

1 **Supplementary Information**

2 **Supplementary Methods**

3 We collected fine-scale network data concerning fishing-related information-sharing for the gillnet
4 skipper community in San Jose, Lambayeque, Peru ($6^{\circ}46' S$ $79^{\circ}58' W$). Sea turtles captured in gillnets
5 are the marine group of primary conservation concern in San Jose; thus, information-sharing about sea
6 turtle bycatch is one of the study's primary interests. Nevertheless, the sea turtle bycatch reduction
7 initiative (LEDs on gillnets) may potentially add value to catch (fishing finance), as well as relating to
8 other fishing-related information shared, such as fishery regulations, weather conditions, crew
9 management, and vessel technology and maintenance. Thus, nine individual information-sharing
10 networks were collected. This study aimed to investigate the social structure amongst skippers;
11 therefore, the study's data pertain to respondent-to-respondent networks only (which also allowed for
12 consistent respondent numbers between cross-network comparisons).

13

14 The gillnet is the most common fishing gear used in Peru's small-scale fishing fleet¹. Several marine
15 megafauna taxonomic groups are incidentally captured in gillnets in San Jose²⁻⁵, of which sea turtles
16 have been highlighted as a major conservation issue that warrants further management^{2,6}. Across the
17 inshore-midwater vessel class in San Jose, sea turtle captures per trip has been calculated at 0.71 for
18 green turtles (*Chelonia mydas*), 0.08 for olive ridley turtles (*Lepidochelys olivacea*), and 0.02 for
19 leatherback turtles (*Dermochelys coriacea*). Turtles released alive without visible injury comprise
20 nearly 62% of 461 fishing trips observed from San Jose between August 2007 to March 2019. Live
21 releases with injuries 28% of captures and 8% mortalities⁶. Gillnetting across two distinct fleets in
22 San Jose has been defined as posing an extreme and major risk to population recovery goals of the
23 East Pacific Regional Management Unit (RMU)⁷ populations of these species⁶.

24

25 Because of the threat that gillnets pose to sea turtles in Peru's northern fishing ports, our chosen study
26 population was actively fishing San Jose skippers deploying gillnet gear year-round, including those
27 who owned and operated their vessels and those who skippered for others. Skippers were chosen as

28 they are in charge of the fishing gear and crew when the boat is in the water and the gears deployed,
29 and therefore their decisions are most influential in opportunities to reduce turtle bycatch (for
30 example, through better live release, or the use of LED lights on nets to reduce incidental captures⁸).

31

32 Five gillnet skippers and their crew are currently involved in a trial community co-management
33 bycatch reduction scheme operating from San Jose that requires fishers to use light-emitting diodes
34 (LEDs) on their nets a technology shown to reduce green turtle bycatch by 64% while maintaining
35 levels of target catch in randomized control gear trials in Sechura Bay (Main text Figure 2a), located
36 approximately 150 kilometers north of San Jose⁸. Acoustic alarms ('pingers') are also fitted to nets to
37 reduce small cetacean bycatch⁴, and a remote electronic monitoring device is under trial to improve
38 data paucity⁹.

39

40 *Determining population size*

41 During months with warmer weather (and hence better fishing conditions), the number of skippers
42 can more than double as fishers arrive from inland areas seeking fishing work. Skippers typically
43 operate with 1–4 crew¹⁰. Peruvian law defines small-scale fishing vessels as displacing a maximum of
44 32.6m³ Gross Registered Tonnage (GRT), up to 15m in length, and operated predominantly manually.
45 San Jose's small-scale gillnet vessels can be subdivided into two fleets. The first fleet comprises a
46 class of open-welled boats known as 'chalanas', with a capacity range from 1–8 t. The second fleet
47 comprises a predominantly larger vessel class known as 'lanchas', with small closed bridges ranging
48 in capacity from 5–32 t¹. The survey interviewed actively fishing gillnet skippers on both chalana and
49 lancha vessels. Previous estimates of gillnet activity in San Jose recorded 95 gillnet vessels fishing in
50 January–April 2004¹⁰, and 47 gillnet vessels fishing in November 1995–April 1996¹¹.

51

52 The total population (n=168) was determined by triangulating data obtained from membership lists of
53 the two main fishing groups in San Jose, lists of vessels daily launching and landing logs, and key
54 informant interviews. We restricted the network analysis to gillnet skippers – who owned their own

55 vessel(s) or who skippered a vessel owned by someone else, and who launched and landed their
56 vessels from the beach at San Jose, Lambayeque, Peru (6°46' S, 79°58' W). Gillnet skippers were
57 required to be identified as actively fishing at least once during the winter period of 1 July – 30
58 September 2017 using one or more of the data sources used for the analysis.

59

60 There are two main at-sea fishing groups in San Jose (the Maritime Union of Fishermen Society, and
61 the Artisanal Fishermen and Hydrobiological Extractors Association). Following initial introductions
62 being made with both of the fishing groups leaders during which time we presented a description of
63 the study and associated ethical clearance, we were granted access to the fishing groups membership
64 lists, which contained information on gillnet skipper name, vessel name, and vessel unique
65 identification (plate number). During our survey period, the fishers in San Jose were pushing and
66 pulling their fishing vessels in and out of the water from the beach using large tractors that were
67 driven by employees of a local company that specialized in providing this service. Subsequent
68 information from San Jose in early 2019 indicates that this service is no longer being provided due to
69 legal implications imposed by recently implemented Government legislation. Skippers were charged a
70 fee and the tractor drivers record each vessel (using the plate number) as they are pushed each vessel
71 out to sea and pulled each vessel back onto the beach following a fishing trip. The daily launching and
72 landing logs were provided following a meeting with the company owner and the tractor drivers,
73 during which we presented a description of the study and associated ethical clearance. The daily
74 launching and landing logs were cross referenced with the list of active fishing group members and
75 the list of actively fishing gillnet skippers was checked by several key informants during two key
76 informant interviews held in San Jose in July 2017. Between 1 July – 30 September 2017 every
77 actively fishing gillnet skipper (n=168) was identified and asked if they would like to partake in the
78 interview; only three actively fishing gillnet skippers declined.

79

80 *Social Network Analysis structured questionnaire*

81 We surveyed with a fixed choice survey design, where respondents were asked to consider up to ten
82 individuals with whom they exchange useful information about fishing and whom they considered

83 valuable to their fishing success (see full questionnaire below). The decision to limit the number of
84 skippers each respondent could specify was made for practical survey purposes as the network we
85 surveyed was relatively large. The fixed-choice survey design also had the secondary benefit to help
86 respondents understand what is required of them during the survey, as a free-choice survey design can
87 result in subjective interpretations of the desired links¹². While the number of out-going links was
88 limited to ten, there was no limit on the in-degree of links in the network (i.e., there was no limit to
89 the number of times others could nominate a skipper), which was the main focus of our analysis.

90

91 Respondents were asked to consider people from San Jose that they share useful information about
92 fishing with; considering those that they thought may influence their fishing success. Respondents
93 were reminded that the shared information and names will remain anonymous and will not be
94 revealed. We highlighted that the information provided will help us understand how information that
95 relates to fishing flows between fishers. Prior to the fixed response, respondents were asked to
96 consider relationships that they have had with other vessel owners, captains, owner/captains (owners
97 who also captain their vessel), other fishery leaders, fishery management officials, members of the
98 scientific or not-for-profit community, boat launching / landing support, fish transport associations,
99 fish sellers/market operators, their family and friends, and any other people they have fished with, or
100 shared information with about fishing over the last 5 years.

101

102 We classified two broad categories about which we expect gillnet skippers to exchange fishing related
103 information. These include 1) the process of fishing, and 2) the business and governance of fishing.
104 We then disaggregated these two broad categories into nine fine-scale information-sharing types that
105 relate to fishing, including i) turtle bycatch, ii) gillnet type and maintenance, iii) weather conditions,
106 iv) fish location and catch sites, v) fishing activity (how many people are fishing, who is fishing, who
107 caught what), vi) vessel technology and maintenance, vii) fishing regulations (laws and rules), viii)
108 fishing finances (market prices, loans, fines, penalties), and ix) crew management. Fishing-related
109 information categories were randomized before interviewing each respondent using a random number
110 generator

111

112 *Assessing cross-network correlations*

113 The basic properties of each information-sharing network, and the nomination structure in general,
114 will have a larger deterministic influence on the cross-network correlations. For instance, considering
115 a network of ‘any nomination in any information-sharing network’, we would expect each network to
116 hold a correlation equal to that of the number of nominations in each network (Supplementary Fig. 7).
117 Similarly, networks with similar numbers of nominations are more likely to be more correlated with
118 one another than those with very different numbers of nominations. Simply carrying out edge-
119 permutations, even conservative ones controlling for the number of nominations, or degree
120 distributions, for example, would, by definition, randomize the underlying dyadic structure (who can
121 nominate who) and thus means all observed cross-network correlations would differ largely from
122 expected under this null model just due to this alone. To infer the extent to which networks are more,
123 or less, similar than expected under the general dyadic social structure, we carried out a cross-network
124 null model: For each dyadic nomination across any of the networks, we randomized the networks that
125 these nominations were made within. For instance, when individual X nominated individual Y for
126 information sharing within three different networks, we allowed these three nominations to be
127 reassigned to any of the networks, but all three still in the direction of individual X nominating
128 individual Y within these networks. In this way, the overall dyadic nomination structure was
129 maintained, but the networks within which these dyadic nominations took place within were
130 randomized. Using this method (termed ‘cross-network null model 1’ – Main text Figure 1c), 1000
131 permuted networks were generated, and the distribution of the expected cross-network correlations
132 was recalculated using this.

