

1 Inter-specific relationships and their ecological role in an oceanic
2 elasmobranch community

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17 **Abstract**

18

19

20 Marine ecosystems support a diverse array of co-occurring species, whose presence and abundance
21 influence the behavior, population dynamics, and distribution of interacting organisms. Elasmobranchs
22 play a central role in marine ecological processes as top and meso-predators across various ecosystems.
23 Previous work has shown that some elasmobranchs are key to ecosystem health and resilience, regulating
24 communities through competitive, predatory, and territorial interactions. However, few studies have
25 evaluated the effect of interspecific interactions on the abundance and diversity of individual populations
26 in marine ecosystems, partly due to the difficulty of obtaining long-term data on species co-occurrence.
27 Our objective was to analyze elasmobranch species co-occurrence in Cocos Island National Park in the
28 Eastern Tropical Pacific and its effect in species abundance, diversity, and interactions within the
29 community. We analyzed a 26-year underwater visual survey dataset through network statistics, diversity
30 indices, and Bayesian species interaction models. Here we show that the elasmobranch community in
31 early 1990s shifted from a sparse, low richness network to a more diverse, densely connected community
32 in the late 2010s, with diversity peaking in 2006. We identified three types of species interactions within
33 the elasmobranch community: competition, predation, and ecosystem preference. The Scalloped
34 hammerhead (*Sphyrna lewini*) and Whitetip reef (*Triaenodon obesus*) sharks were the main contributors
35 to year-to-year variations in diversity. In addition, most species had a negative relationship with the co-
36 occurrence of other elasmobranchs, underscoring the importance of potential competitive and predatory
37 interactions. Our findings offer valuable insights for marine conservation efforts, particularly for
38 understanding how interspecific interactions and ecological processes influence ecosystem stability over
39 time. Our study provides a framework for future research on species co-occurrences in marine

40 ecosystems, emphasizing the need for integrated, interaction-focused models that can better predict
41 ecosystem responses to threats like overfishing and environmental change.

42

43 Resumen

44 Los ecosistemas marinos sostienen un diverso conjunto de especies, cuya abundancia afecta el
45 comportamiento, dinámica poblacional y distribución de los organismos con los que interactúan. Los
46 elasmobranquios juegan un papel central en los procesos ecológicos marinos al ser meso-depredadores y
47 depredadores tope en diversos ecosistemas. Estudios demuestran la importancia de los elasmobranquios
48 para la salud y resiliencia de los ecosistemas, regulando a las comunidades marinas mediante procesos de
49 competencia, depredación y territorialidad. No obstante, pocos estudios han evaluado el efecto que las
50 interacciones interespecíficas ejercen sobre la abundancia y diversidad de las poblaciones en ecosistemas
51 marinos, debido en parte a la difícil tarea de recolectar datos de coocurrencia de especies a largo plazo.
52 Nuestro objetivo fue analizar la coocurrencia de elasmobranquios en el Parque Nacional Isla del Coco en
53 el Pacífico Oriental Tropical, y su efecto en la abundancia, diversidad e interacciones de las especies en su
54 comunidad marina. Analizamos 26 años de datos obtenidos de censos visuales submarinos a través de
55 estadísticas de redes, índices de diversidad y modelos interactivos Bayesianos. Demostramos que la
56 comunidad de elasmobranquios pasó de ser una red dispersa con riqueza más baja en los 1990s, a una
57 comunidad más densamente conectada y diversa en los 2010s, con un poco de diversidad en 2006.
58 Identificamos tres tipos de interacciones entre especies en la comunidad de elasmobranquios:
59 competencia, depredación y preferencias ecosistémicas. Los tiburones martillo común (*Sphyrna lewini*) y
60 puntas blancas de arrecife (*Triaenodon obesus*) fueron los principales contribuyentes a la variación anual
61 de diversidad. Adicionalmente, la mayoría de las especies mostró una relación negativa con la
62 coocurrencia de otras especies de elasmobranquios, resaltando la importancia de las potenciales
63 interacciones competitivas y de depredación. Nuestros descubrimientos ofrecen una línea base para
64 futuros estudios sobre la coocurrencia de especies en ecosistemas marinos, y enfatizan la necesidad de
65 producir modelos integrados basados en interacciones para predecir de mejor manera la respuesta de los
66 ecosistemas a amenazas como la sobrepesca y el cambio climático.

67 Introduction

68 Marine ecosystems are shaped by complex biotic interactions, where processes like diel migrations and
69 seasonal shifts influence species' behavioral and physiological responses to their competitors (Lear et al.,
70 2021) as well as ecosystem structure (Carrier et al., 2012). However, evaluating competitive, predatory
71 and ecologically driven species interactions is challenging due to logistic and methodological constraints,
72 especially for large marine predators whose interactions may occur infrequently and across large space
73 and time scales. Additionally, species interactions are complex on their own, occurring through direct
74 pathways such as predation, and indirect mechanisms like competition and predator mediated effects
75 (Twining et al., 2024). Species co-occurrence models and indirect network analysis have emerged as a
76 powerful tool for capturing relevant ecological properties such as trophic and competitive interactions
77 (Russo et al., 2022). By evaluating the relationship between co-occurring species and community
78 composition, we can indirectly provide evidence for complex ecological processes that shape species
79 interactions and affect community resilience (Frank et al., 2005). Understanding the effects of species co-
80 occurrence is crucial for evaluating the ecological health of systems providing services to stakeholders,

81 and the economic stability of said ecosystem services (Frank et al., 2005). Similarly, local processes
82 mediated by behavioral shifts, predation and competitive interactions have been shown to affect national
83 and local stakeholders' economies (Dill et al., 2003; Gregr et al., 2020). Despite the nuanced process of
84 factoring species interactions into management models, biologically appropriate ecological models that
85 factor species abundance and co-occurrence can shed light over complex ecosystems (Twining et al.,
86 2024), and provide better tools for mitigating economic impacts for local and international stakeholders
87 (Ferretti et al., 2020; Frank et al., 2005).

