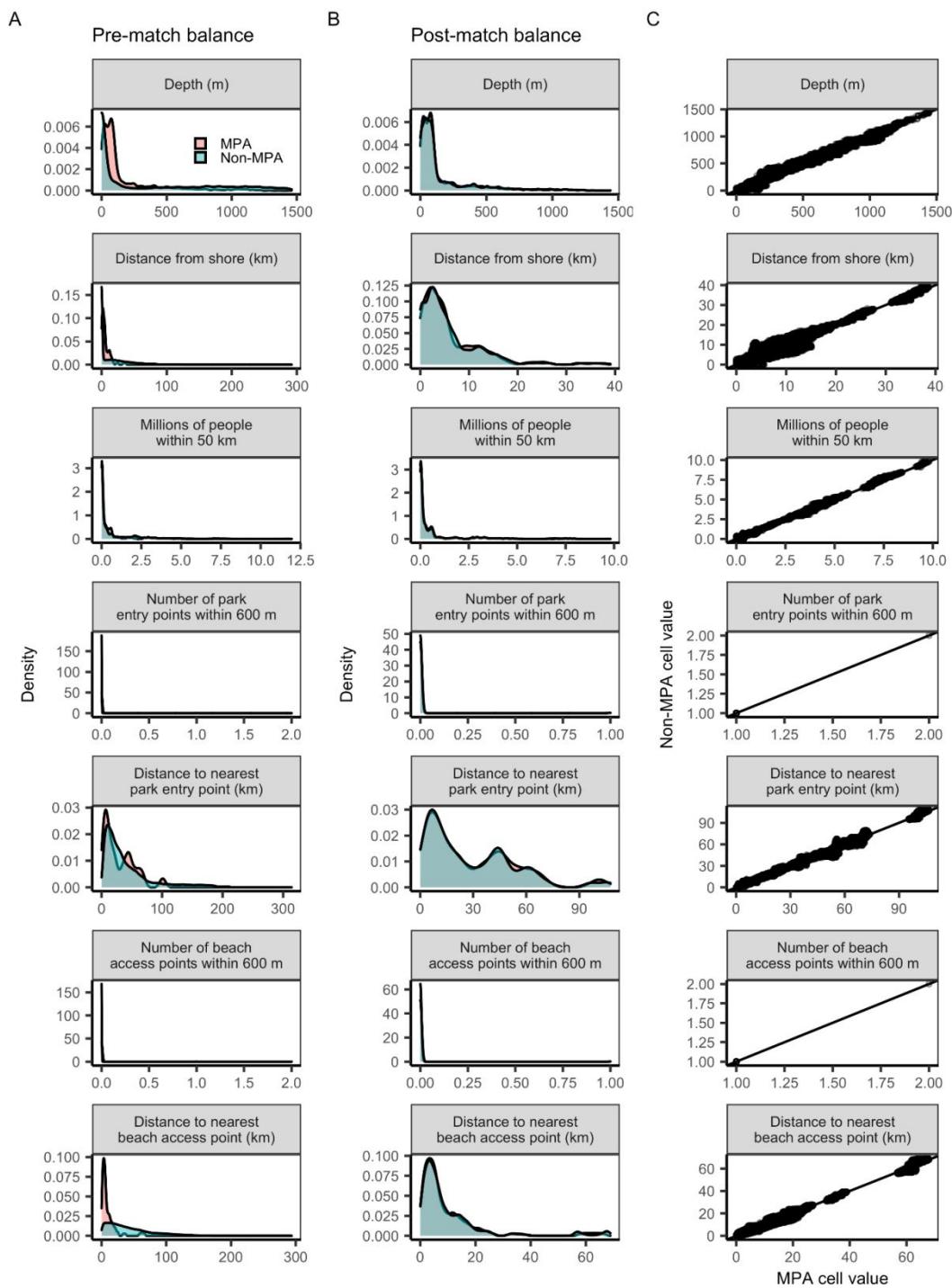




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Figure S18. Correlation between human engagement indicators. The lower section shows pairwise comparisons of engagement indicators. The upper section shows the correlation between each pairwise combination of indicators and the statistical significance of this correlation (* = $p < 0.05$; ** = $p < 0.01$; and *** = $p < 0.001$). The diagonal indicates the distribution of each engagement indicator. See **Table S3** for the choice of displayed indicator.

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Figure S19. The balance of matching variables (A) pre- and (B) post-matching and the (C) correlation between the values of MPA and matched non-MPA raster cells. In (C), the black line is the one-to-one line.

Table S1. California marine protected area (MPA) designations.