133

134 As an even more conservative version of a cross-network null model, we created a new version of
135 these permutations and controlled for the number of nominations that took place overall within each
136 network. For instance, when individual X nominated individual Y for information sharing within
137 three different information-sharing networks, these three nominations were reassigned amongst the
138 networks in a way that was equal to the number of nominations in each network. For example, if

139 network A had twice as many nominations in total as network B, reassigning a nomination between
140 individual X and individual Y would be twice as likely to be reassigned within the network A than the
141 network B. This permutation was done by merely swapping individual network nominations between
142 dyadic nomination pairs. This permutation is similar to a group-by-individual permutation¹³ but where
143 the rows of the matrix were set as the individual-to-individual dyadic nominations, and the columns
144 were set as each of the information-sharing networks. Using this permutation procedure (termed
145 cross-network null model 2 – Main text Figure 1d), we generated 1000 permuted networks (with 100
146 swaps between each network and a burn-in of 2000 swaps; Supplementary Fig. 8) and then calculated
147 the distribution of the expected cross-network correlations under this null expectation.

148

149 **Supplementary Results**

150 *Network summary statistics*

151 While the current study's focus is not on the broader network of non-skipper outgoing links, our analysis
152 showed the number of information-sharing links remained consistent between the respondent-to-
153 respondent network and the broader network that includes non-skipper nominees. Our analysis showed
154 that across nine different information-sharing networks evaluated, turtle bycatch remained the least
155 discussed type of fishing information in the wider network (in 64.2% of possible nominations).
156 Information about the weather and fishing activity were discussed the most (with 95.7% and 95% of
157 possible links, respectively). Turtle bycatch and fishing regulations were the only two information-
158 sharing networks that had a relative increase (both by 3%) in the amount they were discussed in the
159 wider network, compared to the respondent-to-respondent network that contained only skippers
160 (Supplementary Table 1b).

161

162 *Structural differences between information-sharing networks*

163 Degree assortativity

164 Our analysis of network degree assortativity (presented in the main text and akin to degree
165 homophily) found that networks of sea turtle bycatch information sharing nominations show no

166 significant assortativity in comparison to both the edge permutation null models (Main text Figure
167 2c). Individual gillnet skippers had a propensity to be disproportionately connected to other gillnet
168 skippers who had nominated a similar number of people as they had (out-degree assortativity).
169 Although none of the information-sharing networks were significantly different from the edge null
170 models in their out-degree assortativity, the sharing of information regarding sea turtle bycatch was
171 the only network that was slightly lower than expected, whilst all other networks were higher than
172 expected (Supplementary Fig. 2 and Supplementary Table 2). The lack of significant differences here
173 is probably due to the relatively low variance in out-going links in comparison to in-going links (i.e.,
174 due to the questionnaire set-up the number of nominations an individual could make was limited – see
175 Main text, Methods), and is most likely driven by a carry-over of the strong patterns evident in the in-
176 going nomination assortativity.

177

178 Our analysis shows that the lack of degree assortativity in the turtle bycatch context is most likely a
179 result of more complex dyadic-level behavior patterns driving each individual's attitudes and
180 behaviors. This is because the degree assortativity statistic itself is the level of like-to-like
181 connectivity given the total number of links. The edge permutations (edge null model 2) also (a)
182 directly control for the number of out-going and in-going links in each information-sharing network
183 (Main text Figure 2d), and (b) still find that degree assortativity is not significantly different in the sea
184 turtle network, but significantly differently in the other fishing-related information-sharing networks.
185 These comparisons are over and above that which would be expected from the differences in the
186 number of links, or even the degree distributions, specific to each network assessed.

187

188 Individual Centrality

189 As we aimed to examine the use of social network analysis for conservation-relevant systems, we did
190 not want to use simple node-level metrics that can be inferred without building social networks (e.g.,
191 using 'degree' is simply equivalent to counting the number of nominations an individual receives and
192 requires no knowledge of the network structure).

193

194 When considering the variance in betweenness (as an alternative measure of centrality;
195 Supplementary Fig. 3), or the mean eccentricity of each network's nodes (rather than the variance;
196 Supplementary Fig. 4), we found that the observed statistics from all networks (including sea turtle
197 bycatch) were lower (and mostly strongly significantly lower) than the statistics generated from edge
198 null model 1. This is most likely due to the random reassignment of in-going links in this permutation
199 causing (i) the assignment of in-going links to nodes which are originally disconnected in this context
200 and thus increasing the mean and (ii) the randomization of the in-going degree distribution increasing
201 the betweenness variance.

202

203 Seven of the nine information-sharing networks fell within the expected range of both the edge model
204 permutations for node eccentricity (how far an actor is from the furthest other), with bycatch and
205 fishing activity the only exceptions. We found that the observed variance in node eccentricity
206 (Supplementary Fig. 5) was lower than expected for information sharing regarding sea turtle bycatch,
207 in comparison to the null distributions (generated from the context permutations), which had higher
208 than expected observed variance in node eccentricity. The opposite was true for fishing activity. The
209 observed mean node eccentricity (Supplementary Fig. 6) followed a similar pattern to the variance in
210 node eccentricity, with information sharing regarding sea turtle bycatch being the only network that
211 was lower than expected in comparison to the null distributions. Mean node eccentricity for
212 information sharing regarding fishing activity illustrated the greatest contrast to the sea turtle bycatch
213 network with higher than expected observed statistics. This supplementary analysis demonstrates that
214 the sea turtle bycatch information-sharing network holds some structural dissimilarities in mean node
215 eccentricity, not only when being compared to the edge null models (Main Text Figure 3), but also
216 given the underlying social structure of who is connected to who within the network.

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218 *Cross-network correlations of dyadic links*

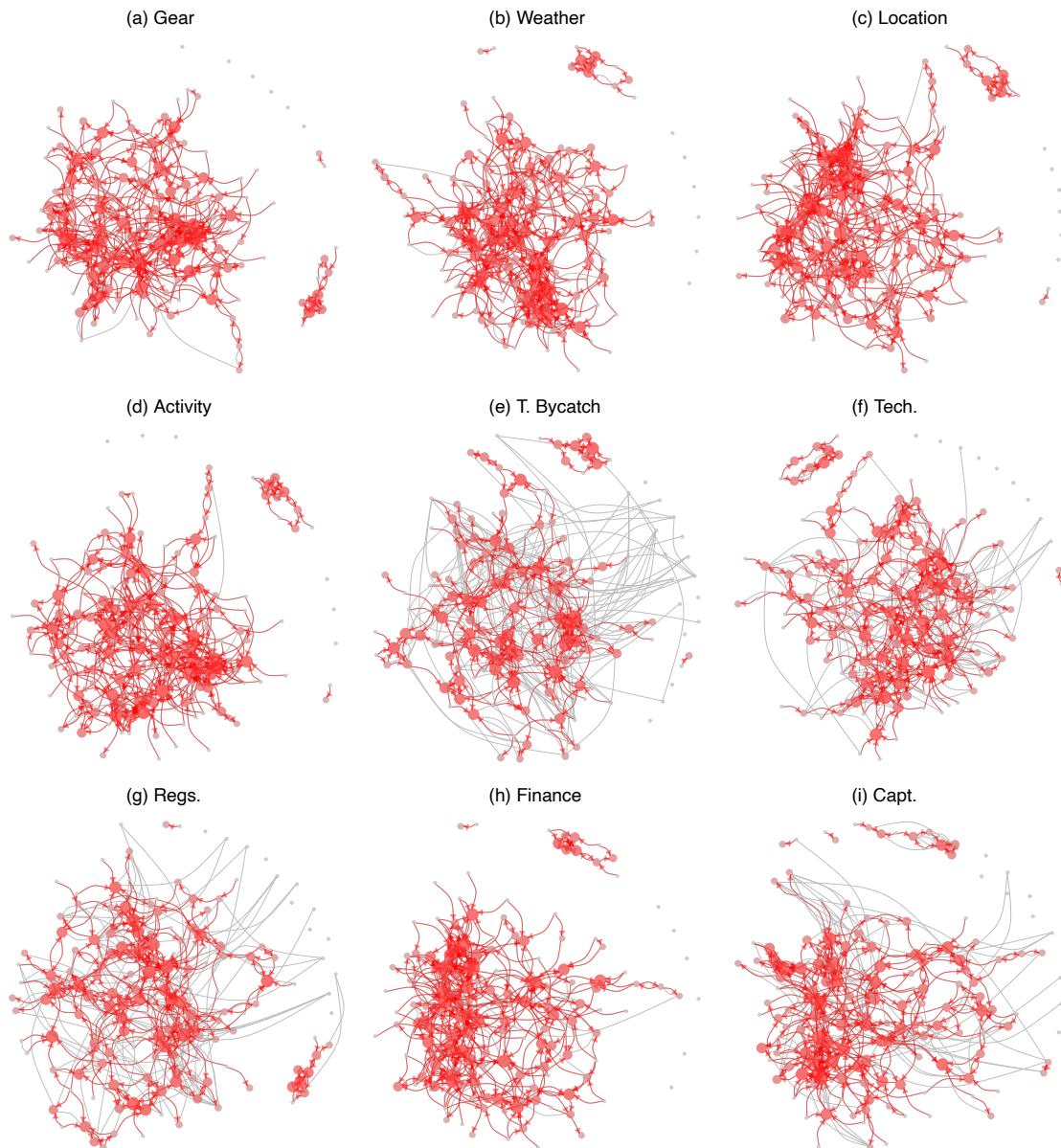
219 Along with focussing on the ability of each fishing-related information-sharing network to predict to
220 sea turtle bycatch information-sharing links, we also considered the correlation between all networks
221 and how these differed from the correlations expected under the cross-network permutation null

222 models (Supplementary Fig. 9). We found that the dyadic directed links within the ‘technology’
223 information-sharing network was more correlated with all the other networks than expected under the
224 general social structure of the network. This suggests that the technology information-sharing network
225 was particularly predictive of fishing activity in general.