88 As top predators and meso-predators, elasmobranchs play critical roles in regulating marine ecosystems
89 through direct and indirect interactions with their prey and competitors (Dedman et al., 2024; Flowers et
90 al., 2021; Lear et al., 2021). For example, A study in Western Australia demonstrated behaviorally
91 mediated niche shifts in dugongs (*Dugong dugon*) following the arrival of tiger sharks (*Galeocerdo cuvier*)
92 (Dill et al., 2003). The presence of tiger sharks also displaced non-prey competitors, such as dolphins, from
93 otherwise attractive foraging areas. Similarly, a study of six shark species in the Gulf of Mexico revealed
94 local changes mediated by diel patterns of site use by sharks, as well as evidence of larger species such as
95 tiger and bull sharks (*Carcharhinus leucas*) excluding smaller species such as blacktip sharks (*Carcharhinus*
96 *limbatus*) (Lear et al., 2021). Batoid species also play an important role in ecosystem function as meso-
97 predators and benthic feeders (Flowers et al., 2021). Ajemian et al. (2012) evaluated the impacts of an
98 increase in spotted eagle ray (*Aetobatus narinari*) population in Bermuda. Authors found shifts in
99 preferred mollusk prey items according to eagle rays' size and prey density, suggesting moderate impacts
100 on restoration programs caused by an increased number of rays, and underscoring the need for further
101 interaction focused studies. Addressing gaps in the analysis of multi-species networks, predator co-
102 occurrence, and competition-driven community shifts is crucial for understanding and managing complex
103 marine systems (Hollowed et al., 2000; van Zinnicq Bergmann et al., 2022).

104 The Eastern Tropical Pacific (ETP), is a vast ecoregion which extends along the Pacific Coast from southern
105 Mexico to northern Peru, hosting a diverse community of co-occurring elasmobranch species (Navia et
106 al., 2024). Cocos Island National Park (hereafter Cocos Island) is unique within the ETP for its systematic
107 long-term monitoring of elasmobranch communities. As one of the world's oldest oceanic marine
108 protected areas (MPAs), Cocos Island plays a vital role in Costa Rica's cultural, economic, and natural
109 heritage (Salas et al., 2012). Despite its small size of 28.8 km², Cocos Island is recognized globally for its
110 rich biodiversity, serving as a critical hotspot (Moreno et al., 2021) for marine endangered species and
111 migratory macropredators (Klimley et al., 2022; Nalesso et al., 2019). However, the elasmobranch
112 community at Cocos Island has undergone population shifts beyond the expected seasonal variability in
113 the area during recent years. For example, White et al. (2015) observed changes in elasmobranch
114 population dynamics in shallow waters, driven by both climatic and oceanographic factors. The authors
115 noted that although environmental conditions were important for most species, others were increasing
116 in number regardless of yearly variations. Osgood et al. (2021) further addressed the changes in Cocos
117 elasmobranch community by suggesting that acute temperature anomalies may be pushing more mobile
118 and migratory species away from the island. Abundant species, such as scalloped hammerhead sharks and
119 mobula rays, exhibited strong but inconsistent responses to temperature changes, indicating the potential
120 influence of additional, unidentified factors such as seasonal patterns and biotic interactions. Saltzman
121 and White (2023) on the other hand, examined environmental and biological drivers of filter feeder
122 abundance at Cocos Island, finding that while some species responded strongly to temperature shifts,
123 other species were only vaguely associated with factors such as primary productivity. Thus, recent

124 research on Cocos Island elasmobranch communities underscores the need to explore alternative sources
125 of variation in both seasonal and yearly community dynamics.

126 Despite continued interest in Cocos Island as a case study for elasmobranch community processes,
127 research on competition, niche partitioning, and network connectivity within its predator populations
128 remains limited. While some studies highlight the role of competition and niche partitioning in the Eastern
129 Tropical Pacific (Estupiñán-Montaña et al., 2017, 2024), the ecological and behavioral effects of species
130 co-occurrence at Cocos Island remain largely unexplored. Evidence suggests that certain species
131 experience behaviorally mediated niche compression in the presence of larger predators and competitors
132 (Espinoza et al., 2024), yet the broader impacts of predator and competitor exclusion on community
133 structure are not well understood. Notably, Saltzman et al (2024) examined the changes in Cocos Island
134 marine communities caused by the arrival or establishment of large predators such as Galapagos and Tiger
135 sharks that were relatively uncommon or absent in the 1990s. The authors reported a 43% decline in the
136 relative abundance of migratory sea turtles, an effect likely driven by tiger shark predation, but
137 accentuated due to illegal fishing and external migration-related mortality. Further analysis of predator
138 co-occurrence and network dynamics is essential to better understand the ecological processes shaping
139 Cocos Island's elasmobranch community.

140 In this study, we investigate the network structure and shifts in diversity of the elasmobranch community
141 in Cocos Island, Costa Rica. We also analyzed the effects of elasmobranch species co-occurrence in Cocos
142 Island and its effect within the community. We hypothesize our analysis will allow us to infer which species
143 interactions are taking place, and if these interactions could be caused by competition between species,
144 predation, or as a response to environmental preferences. Our specific objectives were to: 1) evaluate
145 ecological network and biodiversity metrics using a long-term database (1993-2019) of the elasmobranch
146 community; 2) assess changes in seasonal trends in the elasmobranch community of Cocos Island; and 3)
147 determine the effect of including species interactions as a covariate in ecological models. Given the range
148 of strategies elasmobranchs have developed for resource partitioning and the narrow trophic niches these
149 organisms occupy, we hypothesized that the abundance and community shifts of less dominant and
150 migratory species would be most affected by the relative abundance of competing or dominant species.