Designation	# / area (km²)	Restrictions
<i>State marine protected areas</i>	<i>124 / 2207 km²</i>	
State marine reserve (SMR)	49 / 1229 km ²	Prohibits comm/rec take of all marine resources*
State marine conservation area (SMCA)	60 / 880 km ²	Prohibits comml/rec take of selected marine resources
State marine conservation area (no take)	10 / 86 km ²	Prohibits comm/rec take of all marine resources but allows permitted activities that cause damage (e.g., dredging)
State marine recreational management area (SMRMA)	5 / 12 km ²	Limits comm/rec take of marine resources but allows legal waterfowl hunting
<i>Other state management areas</i>	<i>22 / 20 km²</i>	
State marine park (SMP) - all in SF Bay	7 / 17 km ²	Prohibits damage or commercial take of all marine resources; recreational take is allowed
State marine conservation area (SMCA) - SF Bay	1 / 0 km ²	Prohibits comml/rec take of selected marine resources
Special closure	14 / 8 km ²	Prohibits or restricts activities in waters adjacent to seabird rookeries or marine mammal haul-out sites
<i>Federal marine protected areas</i>	<i>9 / 394 km²</i>	
Federal marine reserve (FMR)	8 / 388 km ²	Extends SMRs around the Channel Islands into federal waters
Federal marine conservation area (FMCA)	1 / 6 km ²	Extends SMCAs around the Channel Islands into federal waters

* Marine resources can be living, geologic, or cultural

Table S2. Social vulnerability indicators and metrics used to calculate the social vulnerability index. Direction of influence indicates whether the metric was assumed to increase (positive) or decrease (negative) vulnerability. Data used were from the 2010 American Community Survey. Geographic unit for all data is the census tract.

Indicator	Direction of Influence
<i>Housing Characteristics</i>	
Median monthly mortgage (USD)	Negative
Median monthly rent (USD)	Negative
Median number of rooms per housing unit	Negative
Percent of all family households that live in mobile homes	Positive
<i>Labor Force Structure</i>	
Percent of age 16+ population that are working females	Negative
Percent of age 16+ population that are working	Negative
<i>Population Composition</i>	
Percent of families with single females as head of household	Positive
Percent of population age 5 and older that speak English less than “very well”	Positive
Percent of population whose race is white, with no other descriptors	Negative
Percent of the population younger than age 6	Positive
<i>Poverty Status</i>	
Percent of families living below the poverty line	Positive
Percent of people under age 18 living below the poverty line	Positive
Percent of people receiving cash assistance or SNAP payments	Positive
Percent of people age 65 and over living below the poverty line	Positive

Table S3. Indicators of human engagement evaluated in this paper. The bolded metric indicates the metric used in the scorecard and accumulation curve analysis.

Indicator and source	Description	Metrics	Limitations
MPA Watch (www.mpawatch.org)	Recreation: MPA Watch is a community science program that trains volunteers to observe and collect data on human uses of protected areas (MPA Watch, 2022b). Volunteers use a standardized survey protocol (MPA Watch, 2022a) to record consumptive (e.g., fishing) and non-consumptive (e.g., surfing, boating, tidepooling, running, etc.) activities occurring offshore and onshore of coastal sampling sites.	(1) the median number of activities observed per hour for surveys in which activities were observed (i.e., zeroes excluded); (2) percent of surveys in which an activity was observed	There is limited ability to infer the legality of the consumptive activity observed by MPA Watch volunteers because some MPAs allow some forms of harvest and MPA Watch volunteers, while well-trained, are not legal authorities on MPA boundaries and regulations.
iNaturalist (www.inaturalist.org)	Recreation/education: iNaturalist is a web- and app-based platform that allows observers to submit wildlife photos for identification by amateur and professional naturalists (iNaturalist, 2022).	(1) number of iNaturalist users who submitted observations; (2) number of submitted observations	Participation in iNaturalist largely depends on smartphone ownership and fluency and likely represents a biased (younger, wealthier) demographic.
eBird (www.ebird.org)	Recreation/education: eBird is a global citizen science program that collates observations of birds submitted by birdwatchers (eBird, 2022).	(1) number of eBird users who submitted observations; (2) number of submitted observations	Participation in eBird is eased by smartphone ownership and fluency and likely represents a biased (younger, wealthier) demographic.
REEF (www.reef.org)	Recreation/education: REEF is an international marine conservation organization that trains volunteer divers and	(1) number of surveys conducted; (2) number of years in which a survey was	The REEF Program is based in southern California and trainings are largely conducted

	snorkelers to collect and report information on marine fish and selected invertebrate and algae species (REEF, 2022).	conducted	in this area. As a result, participation is greatly biased towards southern California.
Scientific permits (CA Dept. Fish & Wildlife)	Scientific research: Permits issued by CDFW for scientific research provide an indicator of scientific research activity throughout California's MPA network.	(1) number of permits issued; (2) number of years in which permits were issued.	There are no apparent limitations to this dataset. However, we note that there are more scientific institutions in Southern/Central CA than Northern CA and that conditions for fieldwork are worse in Northern CA.
Law enforcement citations (CA Dept. Fish & Wildlife)	Non-compliance: Regulatory citations from CDFW's Law Enforcement Division provide an indicator of where non-compliance occurs throughout California's MPA network.	(1) number of citations issued; (2) number of years in which citations were issued.	The lack of effort information (e.g., number of patrol hours) associated with the citation frequency data limits ability to infer non-compliance rates. For example, high frequency could reflect either greater enforcement activity or greater illegal activity.