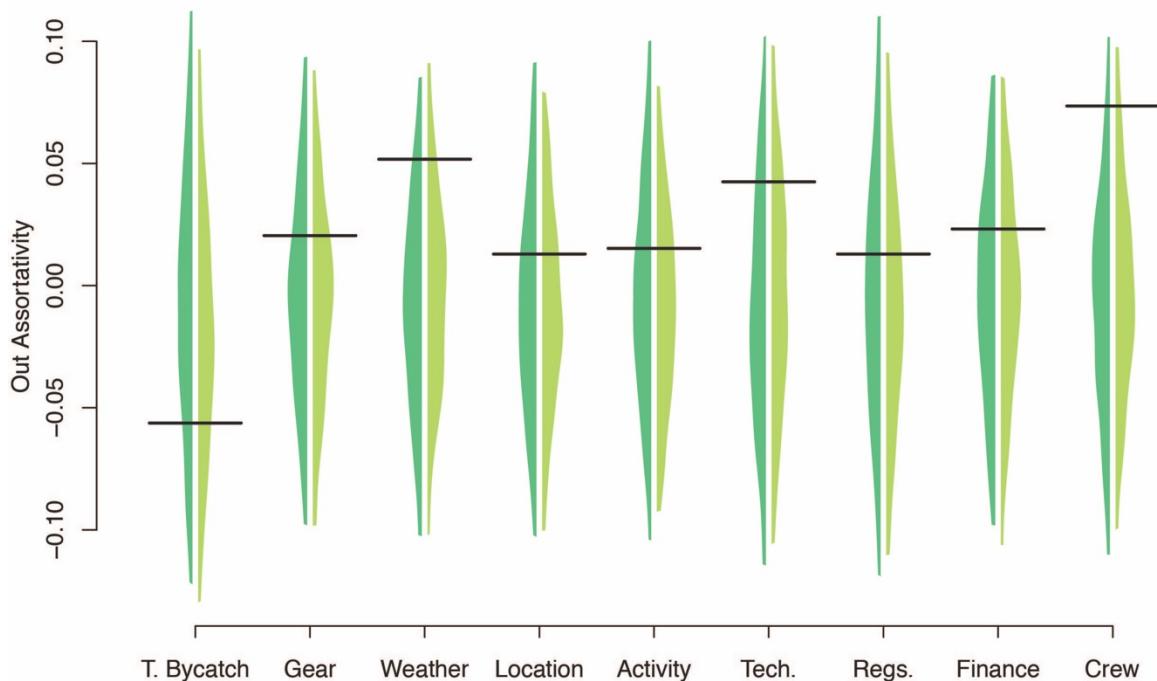
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227 As expected, when comparing the correlations to those generated from edge-permutations (rather than
228 cross-network permutations), the observed statistics were vastly different even though these
229 permutations were controlling for the number of nominations, degree distributions etc. due to
230 randomizing the underlying dyadic social structure (in terms of who can nominate who)
231 (Supplementary Fig. 10).

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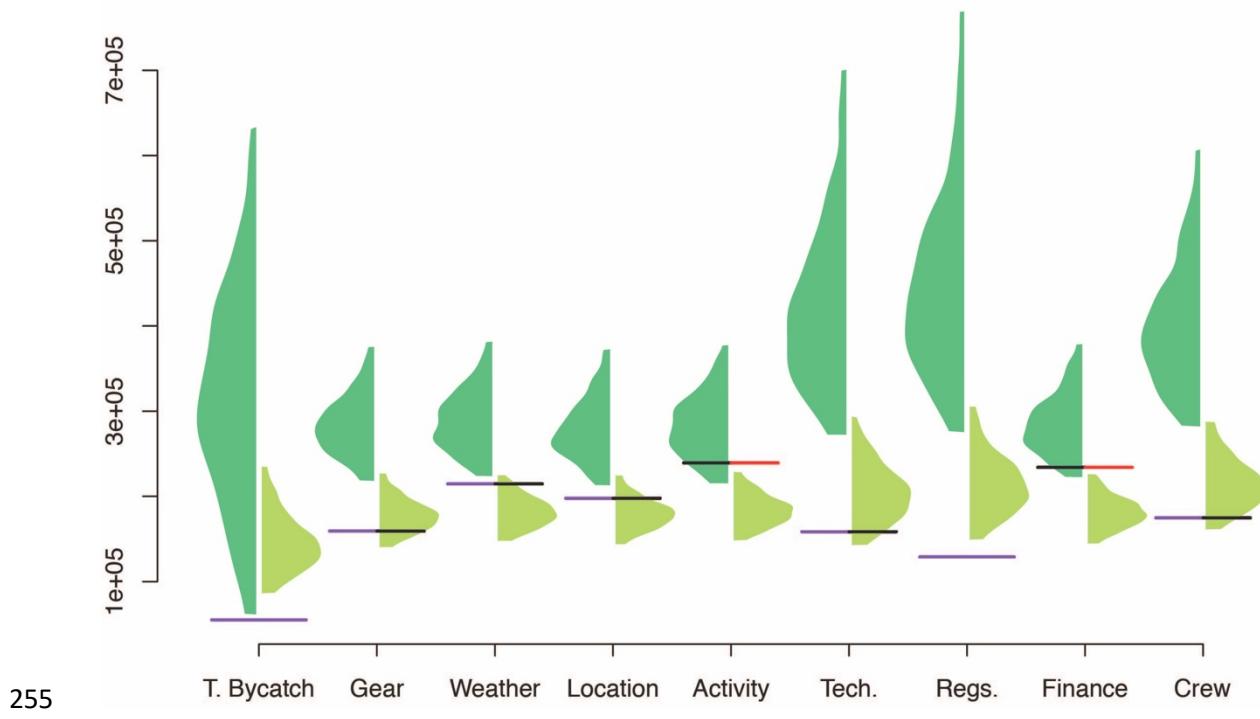
233
234 **Supplementary Figure 1. Illustrative network of the structure of information sharing across**
235 **fishng-related information-sharing networks.** The nodes show each of the skippers and the
236 adjoining lines show which dyads shared information in at least one network, and nominations within
237 the focal network (as indicated by heading) is highlighted as a directed red arrow here (arrow points
238 to the one that was nominated). Node size and shading shows the number of nominations each
239 individual received for the focal network (largest and most red = most nominations, small and grey =
240 no nominations). Layout was set as a spring layout of edges within each focal network (to minimize
241 overlap).



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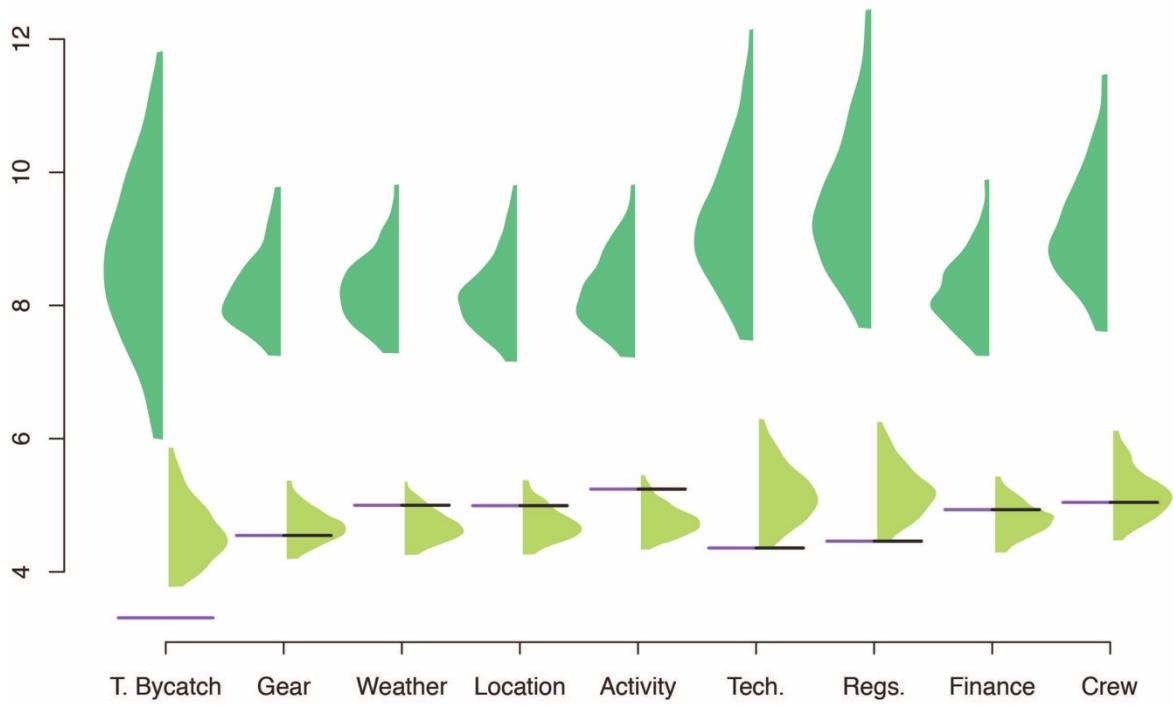
243

244 **Supplementary Figure 2. The observed assortativity coefficient for outgoing links in comparison**
 245 **to the null distributions for the different information-sharing networks.** Horizontal lines show the
 246 observed values from the actual networks (red = observed values are above the permutations, black =
 247 observed values are within the range of the permutations, purple = observed values are below the
 248 permutations). Polygon distributions show those generated by permutations (dark green = outgoing
 249 edge permutation that maintains the no. of nominations each individual makes, light green = edge
 250 swap that maintains the no. of nominations each individual makes and also the number of times each
 251 individual was nominated). Outgoing links also show the same pattern seen in figure 1 (i.e., the turtle
 252 bycatch network is the only information network measured which is not positively homophilous) but
 253 with no significant difference. For details on fishing-related information-sharing networks refer to
 254 Table 1 in the main text.