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Materials and methods

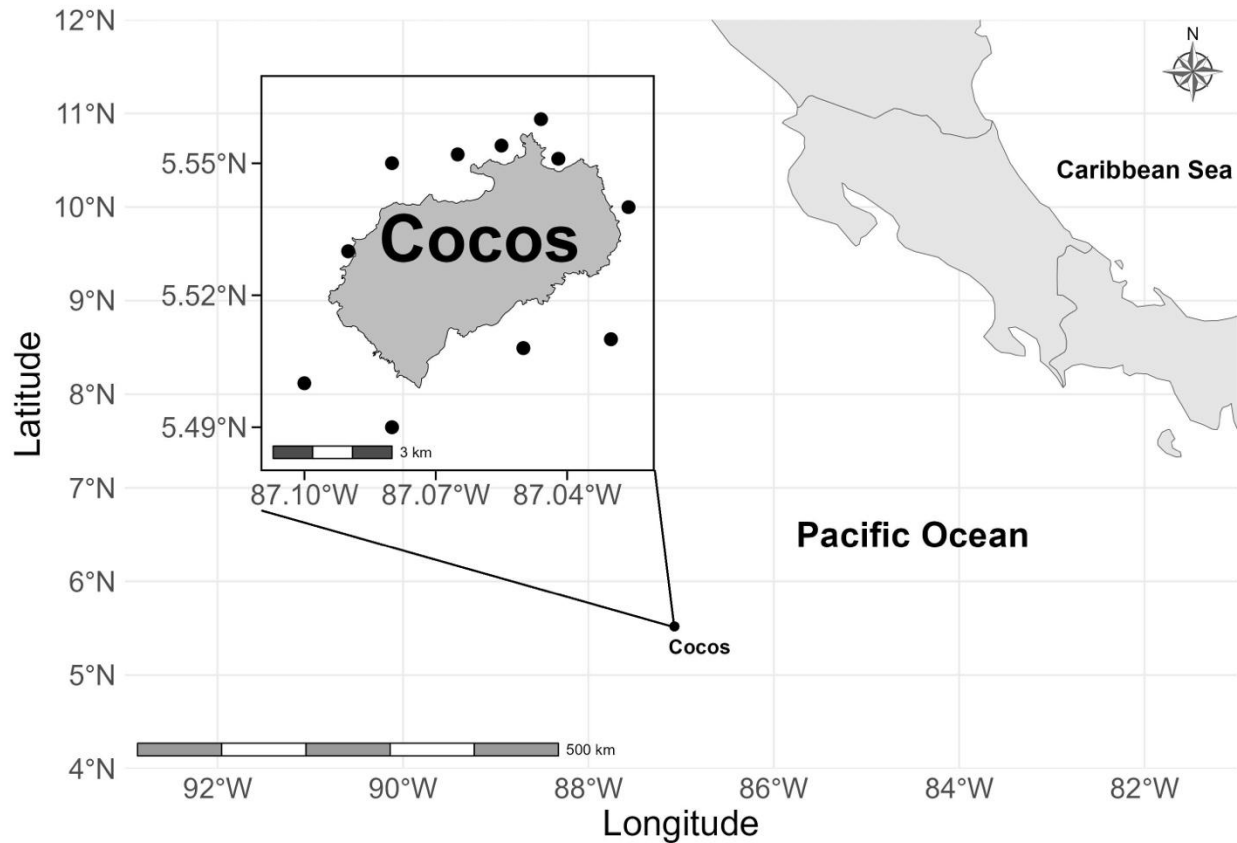
153 Cocos Island National Park (N 05°31'08", W 87°04' 18") is a small (28.8 km²) uninhabited Island located
154 550 km off Costa Rica. Cocos Island experiences two seasons, wet from June to November and dry from
155 December to May. The Island is a hotspot for marine biodiversity thanks to its isolation, complex
156 geomorphology and the influence of several sea currents (Garrison, 2006). Cocos Island is a tropical
157 marine environment with waters ranging from 24 to 30 °C in sea surface temperature, and experiences
158 thermal anomalies such as El Niño Southern Oscillation (ENSO) caused by the strengthening and
159 weakening of oceanic currents (Osgood et al., 2021). Being a National Park, Cocos Island is an attractive
160 site for wildlife tourism, with several companies taking more than 1900 tourists to diving trips to the island
161 most of the year (Moreno et al., 2021). Despite the flourishing tourism industry in the island and its long-
162 standing status as a protected national park, illegal fishing activities and proper law enforcement remain
163 as the main conservation challenges for the area (Arias et al., 2016).

164 We used underwater visual census data collected from 1993 to 2019 by experienced diving guides (n =
165 53) from the Undersea Hunter Group (underseahunter.com), resulting in a total of 35,706 dives at 17 sites
166 around Cocos Island (Figure 1). This dataset is one of the largest and longest-running underwater visual
167 censuses (UVC) for elasmobranchs (Osgood et al., 2021). Average dive time was approximately 40
168 minutes, with depth ranging from 25 to 32 m, remaining consistent over time for each site despite not
169 being standardized as scientific UVCs. Divers used a standardized data sheet to record observed numbers
170 of the most common elasmobranch species. Species were recorded as either the presence/absence or
171 counts of 12 species of elasmobranchs including filter feeders, meso-predators and top-predators
172 (Supplementary Table 1). We merged species of the genus *Mobula* in a single category due to their spatial
173 and temporal overlap excluding manta rays (*Mobula birostris*). Additionally, we included sea turtles
174 (*Chelonia spp.*) in our analysis due to their role as prey for large shark species. The database was manually
175 reviewed to correct transcription errors. Blacktip (*Carcharhinus limbatus*), Galapagos (*Carcharhinus*
176 *galapaguensis*), and tiger sharks (*Galeocerdo cuvier*) were initially recorded only as observational notes,
177 with systematic counts introduced later (Supplementary Figure 1). While some human error (e.g., double
178 counting) may have occurred, such bias remained consistent throughout the survey period. Models
179 potentially affected by this issue account for it through offset terms. Although sampling could be
180 considered a form of citizen science, all UVCs were conducted by experienced dive guides, ensuring high
181 data quality. Sensitivity analyses were performed when needed, and the shift in data collection reflects
182 changes in community composition at Cocos Island over time.

183 If more than one diver observed a species during a day, we averaged counts for that species across all
184 involved divers. We used the total number of dives in a day as an offset term in our models to account for
185 sampling effort (Kéry & Schaub, 2012). Visual censuses occurred within sites relatively close to one
186 another (n=17); therefore, we accounted for observations amongst all sites by merging observations of
187 all sites around Cocos Island (Supplementary figure 6). Site-by-site species counts were treated separately
188 for analysis in which site variation could be important.

189

190



191

192 *Figure 1. Study site: Cocos Island), Costa Rica (CR). The inset plot shows the actual contour of Cocos Island and its coordinates;*
 193 *points show the approximate location of dive sites.*

194

195 **Network Analysis**

196 We conducted a network analysis to examine the inter-specific interactions of the elasmobranch
 197 community in Cocos Island from 1993 to 2019. Our analysis was based on species detection and non-
 198 detection data aggregated for each day. We represented species as nodes, node size as species
 199 abundance, and edges between nodes as the co-occurrence of species within the same period. We
 200 quantified the weight of each edge as the number of days when two species cooccurred within a given
 201 year. We then calculated network strength as the weighted number of connections between nodes.
 202 Finally, we built separate networks for each of the observed seasons in Cocos Island (wet from June to
 203 November and dry from December to May) to account for climatic and seasonal variability in the region
 204 for each year (for example, a network for the dry season in 1993, wet season in 1993, dry in 1994 and so
 205 on).

206 We used the iGraph (Csardi & Nepusz, 2006) package and its dependencies in the R programming
 207 environment (R Core Team, 2016) for statistical analysis and network construction. We performed
 208 centrality and betweenness analysis to identify key species within the network and highlight shifts in the
 209 community over time. We used this information to build graphical representations of the nodes and edges

210 in the network. We tested significant changes in network metrics between years and seasons using simple
211 linear regression models.

212 Additionally, we analyzed the variation in network metrics on a site-by-site basis amongst the 17 most
213 popular dive sites around Cocos Island (Supplementary figure 6). We repeated the analysis according to
214 the mean number of individual species detected on each site used by divers in Cocos Island looking for
215 deviations to the general network patterns detected in Cocos Island as a whole, producing many individual
216 networks for each site-season-year combination.