Table S4. Human use activities recorded by MPA Watch volunteers.

Activity type	Activity	Subcategories
<i>Non-consumptive</i>		
Onshore recreation	Beach recreation	sandy/rocky
Onshore recreation	Wildlife watching	sandy/rocky
Onshore recreation	Domestic animals	sandy/rocky; on/off leash
Onshore recreation	Driving on the beach	
Onshore recreation	Tide-pooling	
Offshore recreation	Offshore recreation (e.g. swimming)	
Offshore recreation	Surfing	
Offshore recreation	Other board sports	
Offshore recreation	SCUBA diving or snorkeling	
Boating	Kayaking	
Boating	Paddleboarding	
Boating	Other paddleboating	
Boating	Sailing	
Boating	Windsurfing	
Boating	Jet skiing	
Boating	Power boating	
Boating	Whale watch boat	
Boating	Dive boat	
Boating	Work boat	
Boating	Law enforcement boat	
Boating	Other boating	
<i>Consumptive</i>		
Fishing	Hand collection of biota	sandy/rocky
Fishing	Hook and line fishing	onshore/boat; sandy/rocky (if shore); rec/comm/unknown (if boat); active/inactive (if boat)
Fishing	Trap fishing	onshore/boat; sandy/rocky (if shore); rec/comm/unknown (if boat); active/inactive (if boat)
Fishing	Net fishing	onshore/boat; sandy/rocky (if shore); rec/comm/unknown (if boat); active/inactive (if boat)
Fishing	Spear fishing	onshore/offshore/boat; sandy/rocky (if shore); rec/comm/unknown (if boat); active/inactive (if boat)
Fishing	Dive fishing	offshore/boat; rec/comm/unknown (if boat); active/inactive (if boat)
Fishing	Kelp harvesting	active/inactive
Fishing	Passenger fishing	active/inactive/unknown
Fishing	Unknown fishing	

Table S5. Sources of explanatory variables included in logistic regressions evaluating traits associated with charismatic and underutilized MPAs.

Variable	Source
<i>Design feature</i>	
MPA age (yr)	CDFW (2019) MPA GIS file. Available at: https://filelib.wildlife.ca.gov/Public/R7_MR/MANAGEMENT/MPA/
MPA size (km ²)	CDFW (2019) MPA GIS file. Available at: https://filelib.wildlife.ca.gov/Public/R7_MR/MANAGEMENT/MPA/
Protection status (no-take, some take)	CDFW (2019) MPA GIS file. Available at: https://filelib.wildlife.ca.gov/Public/R7_MR/MANAGEMENT/MPA/
<i>Habitat type</i>	
Sandy beach (km)	CDFW MPA habitat mapping
Rocky intertidal (km)	CDFW MPA habitat mapping
Estuary (km)	CDFW MPA habitat mapping
Maximum kelp canopy (km ²)	CDFW MPA habitat mapping
<i>Accessibility and amenities</i>	
Distance to port (km)	CDFW (2022) Fishing ports. Available from CDFW.
Number of parks within 1 km	ESRI (2022) USA Parks. Available at: https://www.arcgis.com/home/item.html?id=578968f975774d3fab79fe56c8c90941
Number of parking lots within 1 km	CDPR (2022) Parking. California Department of Parks and Recreation. Available at: https://www.parks.ca.gov/?page_id=29682
Number of picnic areas within 1 km	CDPR (2022) Picnic Grounds. California Department of Parks and Recreation. Available at: https://www.parks.ca.gov/?page_id=29682
Number of campgrounds within 1 km	CDPR (2022) Campgrounds. California Department of Parks and Recreation. Available at: https://www.parks.ca.gov/?page_id=29682
<i>Social vulnerability</i>	
Social vulnerability index	See methods for details.

Table S6. Matching variables used in the design of counterfactual areas and their sources.