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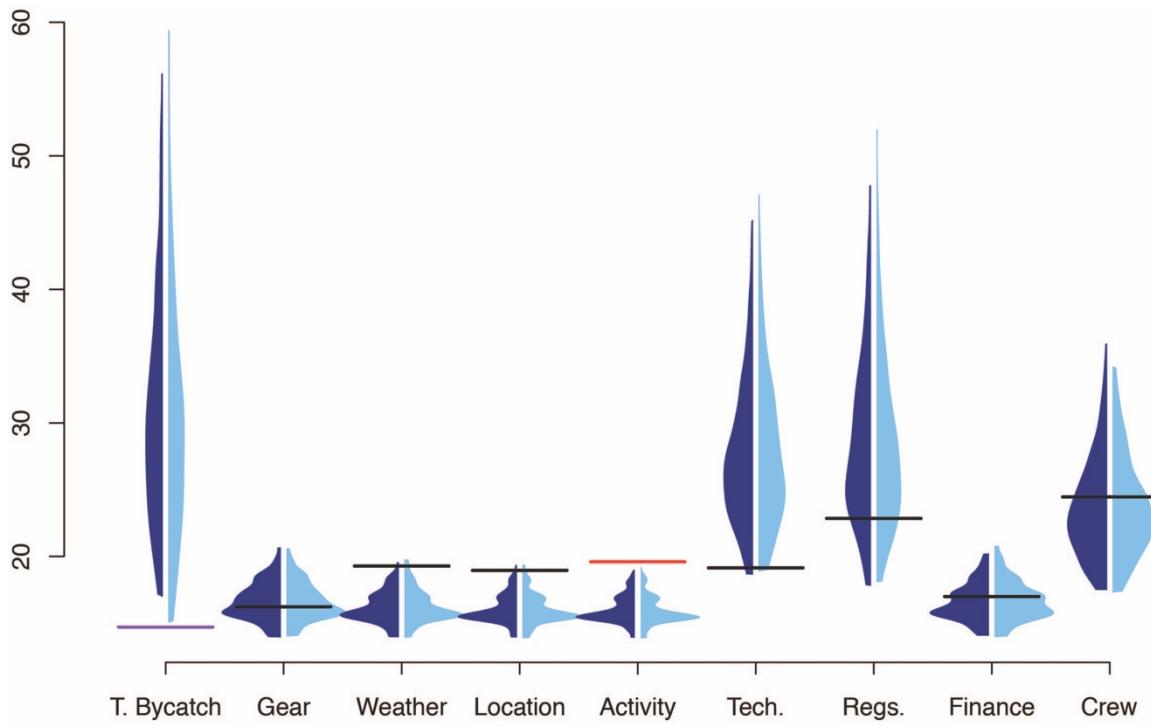
257 **Supplementary Figure 3. The observed variance in node betweenness in comparison to the null**
 258 **distributions for the different information-sharing networks.** Horizontal lines show the observed
 259 values from the actual networks (red = observed values are above the permutations, black = observed
 260 values are within the range of the permutations, purple = observed values are below the
 261 permutations). Polygon distributions show those generated by permutations (dark green = outgoing
 262 edge permutation that maintains the no. of nominations each individual makes, light green = edge
 263 swap that maintains the no. of nominations each individual makes and also the number of times each
 264 individual was nominated). Here a similar pattern to the degree assortativity (homophily) coefficient
 265 is also seen. For details on fishing-related information-sharing networks refer to Table 1 in the main
 266 text.



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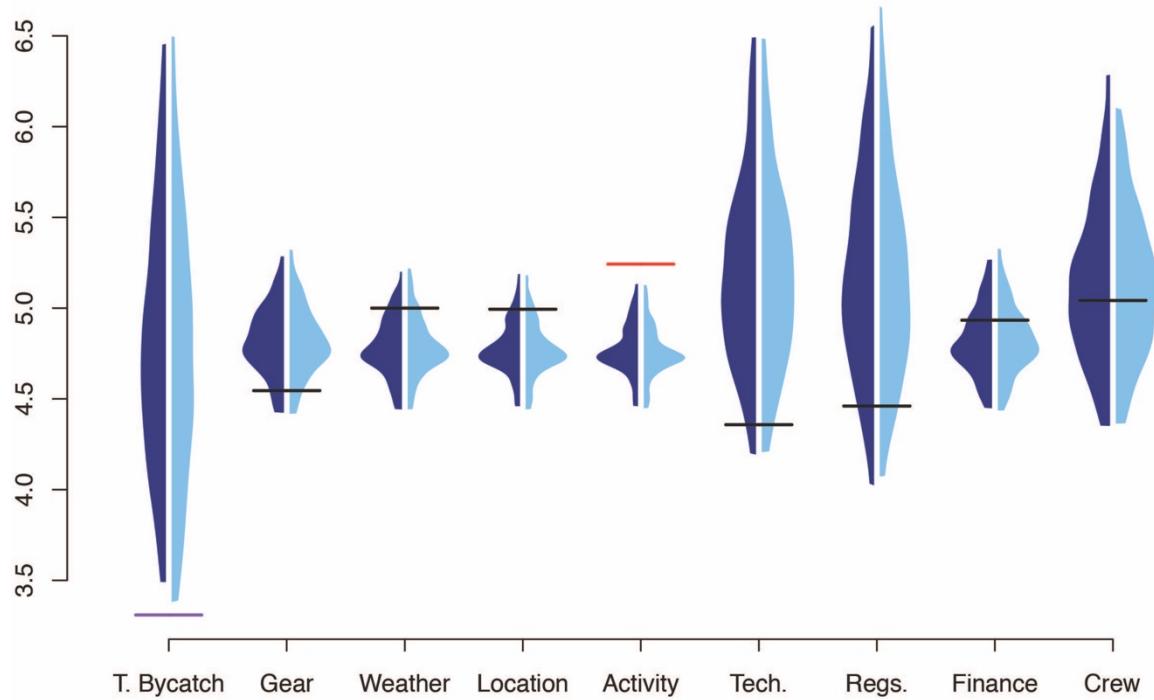
269 **Supplementary Figure 4. The observed mean node eccentricity in comparison to the null**
 270 **distributions for the different information-sharing networks.** Horizontal lines show the observed
 271 values from the actual networks (red = observed values are above the permutations, black = observed
 272 values are within the range of the permutations, purple = observed values are below the
 273 permutations). Polygon distributions show those generated by permutations (dark green = outgoing
 274 edge permutation that maintains the no. of nominations each individual makes, light green = edge
 275 swap that maintains the no. of nominations each individual makes and also the number of times each
 276 individual was nominated). Here a similar pattern to the degree assortativity (homophily) coefficient
 277 is also seen. For details on fishing-related information-sharing networks refer to Table 1 in the main
 278 text.



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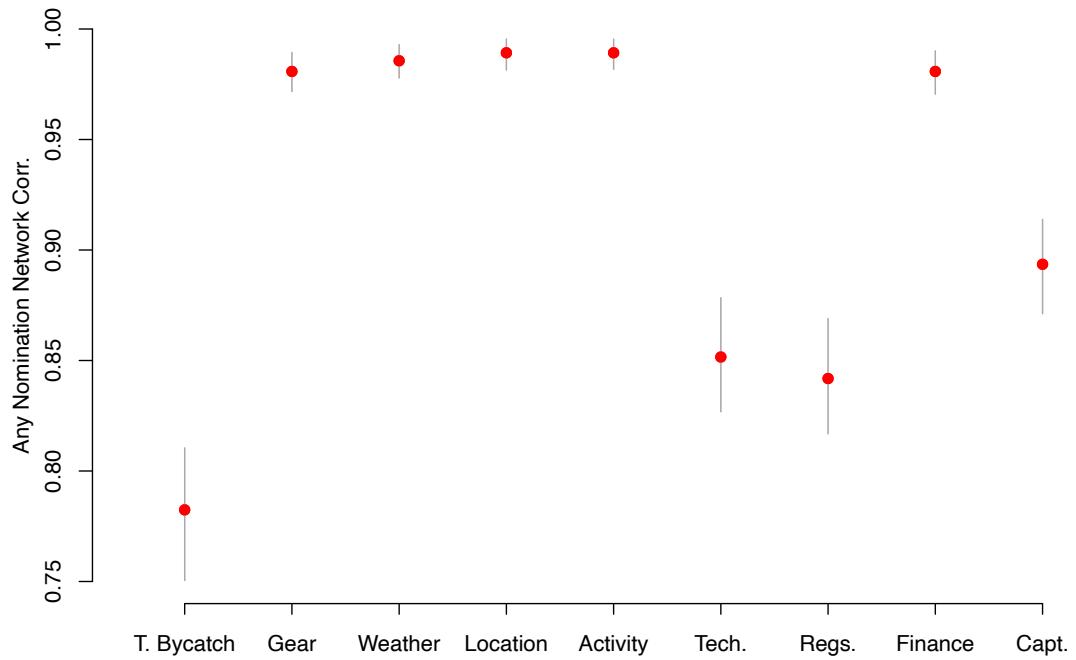
281 **Supplementary Figure 5. The observed variance in node eccentricity in comparison to the null**
 282 **distributions (generated from the cross-network permutations) for the different information-**
 283 **sharing networks.** Horizontal lines show the observed values from the actual networks (red =
 284 observed values are above the permutations, black = observed values are within the range of the
 285 permutations, purple = observed values are below the permutations). Polygon distributions show
 286 those generated by permutations (dark blue = cross-network swap that maintains the no. of
 287 nominations each individual makes and also the number of times each individual was nominated, but
 288 swaps the network these were made within whilst maintain the number of times each network was
 289 nominated as overall, light blue = conservative cross-network swap that is the same as dark blue, but
 290 also maintains the number of networks each dyad nominated each other for – but changes those
 291 networks (same as a gbi permutation but on the dyad-by-network edges). For details on fishing-
 292 related information-sharing networks refer to Table 1 in the main text.