217

218 **Diversity Analysis**

219 We conducted diversity and similarity analyses using the vegan package (Oksanen et al., 2022) in R. We
220 calculated the Hill transformed Shannon Diversity Index on a yearly and monthly basis for the study
221 period, aiming to assess temporal changes in species diversity. This index provides a measure of species
222 diversity by accounting for both species richness and evenness, presenting values as easily interpretable
223 effective numbers of species. We analyzed the statistical significance of yearly and monthly variations in
224 diversity by using simple linear regression models.

225 We calculated Bray-Curtis dissimilarity matrices to quantify differences in species composition over time
226 for the entire dataset. We generated separate matrices for our data aggregated by year and season,
227 allowing us to examine both annual and seasonal variation. We applied an Analysis of Similarities
228 (ANOSIM) to these matrices to evaluate whether significant temporal changes in species composition
229 occur across years and between months.

230 To further explore the drivers of community composition changes, we performed a Similarity Percentage
231 (SIMPER) analysis on the Bray-Curtis dissimilarity matrices. This analysis identifies the contribution of
232 individual species to the overall dissimilarity between groups of paired years, highlighting which species
233 are the primary drivers of community change over time (Clarke, 1993).

234

235 **Interactive Abundance Models**

236 Our network and diversity analysis examined interspecific interactions as a function of the joint presence
237 of elasmobranch species. Although useful, indirect network analysis does not consider how the
238 occurrence of a species or group of species affects the abundance of the others. Therefore, we also
239 analyzed our data with hierarchical regression models “Interactive Abundance Models” using a Bayesian
240 framework. We grouped data by week to make this analysis computationally tractable.

241 We performed our analysis with the package r2jags (Su & Yajima, 2015) in R. This package interfaces the
242 R coding language with JAGS, a Bayesian analysis software (Plummer, 2017). We ran 10,000 iterations
243 with a 1,000 iteration burn-in and three chains to obtain model convergence (Zuur et al., 2008). We did
244 not apply thinning to our models since it is computationally inefficient and unnecessary for ecological
245 models (Link & Eaton, 2012). Before running our models, we applied a correlation analysis among our
246 model co-variates (each species, year, weeks, sea surface temperature and visibility, supplementary table
247 2) and merged covariates over the generally accepted correlation cutoff of 0.7 (Dormann et al., 2013). We
248 built separate models for each independent species with diffuse priors as slope and intercept terms, and

249 environmental variables as well as the abundance of every other species as covariates. We considered the
250 effect of a covariate significant if the 95% credible interval did not cross zero, indicating a consistent result
251 across posterior samples (McElreath, 2015). Although our output effects were correlative in nature, we
252 will be referring to our species-to-species covariate effects as “interactions” for simplicity.

253 For most species, abundance was modeled through a negative binomial distribution expressed as a
254 Poisson model with a diffuse over dispersion term ϵ and diffuse priors for each covariate. This distribution
255 was chosen to address the over dispersion caused by the contrasts between counts for rare and abundant
256 species (Zuur et al., 2008). The base equation for this model was:

257

$$258 \text{ Counts}_i \sim \text{Poisson}(\lambda_i)$$

259 Where λ_i represents the expected counts during the i th week, according to the following equation:

$$260 \log(\lambda_i) = \alpha + \beta_{week} * \text{WeekCyclic}_i + \beta_{visibility} * \text{Visibility}_i + \beta_{SST} * \text{SST}_i + \beta_{Year} * \text{Year}_i + \beta_j * \\ 261 \text{Count}_{ji} + \log(\text{offset}_i)$$

262 Weeks represented as “WeekCyclic” were transformed using a sin-cosin cyclic transformation to account
263 for seasonal variation. We chose to only include sea surface temperature (SST) and visibility as
264 environmental variables since they have previously been demonstrated to have an impact on the relative
265 abundance of elasmobranchs in the area (Osgood et al., 2021; White et al., 2015). We included the counts
266 of other species as covariates in the model though a log link function in λ_i for each i th weekly observation
267 for each j species, where offset_i equals the number of dives performed for each i th week. α and β_{week}
268 are the slope and intercept terms respectively.

269 Additional models were built for blacktips, eagle rays and manta rays, due to their low abundance. We
270 recorded observations in these models as a binomial detection/non-detection variable, adjusting them to
271 a logistic binomial model as:

272

$$273 \text{ Counts}_i \sim \text{Bernouli}(\rho_i)$$

274

275 We used a logit link function instead of a log link, otherwise, the model equation followed the same
276 formulation as the over dispersed Poisson.

277 We re-ran our models while excluding turtles as a predictor checking for significant changes in model
278 outputs to evaluate model sensitivity to species removal. Additionally. We re-ran the model using only
279 the last 5 years of data to evaluate temporal variation and reduce-sampling sensitivity.