Matching variable	Source
Depth (m)	CDFW (2011) Bathymetry. California Department of Fish and Wildlife. Available at: https://filelib.wildlife.ca.gov/Public/R7_MR/BATHYMETRY/
Distance from shore (km)	CDFW (2011) Coastline. California Department of Fish and Wildlife. Available at: https://filelib.wildlife.ca.gov/Public/R7_MR/BASE/
Number of people within 50 km	USCB (2010) US Census Data. United States Census Bureau. Available at: https://www.census.gov/programs-surveys/decennial-census/decade/2010/about-2010.html
Number of park entry points within 600 m	CDPR (2022) Park Entry Points. California Department of Parks and Recreation. Available at: https://www.parks.ca.gov/?page_id=29682
Distance to nearest park entry point (m)	CDPR (2022) Park Entry Points. California Department of Parks and Recreation. Available at: https://www.parks.ca.gov/?page_id=29682
Number of public beach access points within 600 m	CCC (2022) Public Access Points. California Coastal Commission. Available at: https://gis.data.ca.gov/datasets/coastalcomm::public-access-points/about
Distance to nearest public beach access point (m)	CCC (2022) Public Access Points. California Coastal Commission. Available at: https://gis.data.ca.gov/datasets/coastalcomm::public-access-points/about

Table S7. Attributes of ‘charismatic’ and ‘underutilized’ MPAs by type of engagement, based on the results of stepwise logistic regressions. Missing values indicate the best fit model does not include the associated predictors*. In each model, “typical” MPAs were set as the reference level and evaluated against charismatic or underutilized MPAs. Coefficients returned by each model are reported as odds ratio. CI = 95% confidence interval; AIC = Akaike Information Criterion.

Predictors	Charismatic vs. typical			Underutilized vs. typical		
	Odds Ratios	CI	p	Odds Ratios	CI	p
(Intercept)	0.00	0.00 – 0.13	0.007	0.62	0.24 – 1.53	0.302
Distance to port (km)	1.00	1.00 – 1.00	0.065	1.00	1.00 – 1.00	<0.001
MPA size (km ²)	0.94	0.87 – 1.01	0.121			
Take? (yes/no)	0.26	0.05 – 1.18	0.093			
Sandy beach (km)	1.49	1.08 – 2.19	0.022	0.61	0.39 – 0.87	0.016
MPA age (yr)	1.58	1.15 – 2.29	0.007			
# of parks within 1 km	1.28	1.09 – 1.56	0.006			
Rocky intertidal (km)				0.80	0.61 – 1.03	0.101
# of parking lots within 1 km				0.42	0.15 – 0.71	0.019
Observations	71			92		
R ² Tjur	0.466			0.446		
AIC	59.527			84.254		

* Predictors not included in the reduced models include: maxim kelp canopy (km²), estuary extent (km), number of campgrounds within 1 km, number of picnic areas within 1 km (see **Table S5** for details).

If you build it, they will come: coastal amenities facilitate human engagement in marine protected areas

Marine protected areas (MPAs), areas in the ocean where fishing is prohibited or restricted, are commonly used as tools to protect biodiversity, recover fisheries, and promote other beneficial human experiences. While the conservation and fisheries impacts of MPAs have been well studied, the impacts of MPAs on other dimensions of human use -- such as recreation, education, and scientific research -- have received less attention. Identifying traits of MPAs that promote or limit human engagement is critical to designing MPA networks that achieve multiple goals effectively, equitably, and with minimal environmental impact.

In our recent paper, we develop a novel and transferable framework for quantifying human engagement in California's MPA network, one of the largest MPA networks in the world. We assemble and compare diverse indicators of human engagement -- leveraging information from citizen science programs, social media platforms, and government datasets -- that capture recreational, educational, and scientific activities across California's MPAs.

We find that human engagement is correlated with local population density: unsurprisingly, the more people that live close by, the more people that engage in an MPA. However, we also find that MPAs near tourist destinations, adjacent to state parks and their amenities, and with long sandy beaches generate more engagement than would be expected based on population density alone. Conversely, remote MPAs without sandy beaches or parking lot access had lower than expected human engagement.

What does this mean as the world aims to expand MPA coverage to protect 30% of the ocean by 2030? On one hand, human engagement can be promoted by developing land-based amenities that increase access to coastal MPAs or by locating new MPAs near existing amenities during the design phase. On the other hand, human engagement can be limited by locating MPAs in areas far from population centers, coastal amenities, or sandy beaches. This choice depends on management goals. Our paper provides a transferable framework for current and future MPA networks to track progress towards meeting their own human use objectives.



Kayakers exploring the Matlahuayl State Marine Reserve off of La Jolla Cove in San Diego, California, USA. Photo by Jacob Eurich.

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We develop a framework for measuring recreational, educational, and scientific engagement in California's marine protected area network. We find high engagement in MPAs close to population centers, tourist destinations, state parks, and sandy beaches. @ChrisFree14
@PISCOScience

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