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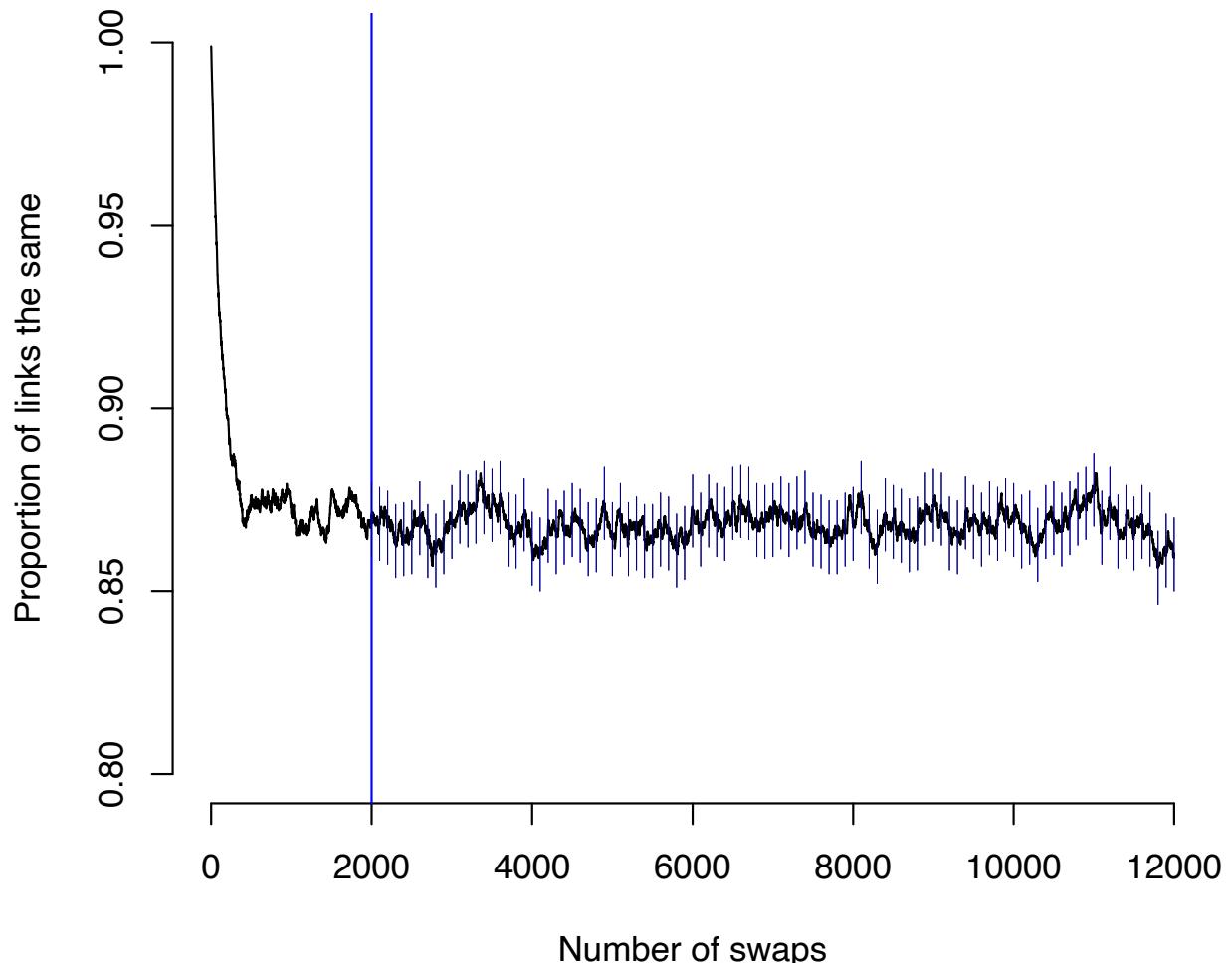
295 **Supplementary Figure 6. The observed mean node eccentricity in comparison to the null**
 296 **distributions (generated from the cross-network permutations) for the different information-**
 297 **sharing networks.** Horizontal lines show the observed values from the actual networks (red =
 298 observed values are above the permutations, black = observed values are within the range of the
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 300 those generated by permutations (dark blue = cross-network swap that maintains the no. of
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 304 also maintains the number of networks each dyad nominated each other for – but changes those
 305 networks (same as a gbi permutation but on the dyad-by-context edges). For details on fishing-related
 306 information-sharing networks refer to Table 1 in the main text.



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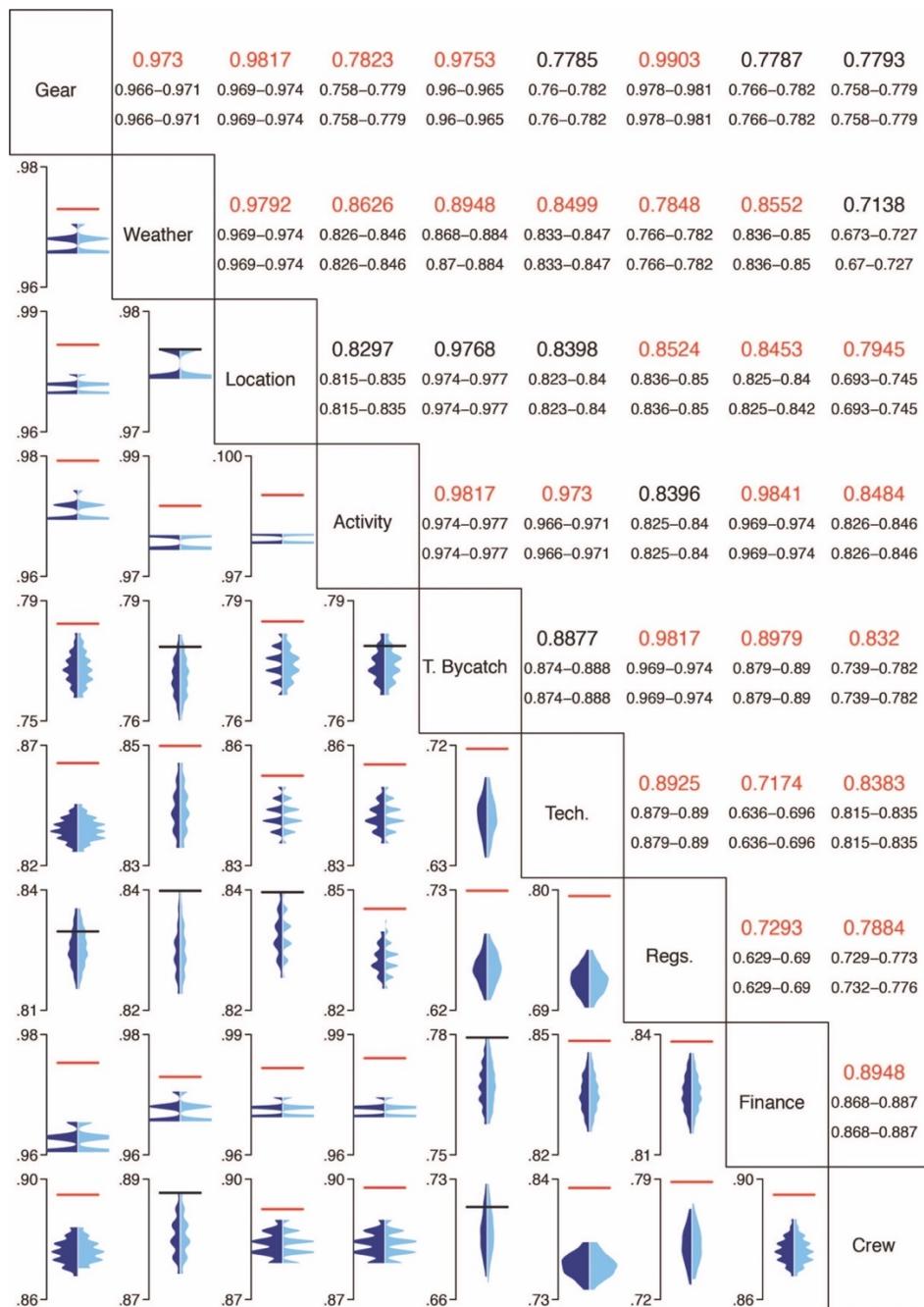
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309 **Supplementary Figure 7. Network differences to the ‘any’ nomination network.** Differences seen
 310 between different networks in how predictive/correlated they are to the ‘any’ nomination network
 311 (lines show bootstrap). For details on information-sharing networks see main text Methods –
 312 Experimental Design –Table 1.



313

314 **Supplementary Figure 8. Output for evaluation of cross-network null model 2.** This permutation
 315 procedure required sequential swaps of the networks in which nominations occurred between dyads
 316 (see Supplementary methods - Assessing cross-network correlations) to generate the null networks.
 317 The y-axis illustrates the number of nominations between individual-to-individual dyads that are in
 318 the same network as those in the observed data, and the x-axis shows the number of swaps that took
 319 place during the permutation procedure. The long vertical blue line indicates the burn-in period for the
 320 randomization swaps (2000 swaps before a null network was stored) and the short vertical blue lines
 321 show the points at which the following 999 null networks were stored (i.e., every 100 swaps).



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324 **Supplementary Figure 9. Observed correlation (and the correlation expected from the cross-**
 325 **network permutations) between all of the information-sharing networks.** Horizontal lines show
 326 the observed values from the actual networks (red = observed values are above the permutations,
 327 black = observed values are within the range of the permutations, purple = observed values are below
 328 the permutations). Polygon distributions show those generated by permutations (dark blue = cross-
 329 network swap that maintains the no. of nominations each individual makes and also the number of
 330 times each individual was nominated, but swaps the network these were made within whilst maintain

331 the number of times each network was nominated as overall, light blue = conservative cross-network
332 swap that is the same as dark blue, but also maintains the number of networks each dyad nominated
333 each other for – but changes those networks (same as a gbi permutation but on the dyad-by-network
334 edges). For details on information-sharing networks refer to Table 1 in the main text.