280

281 Results

282

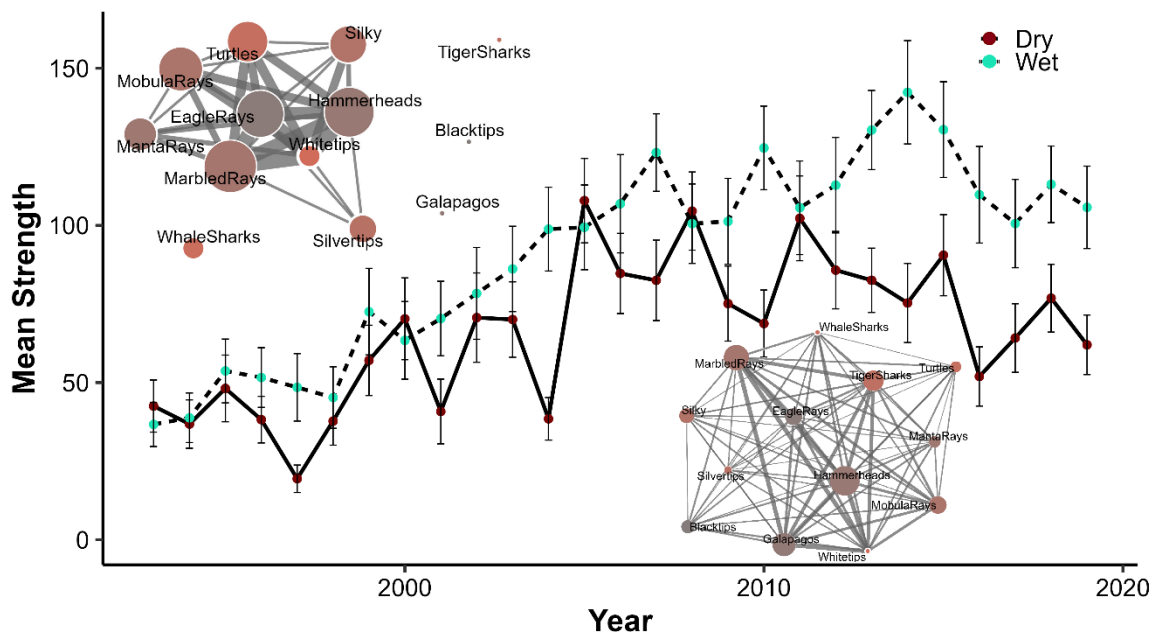
283 Network analysis

284 Our analysis showed that the elasmobranch network in Cocos Island shifted over time from fewer
 285 interactions in both seasons (prior to 2006) towards a more interconnected community (after 2006, more
 286 noticeable in recent years), forming a denser network (Figure 2). Network strength increased significantly
 287 over time and during the wet season (lowest mean strength of 19.4 + 4.36 SD in 1997 dry season, highest
 288 of 142.34 + 16.43 SD in 2014 wet season, $p < 0.01$, supplementary table 2). More species co-occurred in
 289 recent years and during the wet season, resulting in a reciprocal network where species pairs had a similar
 290 number of connections or “degree” (maximum of 24, for reciprocal connections between the 12 other
 291 species).

292 We generated a separate network (nodes and edges) for each year-season combination, but we chose to
 293 display only the two most representative networks as insets in Figure 2. The dry season of 1997 is shown
 294 as an example of the least interconnected network, while the wet season of 2014 represents the most
 295 interconnected network.

296 Network strength increased over time, with a marked rise after 2006 (Figure 2). In other words, species
 297 co-occurrences at Cocos Island became more frequent over the study period, peaking in the wet season
 298 of 2014. Prior to 2006, network strength remained relatively similar between seasons, but after this point,
 299 interactions became notably stronger during the wet season. The increased in network strength was
 300 mainly caused by the arrival and increased occurrence of blacktips, Galapagos and tiger sharks.

301



302

303 *Figure 2. Network Dynamics through time in Cocos Island Elasmobranch community. Mean network strength is represented by a*
 304 *solid line for the dry season and a dashed line for the wet season. Wisker represents standard deviation. Neural networks plots*
 305 *show the least interconnected network (1997, left) and the most interconnected network (2014, right). Each node (colored*
 306 *circles) represents a species in the network, size shows the species relative abundance. Lines between nodes (edges) represent*
 307 *species co-occurrence. Line thickness shows the relative edge weight or proportion of co-occurrence events.*

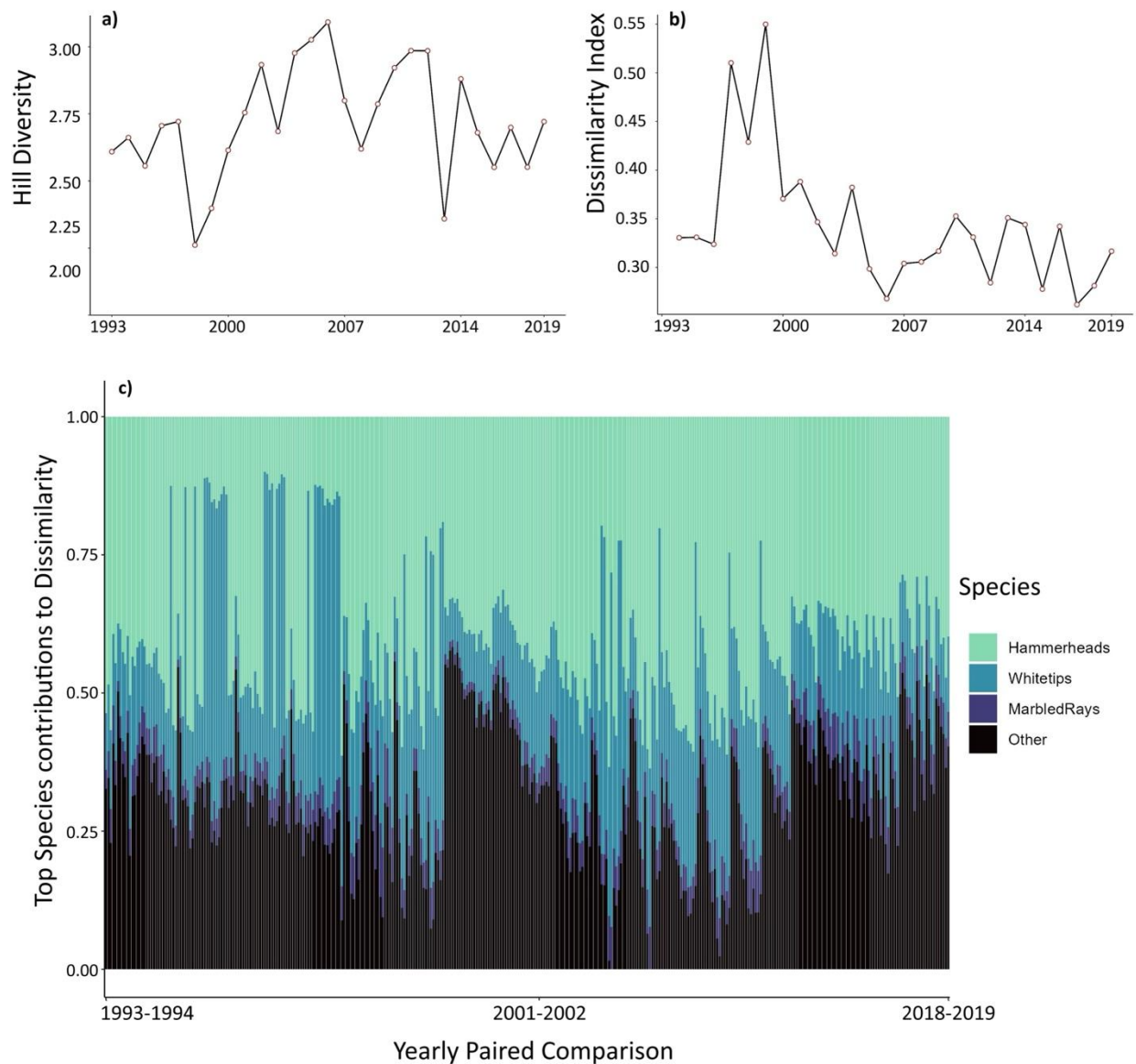
308 Site by site analysis followed a very similar (if more variable) pattern, with most sites following the same
309 yearly and seasonal trends (Supplementary figure 7). Four sites (Manuelita B, SharkfinRock, Silverado and
310 SubmergedRock) differed from the observed network patterns, showing either a decrease in network
311 strength over time or a lack of seasonal differentiation in network strength.

312 **Diversity**

313 Annual Hill diversity ranged from 2.26 to 3.09, with a slight but non-significant increase in diversity after
314 2003 due to the arrival and increased abundance of blacktips, Galapagos and tiger sharks over time (Figure
315 3a). The lowest diversity was observed in 1998, while the highest occurred in 2006. We did not find any
316 significant variations in monthly species diversity, with values ranging from 2.6 in March to 2.84 in
317 November (Supplementary figure 2).

318 ANOSIM results supported the network statistics, as we observed a strong and significant separation in
319 community composition between consecutive years ($R = 0.32$, $p = 0.001$) with lower mean dissimilarity
320 values from 2006 onwards. Dissimilarity percentages ranged between 25% and 55% (Figure 4b). On the
321 other hand, the monthly variation showed a weak but significant level of separation in community
322 composition ($R = 0.06$, $p = 0.001$). Peak dissimilarity was observed during April, descending towards its
323 lowest point in October before increasing again (Supplementary Figure 3).

324 Finally, we identified three species as the primary contributors to dissimilarity within the community
325 thanks to the SIMPER analysis: Scalloped hammerheads, Whitetip reef sharks, and Marbled rays (Figure
326 3c). Scalloped hammerheads were the dominant contributors to dissimilarity in most years (303 out of
327 351 pairs of years), with their contribution ranging from 9.9% to 64.32% of between year dissimilarity.
328 Whitetip reef sharks followed (48 pair of years), contributing between 7.6% and 64%, whereas Marbled
329 rays consistently contributed the least among the top three species, with a range of 0.13% to 13.34%.



330

331 *Figure 3. Temporal analysis of species diversity and community composition in the elasmobranch community of Cocos Island.*
 332 *Panel (a) shows the yearly Hill Diversity Index, with each point representing average diversity for a given year over the study*
 333 *period. Panel (b) displays the Bray-Curtis dissimilarity percentages between consecutive years. Panel (c) highlights the*
 334 *contribution of the top three species (Scalloped hammerheads “Hammerheads” light teal color, Whitetip reef sharks*
 335 *“Whitetips” blue color, and Marbled rays, dark blue color) to the overall dissimilarity, as determined by SIMPER analysis. X axis*
 336 *shows each paired year in ascending order, with the first vertical bar representing the dissimilarity between 1993 and 1994,*
 337 *followed by 1993 against 1995 and so on.*

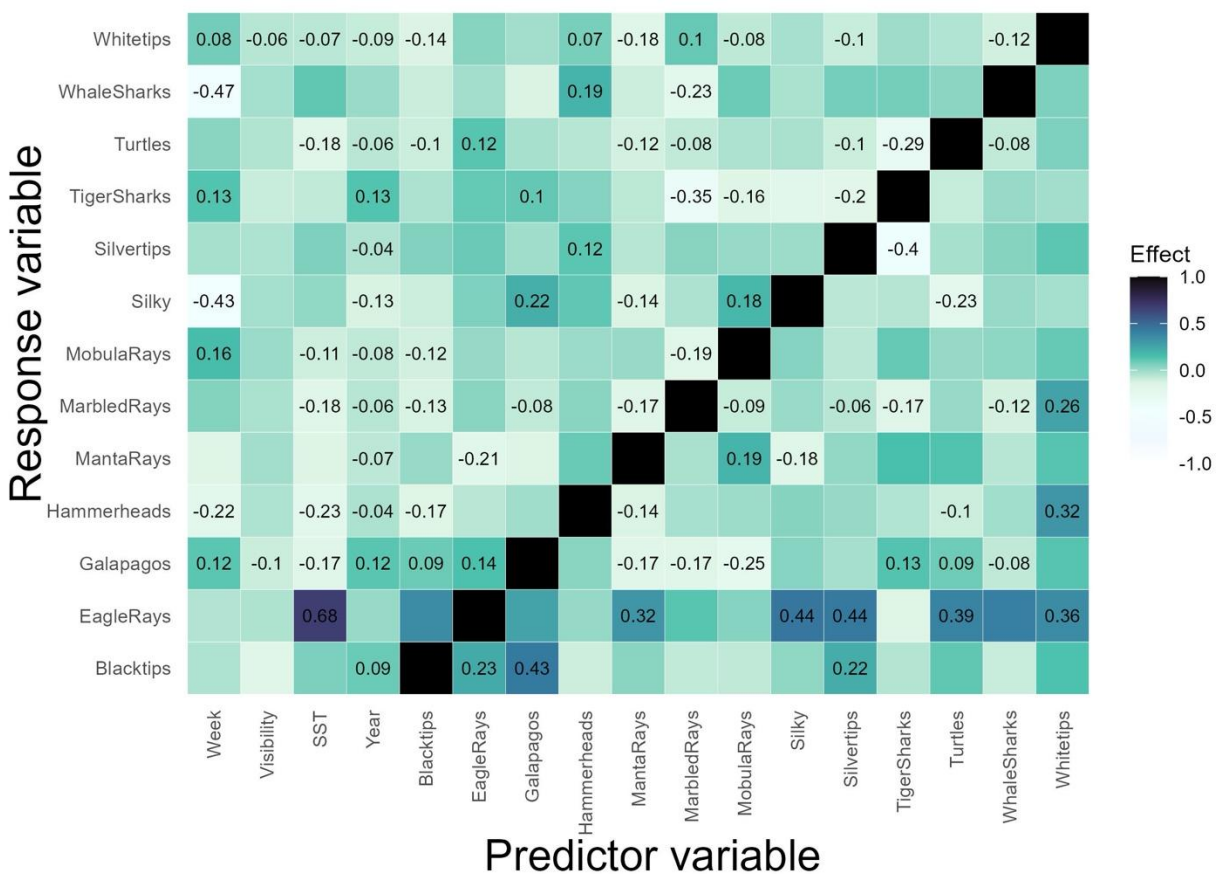
338 Interactive Abundance Models

339 The correlation analysis for our models shows no covariates with correlation values beyond the 0.7 or -
 340 0.7 collinearity cutoff with another covariate (Supplementary Figure 4). Excluding non-biological
 341 covariates, the highest correlation value observed between species counts was 0.48 (between Marbled
 342 rays and Whitetips), the lowest correlation value was of -0.17 (Between Galapagos and Whitetips).

343 All models converged with relatively accurate predictions for mean occurrences of the species studied,
 344 with the highest differences between observed and predicted abundances observed for scalloped
 345 hammerhead and whitetip reef sharks, the most abundant species (Supplementary Figure 5). The lower
 346 predictability observed for the more abundant species was expected due to the high number of individuals
 347 observed during certain periods of time. The errors, however, were relatively small, with average weekly
 348 differences ranging from -2.5 to 5 individuals.

349 Model outputs remained largely unchanged when excluding turtles as a covariate, demonstrating the
 350 robustness of our approach. However, time variability and sample size influenced the 2014–2019 models,
 351 where overall trends remained consistent, but some species interactions lost statistical significance.
 352 Notably, no significant species co-occurrence switched from positive to negative or vice versa. These
 353 results are provided in the supplementary material.