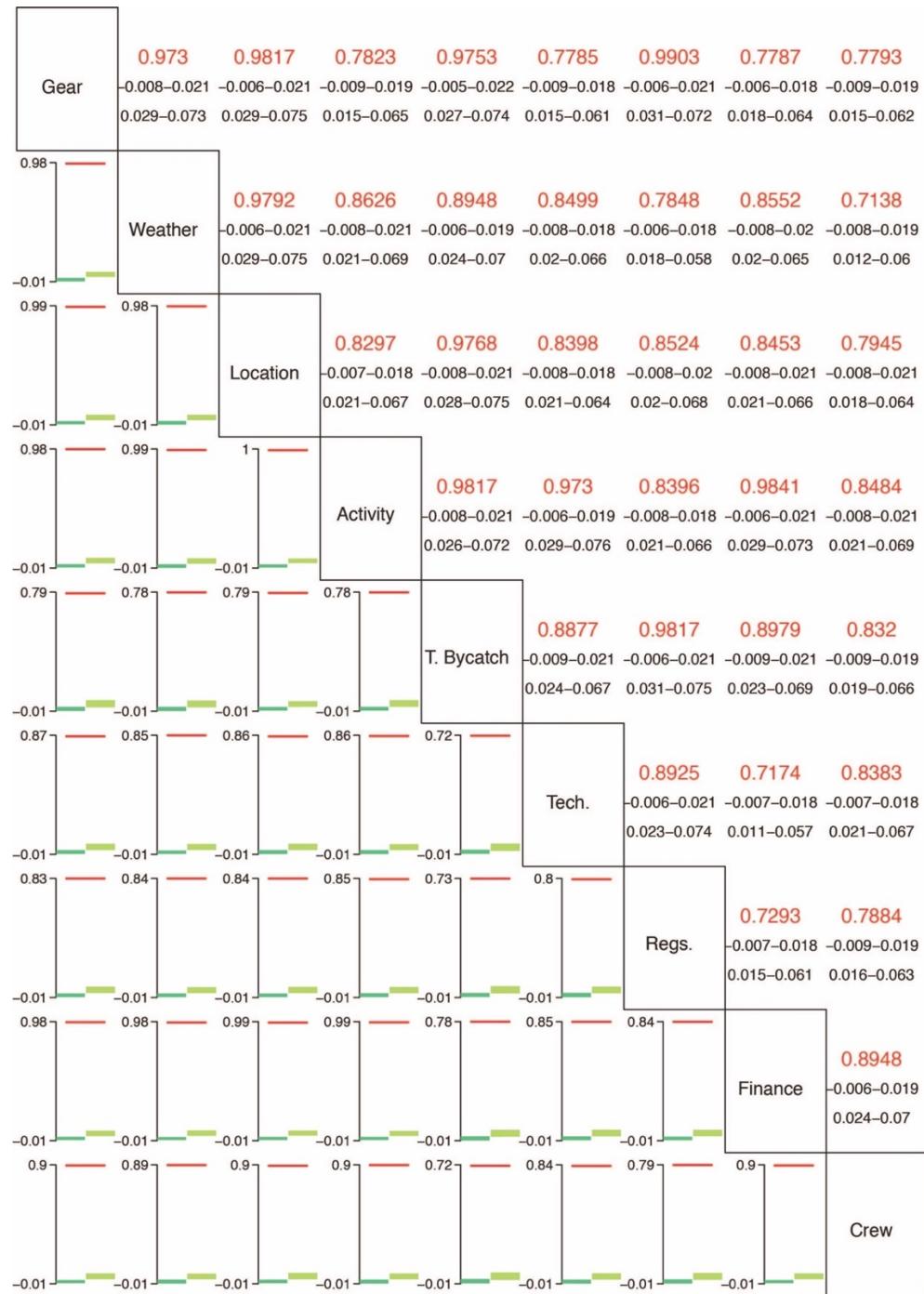
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342 **Supplementary Figure 10. Observed correlation (and the correlation expected from the edge**
 343 **permutations) between all of the information-sharing networks.** Horizontal lines show the
 344 observed values from the actual networks (red = observed values are above the permutations, black =
 345 observed values are within the range of the permutations, purple = observed values are below the
 346 permutations). Polygon distributions show those generated by permutations (dark green = outgoing
 347 edge permutation that maintains the no. of nominations each individual makes, light green = edge

348 swap that maintains the no. of nominations each individual makes and also the number of times each
349 individual was nominated). For details on information-sharing networks refer to Table 1 in the main
350 text.

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375 **Supplementary Table 1. Respondent-to-respondent network summary statistics.** Respondents
 376 nominated up to 10 individuals that included other skippers in their community but also non-skipper
 377 community members that might be deemed valuable to their fishing success. This study only analysed
 378 respondent-to-respondent data but the full network links (i.e., skipper to any relation links) across
 379 information-sharing networks are included in table section B.

380

(A) Respondent-to-respondent network data

	Number
Total no. of ties within all networks	3720
Total no. of ties of one or more networks	427
Total no. of eligible respondents for survey	168
Total no. of respondents surveyed	165
Levels of information-sharing networks	9
Mean number of networks nominated per nominee	7.7
Mean incoming ties of one or more network per respondent	3.7
Mean outgoing ties of one or more network per respondent	2.8
Range of networks nominated per nominee	1 to 9
Range of outgoing ties of one or more network	1 to 8
Range of incoming ties of one or more network	1 to 15

(B) Ties across networks

	Resp-resp	Full network
All	427	1102
Fish location & catch sites	418	1033
Fishing activity	418	1047
Weather conditions	415	1055
Gear type	411	1029
Fishing finances	411	1020
Captain hiring crew and managing them	342	868
Vessel technology & maintenance	311	807
Fishery regulations	304	822
Sea turtle bycatch	263	708

381

382 **Supplementary Table 2. Measures of network structure with statistics describing in-assortment (in assort) and variance eccentricity (var eccent).**

383 Table includes the observed statistic and the statistic from the permutations as the mean, sd, 95% range from 2.5% (lq) to 97.5% (uq), and the p value (when
 384 compared to the observed stat).

385

stat	network	obs1	mean.sd.1	lq.uq.1	p1	mean.sd.2	lq.uq.2	p2
in assort	T.bycatch	0.0377	-0.0049 (0.0593)	-0.1228 to 0.1047	0.512	-0.0107 (0.0593)	-0.1316 to 0.1041	0.39
in assort	Gear	0.105	-0.0034 (0.0481)	-0.0921 to 0.0963	0.034	0.0052 (0.0486)	-0.088 to 0.1028	0.04
in assort	Weather	0.0932	-0.0089 (0.0494)	-0.1113 to 0.0882	0.044	0.0032 (0.0485)	-0.0859 to 0.0983	0.068
in assort	Location	0.1069	-0.005 (0.0471)	-0.095 to 0.0858	0.016	0.0057 (0.0467)	-0.0813 to 0.096	0.022
in assort	Activity	0.1038	-0.0048 (0.0452)	-0.0944 to 0.0822	0.022	0.004 (0.047)	-0.0885 to 0.1024	0.042
in assort	Tech	0.1143	-0.0091 (0.0547)	-0.1108 to 0.1041	0.032	0.0087 (0.0565)	-0.0982 to 0.1189	0.064
in assort	Regs	0.1246	-0.0051 (0.0566)	-0.113 to 0.1077	0.02	0.0038 (0.0507)	-0.0953 to 0.1038	0.026
in assort	Finance	0.1002	-0.0056 (0.0481)	-0.0995 to 0.0915	0.036	0.0049 (0.0473)	-0.0882 to 0.0976	0.048
in assort	Crew	0.1891	-0.0084 (0.0528)	-0.109 to 0.0956	0	0.0198 (0.0525)	-0.0821 to 0.1247	0
var eccent	T.bycatch	14.71	41 (13.5)	22.41 to 73.73	0.006	22.66 (5.335)	15.58 to 36.53	0.02
var eccent	Gear	16.24	8.819 (2.326)	5.209 to 14.11	0.016	12.84 (1.592)	10.67 to 16.65	0.066
var eccent	Weather	19.28	8.717 (2.206)	5.101 to 13.63	0.004	12.76 (1.563)	10.77 to 16.56	0.012
var eccent	Location	18.96	8.366 (2.17)	4.778 to 13.41	0.002	12.39 (1.397)	10.45 to 15.72	0.008
var eccent	Activity	19.6	8.595 (2.251)	5.068 to 13.99	0.004	12.48 (1.362)	10.52 to 15.63	0.002
var eccent	Tech	19.15	27.14 (7.831)	16.93 to 46.5	0.202	20.19 (3.709)	14.95 to 29.94	0.894
var eccent	Regs	22.85	30.48 (9.694)	18.21 to 57.39	0.392	19.87 (3.483)	15.04 to 28.13	0.342
var eccent	Finance	17	8.884 (2.249)	5.369 to 14.09	0.014	12.74 (1.405)	10.65 to 16.2	0.03
var eccent	Crew	24.46	19.43 (5.77)	11.87 to 34.63	0.264	16.87 (2.629)	13.3 to 23.2	0.022

386

387 **Supplementary Table 3. Measures of network structure with statistics describing assortativity coefficient for outgoing links (out assort), mean node**
 388 **eccentricity (mean eccent) and variance in node betweenness (var between).** Table includes the observed statistic and the statistic from the permutations as
 389 the mean, sd, 95% range from 2.5% (lq) to 97.5% (uq), and the p value (when compared to the observed stat).