354



355

356 *Figure 4. Visualization of results from Bayesian models. Colors represent the mean posterior estimates of the Bayesian beta*
 357 *coefficients from each species interaction model. Response variables are shown in the y axis, predictor variables are shown in*
 358 *the x axis. Note that values are not mirrored due to the asymmetric nature of interactions. Darker colors for positive effects,*
 359 *lighter colors for negative effects. Cells with numbers highlight significant effects (95% credible interval for posterior distribution*
 360 *did not cross zero did not cross 0).*

361 Eight out of the 13 species studied showed a significant declining trend in abundance with only three
 362 species showing a positive trend (Table 1). Sea surface temperature had a significantly negative effect
 363 on the occurrence of six species, with Eagle rays being the only species positively influenced by an increase
 364 in temperature (Figure 4). Visibility had a weak but significant negative effect on Galapagos and Whitetips.
 365 Finally, seasonal variation, represented as the cyclic week of the year, had a significantly negative effect
 366 (decrease in occurrence during certain seasons) for scalloped hammerheads, Silky and Whale sharks, and
 367 a positive effect for Mobulas, Galapagos, Tiger and Whitetip reef sharks.

368 Regarding species interactions. Galapagos sharks, Marbled rays and Turtles were tied for the greatest
 369 number of significant interactions with eight each. In contrast, Mobula rays, Silvertip and Whale sharks
 370 were tied for the least significant interactions count with two each. As expected, most species interactions
 371 were negative (43 out of the 59). We observed the strongest positive effects between Eagle rays and Silky
 372 sharks (0.44 ± 0.21), Eagle rays and Silvertips (0.43 ± 0.22) and Blacktips and Galapagos (0.43 ± 0.11). The
 373 strongest negative effects were observed between Silvertips and Tiger sharks (-0.39 ± 0.09), Tiger sharks
 374 and Marbled rays (-0.35 ± 0.11), and Turtles and Tiger sharks (-0.29 ± 0.05).

375 *Table 1. Summary of significant species interactions. Coefficient shows the mean posterior beta coefficient for year as a predictor*
 376 *of species abundance, with NS showing non-significant effects. The abiotic predictors column show the number of significant*
 377 *effects for the species amongst our Week, Visibility, Sea Surface Temperature (SST) and Year covariates. Predictor species show*
 378 *the count of significant species covariate effects. Species interactions show the number of significant interactions attributed to*
 379 *each hypothesized species interaction. Only significant effects (95% credible interval for posterior distribution did not cross zero*
 380 *did not cross 0) are counted.*

Species	Population	Coefficient	Abiotic predictors	Predictor species	Species interactions		
					Competition	Environmental	Predation
Blacktips	Increase	0.09	1	3	0	3	0
Eagle rays	Stable	NS	1	5	0	3	2
Galapagos	Increase	0.12	4	8	0	5	3
Hammerheads	Decrease	-0.04	3	4	0	4	0
Manta rays	Decrease	-0.07	1	3	0	3	0
Marbled rays	Decrease	-0.06	2	8	0	4	4
Mobula rays	Decrease	-0.08	3	2	0	1	1
Silky	Decrease	-0.13	2	4	0	4	0
Silvertips	Decrease	-0.04	1	2	1	1	0
Tiger sharks	Increase	0.13	2	4	1	1	2
Turtles	Decrease	-0.06	2	7	0	4	3
Whale sharks	Stable	NS	1	2	0	2	0
Whitetips	Decrease	-0.09	4	7	1	5	1

381

382 Discussion

383 Here we use a novel combination of network analysis, diversity metrics, and ecological modeling to
 384 provide an evaluation of elasmobranch community dynamics in the Eastern Tropical Pacific. Our results
 385 corroborate past work on the continued decline of several key elasmobranch species (Figure 4). Several
 386 recent studies suggest that Cocos Island's elasmobranch community is undergoing a shift in species
 387 composition (Espinoza et al., 2020; White et al., 2015). Potential drivers of these changes include the

388 increasing frequency of sea surface temperature anomalies (Osgood et al., 2021; Saltzman & White, 2023),
389 the continued decline of the most abundant elasmobranch species and the arrival of new predators and
390 competitors to the system (Espinoza et al., 2024; Saltzman et al., 2024). We provide new insights into the
391 observed changes in community composition by using indirect methods to evaluate species interaction
392 and community processes.

393 Our diversity and network analysis suggest that changes in the Cocos Island elasmobranch community are
394 driven by complex ecological processes in addition to the previously reported declines in species
395 abundances. Notably, the primary contributors to yearly dissimilarity remain the most abundant yet
396 rapidly declining species: scalloped hammerhead sharks, whitetip reef sharks, and marbled rays. In
397 contrast, network strength, a key measure of community connectivity and resilience (Landi et al., 2018),
398 has experienced significant changes in just a few decades. Peaks in network strength during the dry season
399 of 2006 and the wet season of 2014 were most likely caused by the arrival of blacktips, Galapagos and
400 tiger sharks, their increase in abundance and higher co-occurrence with other species. Nevertheless, the
401 reduction in network strength after 2014 may suggest the community has reached a new stability point.
402 Furthermore, we observed an inverse relationship in network strength between seasons, with higher
403 network strength values during the wet period correlating with reduced network strength in the dry
404 season. Our observations suggest that the community may have split into two temporally distinct sub-
405 communities. The fact that diversity metrics did not capture this seasonal variation implies that, while
406 most species can be encountered year-round, network structure is shaped by resident species during the
407 dry season and by more seasonal species during the wet season. Next, we built on our work with diversity
408 and network metrics by exploring potential interspecific interactions based on our interactive model
409 outputs.

410

411 **Competition**

412

413 Competition encompasses both direct and indirect ecological interactions, including territoriality and
414 interference over limited resources (Twining et al., 2024). We interpreted the negative effect of one
415 species on another's abundance as evidence of potential competition, limited to species unlikely to have
416 predator-prey relationships but sharing ecological or trophic niches. For example, while large blacktip
417 sharks occasionally prey on neonate or juvenile scalloped hammerheads (Castro, 1996), such interactions
418 are unlikely at Cocos Island, where hammerhead sharks are primarily adults. Furthermore, previous
419 studies have reported competitive exclusion of blacktips by larger sharks such as great hammerheads
420 (*Sphyrna mokarran*), bull sharks (*Carcharhinus leucas*) and tiger sharks (Doan & Kajiura, 2020; Lear et al.,
421 2021) suggesting the negative interaction between blacktips and scalloped hammerheads is competitive
422 in nature. Our results also suggest competition is occurring between tiger sharks and silvertip sharks, likely
423 driven by spatial overlap or shared resource use (Espinoza et al., 2024). Even when other large sharks such
424 as silvertips are not regularly preyed upon by tiger sharks, exploitative competition for shared prey items
425 may be taking place in the island (White et al., 2015). Competition for prey may explain the negative
426 interaction between tiger sharks and silvertips, since both shark species were significant negative
427 predictors for turtle and marble ray abundance. Such competitive exclusion events have also been
428 reported in areas such as New Caledonia, where silvertip sharks ignored highly attractive food sources,
429 such as whale carcasses, in the presence of tiger sharks (Clua et al., 2013). Competitive exclusion plays an

430 important role in shaping the structure of predator communities (Condamine et al., 2019; Schoener, 1974)
431 and should be considered in managing and evaluating shifting communities such as Cocos Island's
432 elasmobranchs.