stat	network	obs1	mean.sd.1	lq.uq.1	p1	mean.sd.2	lq.uq.2	p2
out assort	turtle	-0.0563	-0.008 (0.0614)	-0.1223 to 0.1124	0.424	-0.0208 (0.0576)	-0.1297 to 0.0968	0.534
out assort	gear	0.0205	-0.004 (0.049)	-0.0981 to 0.0937	0.584	-0.0055 (0.0471)	-0.0983 to 0.0882	0.544
out assort	weather	0.0517	-0.0078 (0.0475)	-0.1025 to 0.0854	0.212	-0.0044 (0.0471)	-0.1021 to 0.0911	0.226
out assort	loc	0.0129	-0.0078 (0.05)	-0.1028 to 0.0914	0.65	-0.0091 (0.0465)	-0.1004 to 0.0794	0.626
out assort	activ	0.0152	-0.0039 (0.0506)	-0.1043 to 0.1003	0.662	-0.0082 (0.046)	-0.0926 to 0.0818	0.628
out assort	tech	0.0425	-0.0095 (0.055)	-0.1145 to 0.1023	0.36	-0.004 (0.0523)	-0.1058 to 0.0985	0.384
out assort	regs	0.0129	-0.0049 (0.0592)	-0.119 to 0.1104	0.766	-0.01 (0.0533)	-0.1103 to 0.0955	0.648
out assort	financ	0.0232	-0.0053 (0.0481)	-0.0982 to 0.0863	0.58	-0.0044 (0.0479)	-0.1062 to 0.0856	0.544
out assort	capt	0.0735	-0.0057 (0.0529)	-0.1101 to 0.1018	0.13	-0.0025 (0.0493)	-0.0998 to 0.0976	0.136
mean eccent	turtle	3.309	8.754 (1.498)	5.988 to 11.82	0	4.63 (0.5432)	3.776 to 5.867	0
mean eccent	gear	4.546	8.259 (0.6509)	7.242 to 9.782	0	4.701 (0.2994)	4.194 to 5.37	0.612
mean eccent	weather	5	8.28 (0.6363)	7.285 to 9.813	0	4.708 (0.2935)	4.254 to 5.352	0.26
mean eccent	loc	4.994	8.216 (0.674)	7.157 to 9.808	0	4.735 (0.2826)	4.261 to 5.376	0.336
mean eccent	activ	5.242	8.266 (0.6777)	7.218 to 9.813	0	4.792 (0.2752)	4.333 to 5.449	0.13
mean eccent	tech	4.358	9.346 (1.164)	7.472 to 12.15	0	5.188 (0.4979)	4.357 to 6.297	0.052
mean eccent	regs	4.461	9.576 (1.21)	7.652 to 12.44	0	5.233 (0.4729)	4.46 to 6.249	0.056
mean eccent	financ	4.933	8.259 (0.6506)	7.242 to 9.891	0	4.809 (0.2827)	4.291 to 5.431	0.572
mean eccent	capt	5.042	9.198 (0.9606)	7.606 to 11.47	0	5.18 (0.4175)	4.473 to 6.116	0.79
var between	turtle	55170	321700 (147600)	61560 to 633200	0.042	147500 (38430)	86450 to 234900	0
var between	gear	159300	285100 (39540)	218300 to 375500	0	178100 (21950)	140500 to 226800	0.376
var between	weather	214700	290400 (40420)	223900 to 381500	0.016	182700 (20020)	147900 to 224800	0.136
var between	loc	197800	280400 (42240)	213200 to 372600	0.01	180200 (19870)	143900 to 224500	0.344
var between	activ	239300	284800 (42540)	215400 to 377400	0.264	184700 (20600)	148500 to 228600	0.02
var between	tech	158600	428400 (108200)	272400 to 700500	0	206500 (40180)	143000 to 293500	0.18
var between	regs	129000	454000 (125800)	275500 to 768900	0	214900 (38420)	149300 to 305400	0.004
var between	financ	234000	286700 (40670)	222700 to 378600	0.136	182200 (20930)	144800 to 226000	0.034
var between	capt	174800	405800 (84840)	282100 to 606900	0	215600 (32690)	161400 to 287600	0.176

390

391 **Supplementary Table 4. Measures of cross-network comparisons with statistics describing variance in node eccentricity (var eccent) and mean node**
 392 **eccentricity (mean eccent).** Table includes the observed statistic and the statistic from the permutations as the mean, sd, 95% range from 2.5% (lq) to 97.5%
 393 (uq), and the p value (when compared to the observed stat).

394

stat	network	obs1	mean.sd.1	lq.uq.1	p1	mean.sd.2	lq.uq.2	p2
var eccent	turtle	14.71	31.63 (10.14)	16.98 to 56.19	0.02	30.91 (10.73)	15.05 to 59.44	0.038
var eccent	gear	16.24	16.74 (1.665)	13.95 to 20.67	0.874	16.72 (1.661)	14.04 to 20.62	0.926
var eccent	weather	19.28	16.26 (1.372)	13.96 to 19.57	0.058	16.33 (1.449)	13.94 to 19.76	0.088
var eccent	loc	18.96	16.13 (1.282)	13.94 to 19.36	0.06	16.09 (1.308)	13.89 to 19.38	0.074
var eccent	activ	19.6	16.07 (1.207)	13.92 to 18.98	0.034	15.98 (1.247)	13.9 to 19.2	0.042
var eccent	tech	19.15	28.23 (6.785)	18.65 to 45.21	0.07	28.42 (7.11)	18.88 to 47.12	0.068
var eccent	regs	22.85	29.16 (7.625)	17.81 to 47.81	0.338	29.18 (8.034)	18.08 to 51.99	0.382
var eccent	financ	17	16.66 (1.606)	14.07 to 20.22	0.75	16.76 (1.651)	13.98 to 20.8	0.8
var eccent	capt	24.46	24.1 (4.986)	17.49 to 35.94	0.774	24.17 (4.963)	17.3 to 34.21	0.786
mean eccent	turtle	3.309	4.831 (0.7783)	3.491 to 6.455	0.022	4.771 (0.824)	3.381 to 6.498	0.04
mean eccent	gear	4.546	4.824 (0.2212)	4.424 to 5.285	0.216	4.824 (0.221)	4.418 to 5.321	0.182
mean eccent	weather	5	4.786 (0.1861)	4.442 to 5.2	0.238	4.796 (0.1884)	4.442 to 5.218	0.262
mean eccent	loc	4.994	4.786 (0.1698)	4.46 to 5.188	0.202	4.779 (0.177)	4.442 to 5.182	0.208
mean eccent	activ	5.242	4.776 (0.1616)	4.46 to 5.134	0.03	4.765 (0.1657)	4.448 to 5.127	0.03
mean eccent	tech	4.358	5.207 (0.5926)	4.194 to 6.492	0.102	5.208 (0.5861)	4.206 to 6.485	0.102
mean eccent	regs	4.461	5.194 (0.6411)	4.024 to 6.558	0.212	5.184 (0.6511)	4.073 to 6.661	0.228
mean eccent	financ	4.933	4.817 (0.2121)	4.448 to 5.267	0.522	4.831 (0.216)	4.436 to 5.328	0.548
mean eccent	capt	5.042	5.152 (0.4853)	4.351 to 6.285	0.904	5.159 (0.4691)	4.364 to 6.103	0.852

395

396 **Social network analysis questionnaire (English)**

397 **Individual socio-demographic information**

398 First, I'm going to ask you a few questions about yourself. Note that your individual responses to this
399 survey will remain confidential and we will only use the data collected in aggregate form.

400

401

Survey ID

Da

403

404 **Full name**

Nickname

405 **Gender** Male Female

406 **Fisher / decision maker status:** Skipper Vessel owner Skipper AND Owner

407 **Plate number**

Name of boat

408

409 **Q1)** What is your age? _____

410 **Q2)** Do you live in San José. Y _____ yrs., N, where do you live? _____ region / city

411 **Q3)** If < 5 years, where did you live before and why did you move here? _____

412 **Q4)** What generation of gillnet fisherman in San Jose are you? _____

413

414 **For boat owners that are not skippers:**

415 **Q5)** Were you formally a gillnet captain?

416 No

417 Yes (please specify when you stopped fishing) _____

418 **Q6)** Which best describes your situation:

419 My family fish with my boat as we divide the profits evenly. Or some other
420 percentage _____

421

422 I hire my boat to non-family members and receive a percentage of the catch profit: _____

423 **For skippers and skippers AND boat owners**

424

425 **Q7)** How many years have you been fishing? _____

426 **Q8)** Do you launch or land at any other ports?

427 No

428 Yes (please specify) _____

429

430 **Q9)** During which months did you not fish last year? _____

431

432 **Q10)** What is the principal net that you use? Trammel, Lineal, Other: _____

433 Surface / driftnet

434 Mid-water net

435 Bottom net

436 Other net type (please specify) _____

437 **Q11)** Do you ever switch net types from your main net type?

438 No

439 Yes (please explain to what, and under what circumstances) _____

440

441

442

443 **Q12)** What are your three main target species? 1. _____

444

445 2. _____

446

447 3. _____

448 **For everyone:**

449 **Q13)** Which of the following best describes you?

450 President of a gremio / social group (which) _____

451

452 Board member of a gremio / social group (which) _____

453

454 Member of gremio / social group (which) _____

455

456 I'm not a member of any gremio / social group (Individual owner operator)

457 **Q14)** What is your highest level of education?

458 No formal education

459

460 Primary school, please specify if completed _____

461

462 Secondary school, please specify if completed _____

463

464 Trade or technical certificate / fishing course, please specify if completed _____

465

466 University degree, please specify if completed _____

467 **[Personal income]**

468 **Q15)** Is fishing your primary occupation/source of income?

469 Yes

470 No (please specify what is) _____

471
472 **Q16)** How much do you spend on fishing trips per month (on average)? Summer _____ Winter _____

473
474 **Q17)** How many days a month (in average) do you spend on fishing trip? Summer _____ Winter _____

475
476
477 **Q18)** What is your take-home monthly income (in soles) after all expenses in:

478 Summer: Max: _____ Winter: _____ Max: _____
479
480 _____

481
482 Average: _____ Average: _____
483
484 Min: _____ Min: _____
485
486

487 **[Household income]**

488 **Q19)** Which of the following household descriptions best fits you?

489 Couple with children – with some children still living at home
490 Couple with children – with all children having left home
491 Couple without children
492 Single with children
493 Single without children

494 **Q20)** Are you the main wage earner in your household?

495 No
496 Yes
497

498 **Q21)** How many people are currently living in your household? _____

500 **Q22)** Of these, how many are fishermen? _____

501
502

503 **Q23)** Are there any other wage earners in your household that are not fishermen?

504 No
505 Yes (what jobs do they do?) _____
506

507 **Q24)** What percentage of your household income (including all wage earners) comes from fishing?

508 0- 20% 21- 40% 41- 60% 61- 80% 81-100% All Don't know / rather not say

509

510

511

512 **Section B: Social Network Analysis structured questionnaire**

513 We need you to think about the people from San Jose that you share useful information about
514 fisheries with; consider those you think may influence your fishing success. Remember that the
515 shared information and names will remain anonymous and will not be revealed. This will help us
516 understand how the information flows between fishermen.

517 Please consider relationships that you have had with other vessel owners, captains, owner/captains
518 (owners who also captain their vessel), other fishery leaders, fishery management officials, members
519 of the scientific or NGO community, boat launching / landing support, fish transport associations, fish
520 sellers/market operators, your family and friends, and any other people you have fished with, or
521 shared information with about fishing over the last 5 years.

522

523 **Q25)** Please identify up to 10 individuals (providing first and last names, and known nicknames) that you *exchange useful information* with about fishing that
 524 you consider *valuable to your fishing success*.
 525

Full name	Nickname	Rel	Crew	Meet	tMeet	Often	Topic of conversation									Value
							I	II	III	IV	V	VI	VII	VIII	IX	
1																
2																
3																
4																
5																
6																
7																
8																
9																
10																

526
 527 Rel = Relation: A) Professional acquaintance, B) Friend, C) Family
 528 Crew = Crew member: Y / N
 529 Meet = How did you meet: A) family member, B) through a friend, C) through fishing, D) from a family member, E) Other: _____
 530 tMeet = How long have you known this person: A) <1 yr, B) 1-5 yrs, C) >5 yrs
 531 Often = How often do you share useful information about aspects of fishing with this person? A) 1-3 times/yr, B) 1-3 times/month, C) 1-3 times/week or
 532 more
 533 I: Gear type (i.e. Changes, technology, maintenance)
 534 II: Weather conditions
 535 III: Fish location / catch sites
 536 IV: Fishing activity (How many people fishing, who is fishing, who caught what, etc.)
 537 V: Turtle bycatch
 538 VI: Vessel technology / maintenance
 539 VII: Fishery regulations (laws, rules)
 540 VIII: Fishing finances (market prices, loans, fines, penalties)
 541 IX: Hiring new crew / captain
 542 Value: In general, how valuable do you feel the information that you exchange with this individual is to your fishing success? A) Very valuable, B) somewhat
 543 valuable, C) a little valuable
 544 To finish up with the network analysis, I have four more questions on bycatch and new gear uptake

545 **Q26**) Which of the people you've identified is the most influential to you when you are considering
546 making changes to your fishing gear?

547
548

549 **Q27**) Which of the people you've identified is the most influential to you in (potentially) deciding
550 about changing the way you fish (e.g. changing your behaviour such as shorter soak time)?

551
552

553 **Q28**) What do you think about taking on new technologies to reduce bycatch of turtles and dolphins?
554 (-1 Negative, 0 Neutral, +1 Positive)

555
556

557 **Q29**) Are you aware of the work that the NGO ProDelphinus is undertaking with a few fishermen
558 here in San Jose to help reduce the number of turtles and dolphins that are captured in nets? Do you
559 know about the technologies that they are using?

560
561

562 **Q30**) Do you think the Orca underwater acoustic alarm used to deter dolphins attract sea lions to your
563 nets?

564 No
565 Yes
566 I don't know

567 **Q31**) Do you think lights on your nets to deter turtles attract sea lions to your nets?

568 No
569 Yes
570 I don't know

571 If you have any comments on this survey or about information sharing between fishermen within the
572 San José community, please tell us or write them below.



573
574

Thank you very much for your time and help in this survey

575 **Social network analysis questionnaire (Spanish)**

576 **Información socio-demográfica individual**

577 Primero, voy a preguntarte acerca de ti. Ten en cuenta que las respuestas individuales en esta encuesta
578 se mantendrán confidenciales y solo usaremos la información de forma agregada.

579 **ID de la encuesta** **Fecha**

580

581 **Nombre completo** **Apodo**

582 **Género** Masculino Femenino

583 **Estado en toma de decisiones:** Patrón Dueño de embarcación Ambos

584 **Número de matrícula** **Nombre de la embarcación**

586

587 **Q1) ¿Cuál es tu edad?** _____

588 **Q2) ¿Vives aquí? Y** _____ yrs, **N** _____ región / ciudad

589 **Q3) Si < 5 años, ¿dónde vivías antes y por qué te mudaste aquí?** _____

590 **Q4) ¿Qué generación de pescador de redes de enmallé de San José eres tú?** _____

591 **Para dueños de embarcaciones que no son PATRONES.**

592 **Q5) ¿Fuiste alguna vez formalmente un patrón?**

593 No

594 Si (¿hace cuantos años dejaste de pescar?) _____

595 **Q6) ¿Cuál describe mejor tu situación?:**

596 Mi familia pesca con mi bote, dividimos las ganancias igual. Otro porcentaje?

597 _____

598

599 Rento mi bote a un ajeno y recibo un porcentaje de la ganancia, cuanto? _____

600 **Solo para PATRONES y PATRONES que son ARMADORES**

601

602 **Q7) ¿Cuántos años llevas pescando?** _____

603 **Q8) ¿Embarcas o desembarcas de otros puertos?**

604 No

605 Sí (por favor especifica) _____

606

607 **Q9)** ¿En que meses descansaste el año pasado? _____

608 **Q10)** ¿Cuál es el tipo principal de red de enmalle que usas? Trasmallo, Lineal Otro: _____

609 Red de superficie / red de deriva

610 Red de mediagua

611 Red de fondo

612 Otro tipo de red (por favor especifica) _____

613 **Q11)** ¿Cambias tu tipo de red principal por otros?

614 No

615 Sí (por favor especifica a qué, y debido a qué) _____

616

617

618

619 **Q12)** ¿Cuáles son tus 3 objetivos principales de pesca? 1. _____

620

621 2. _____

622

623 3. _____

624

625

626

627 **Para todos**

628

629 **Q13)** ¿Cuál de los siguientes te describe mejor?

630 Presidente de un gremio / grupo social (cuál) _____

631

632 Miembro de consejo de gremio / grupo social (cuál) _____

633

634 Miembro de gremio / grupo social (cuál) _____

635

636 No soy agremiado / no pertenezco a grupos sociales (Dueño operador individual) _____

637 **Q14)** ¿Cuál es tu nivel educativo?

638 Sin educación formal

639

640 Primaria (por favor especificar si completó) _____

641

642 Secundaria (por favor especificar si completó) _____

643

644 Técnico / capacitado en pesca (por favor especificar si completó) _____

645

646 Universitario (por favor especificar si completó) _____

647

648 **[Ingresos personales]**

649 **Q15)** ¿Es la pesca tu principal ocupación / fuente de ingresos?

650 Sí

691

692 **Sección B: Cuestionario estructurado de Análisis de Red Social**

693 Piensa con quienes intercambias INFORMACION UTIL de pesca en San Jose y que sientes que

694 PODRIA INFLUENCIAR en que te vaya bien en la pesca. Los nombres y la informacion que des se

695 mantendran en anonimato y no sera revelada. Esto servira para saber como fluye la informacion entre

696 pescadores.

697 Recuerda a: otros dueños de embarcaciones, capitanes, otros líderes pesqueros, oficiales de manejo

698 pesquero, científicos o ONGs, embarcadores/ayudantes de embarque y desembarque, asociaciones de

699 chalaneros, vendedores de pescado/operadores de mercado, tu familia y amigos, y todas las otras

700 personas con las que hayas pescado o compartido información de pesca en los últimos 5 años.

701

702 **Q25) Social Network Analysis questionnaire (Spanish).** Por favor identifica hasta 10 individuos (nombres y apellidos, no solo apodos) con los que
 703 *intercambias información útil* acerca de la pesca que consideres *valioso para tu éxito pesquero*.
 704

Nombre completo	Apodo	Rel	Crew	Meet	tMeet	Often	Tema de conversaciòn									Valor
							I	II	III	IV	V	VI	VII	VIII	IX	
1																
2																
3																
4																
5																
6																
7																
8																
9																
10																

705
 706 Rel = Relacion: A) Profesional conocido, B) Amigo, C) Familiar
 707 Crew = Colega-tripulante, Y / N
 708 Meet = Como lo conociste: A) familiar, B) por un amigo, C) a traves de la pesca, D) por un familiar, E) OTRO: _____
 709 tMeet = Cuanto tiempo lo conoces: A) <1 año, B) 1-5 años, C) >5 años
 710 Often = Que tan seguido comparten info: A) 1-3 veces/año, B) 1-3 veces/mes, C) 1-3 veces/semana o más
 711 I: tipo de arte (i.e. cambios, tecnologia, mantenimiento)
 712 II: condiciones climaticas
 713 III: ubicacion de los peces y sitios de captura
 714 IV: actividad pesquera (cuanto, quienes estan pescando, que estan pescando, quien cogio que, etc.)
 715 V: Captura incidental de tortuga
 716 VI: tecnologia y mantenimiento de la nave
 717 VII: regulaciones pesqueras (leyes, reglas)
 718 VIII: finanza pesquera (precios del Mercado, prestamos, multas, penalidades)
 719 IX: Contratacion de tripulantes o capitán
 720 Value: Que tan valiosa es la informacion que intercambias: A) muy valiosa, B) algo valiosa, C) un poco valiosa
 721 Solo para terminar el análisis de red social, tengo cuatro preguntas más acerca de pesca incidental y aceptación de nuevos artes de pesca.

722 **Q26)** ¿Cuál de las personas que has identificado es la más influyente para ti cuando se trata de hacer
723 cambios en los artes de pesca?

724

725 **Q27)** ¿Cuál de las personas que has identificado es la más influyente para ti en (potencialmente)
726 decidir cambiar la forma en la que pescas (e.g. cambiar el momento y duracion que pones la red)?

727

728 **Q28)** ¿Qué opinas de adoptar nuevas tecnologias para reducir la captura incidental de tortugas y
729 delfines? (-1 , 0 , +1)

730

731 **Q29)** ¿Estás al tanto del trabajo que la ONG ProDelphinus viene llevando a cabo con un pequeño
732 grupo de pescadores aquí en San José para ayudar a reducir el número de tortugas y delfines que son
733 capturados en las redes? Conoces las tecnologias que usan?

734

735

736 Si tienes comentarios acerca de esta encuesta por favor dinos o escríbelos en el cuadro.



737

738 Muchas gracias por tu tiempo y colaboración con esta encuesta

739 **Supplementary Literature cited**

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