433

434 **Predation**

435

436 We determined the occurrence of potential predation (predator-prey interactions) when we observed
437 either an increase in predator abundance associated with the presence of potential prey or a decline in
438 prey species correlated with predator presence. Blacktip sharks frequently prey on mobulids in the
439 Eastern Tropical Pacific (Estupiñán-Montaña et al., 2018), which may explain why blacktips were negative
440 predictors for mobulas and marbled rays. These batoids are more pelagic species than marbled rays and
441 much smaller than manta rays (Saltzman & White, 2023). Galapagos and silvertip sharks are large
442 predators known to feed on rays at Cocos Island (Garrison, 2006) and could negatively affect marbled ray
443 abundance, suggesting either direct predation or predator avoidance. As expected, tiger shark presence
444 correlated with lower turtle abundance, aligning with studies that highlight turtles as a key component of
445 tiger shark diets (Heithaus, 2001; Saltzman et al., 2024) . Despite being opportunistic predators, tiger
446 sharks did not show significant predation effects for species other than turtles in our models. Instead,
447 their presence negatively affected other top predators, suggesting that predation-driven changes in the
448 Cocos Island community result not only from direct consumption but also from behavioral responses to
449 predator presence. Potential behavioral effects were further highlighted by the negative correlation
450 between silvertip and reef whitetip sharks. Although direct predation of whitetips by silvertips has not
451 been documented, previous studies at Cocos Island suggest larger predators may opportunistically prey
452 on smaller whitetips (White et al., 2015). The large number of potential predation interactions suggested
453 by our results reflect the complexity of trophic dynamics and behaviorally mediated interactions in marine
454 communities with multiple co-existing predators.

455

456 **Environmental preferences**

457

458 Finally, we inferred that some significant species co-occurrences may have been caused by environmental
459 conditions based on previous studies (Osgood et al., 2021; Saltzman & White, 2023). Highly mobile
460 predators such as blacktip, Galapagos, tiger, and silky sharks showed positive correlations, likely reflecting
461 shared habitat preferences (Estupiñán-Montaña et al., 2018; Hoffmayer & Parsons, 2003; Papastamatiou
462 et al., 2006). Previous studies identified the arrival of blacktip, Galapagos and tiger sharks as a potential
463 driver of community shifts at Cocos Island (Espinoza et al., 2024; White et al., 2015), likely influenced by
464 fishing (Burns et al., 2023; Worm et al., 2024) and climate change (Osgood et al., 2021). Batoid species
465 also exhibited strong associations with environmental variables. Research on filter feeders at Cocos Island
466 previously identified oceanographic conditions as key drivers of species composition (Saltzman & White,
467 2023). Model outputs reinforced this finding, as multiple batoid species co-occurred with specific
468 environmental conditions rather than with competing sharks or predators. Eagle rays showed a positive
469 correlation with turtles but a negative correlation with manta rays, possibly indicating habitat partitioning
470 or differences in environmental tolerance. Warmer sea surface temperatures also influence eagle ray
471 distributions, which may explain the observed patterns (Rastoin-Laplane et al., 2023). Marbled rays were

472 positively associated with whitetip reef sharks but negatively associated with turtles, mobula rays, and
473 whale sharks, suggesting a preference for coastal habitats over oceanic environments (Garrison, 2006).
474 Similarly, mobula rays were negatively correlated with coastal predators such as Galapagos, tiger,
475 marbled, and whitetip reef sharks, and positively correlated with oceanic species like silky sharks and
476 manta rays, reinforcing shared habitat preferences and trophic niches (Saltzman & White, 2023). Our
477 findings suggest that environmental preferences play a crucial role in structuring the elasmobranch
478 community at Cocos Island, with species aggregating in response to environmental conditions rather than
479 direct biotic interactions.

480

481 **Study limitations and implications for management**

482 Our analysis evaluated interspecific interactions within the elasmobranch community at Cocos Island but
483 omitted other species in the ecosystem. Future research could implement joint species abundances as
484 predictors, as well as other important functional groups in Cocos Island such as fish and other prey items
485 to determine if more complex indirect relationships are also playing a role in shaping the community. The
486 individual models we used provided interpretable results relevant to our research questions, yet more
487 complex methodological frameworks, such as joint species distribution models or matrix-based time
488 series models (Ovaskainen et al., 2017), could better account for indirect effects and trophic web
489 interactions. However, implementing more specialized analysis using currently available data would
490 require either a more systematic sampling approach or additional methodological adjustments tailored,
491 both of which fall beyond the scope of this study. We addressed the limitations associated with sampling
492 effort in our study by implementing offset terms in our models and averaging species counts, methods
493 often used to work around sampling limitations for modeling (Kéry & Schaub, 2012). Our sensitivity
494 analysis further highlighted structural considerations within our models. For example, only 19 out of 59
495 biological predictors across our interactive models remained significant when re-running our models using
496 only the last 5 years of data (Supplementary figure 8). The small subset of significant effects likely reflects
497 the challenge of detecting subtle species interactions within a limited dataset, a problem overcome in the
498 complete model by the uniquely long series of data collected from Cocos Island. These community shifts
499 may be indicators of broader ecological restructuring rather than methodological issues, and are a
500 phenomenon commonly observed in dynamic ecological networks (Landi et al., 2018). Future work could
501 explore differences in community metrics and network dynamics between historical and recent datasets
502 to further assess how the Cocos Island community is changing. In this study, we found that species co-
503 occurrence data and indirect analytical methods have the potential to provide meaningful insights about
504 inter species interactions. We hope our results work as a steppingstone towards integrating multi-species
505 frameworks in ecological modeling, providing a more comprehensive understanding of community
506 dynamics for the effective management and conservation of marine ecosystems.

507

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512

513 **Data Availability**

514 Data used for this study belongs to the Undersea Hunter Group (underseahunter.com). Data can be made
515 available upon request. Annotated code can be reviewed and branched from our GitHub repository at
516 <https://github.com/Miguelbirostris/CocosElasmoNetwork2025>.

517

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