

Large Scale Marine Protected Areas

Juan Carlos Villaseñor-Derbez*

juancarlos@ucsb.edu

John Lynham†

lynham@hawaii.edu

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Abstract

Large-scale Marine Protected Areas (LSMPAs) have seen a significant increase over the last years. Fishing effort is effectively eliminated within these protected areas upon implementation. The benefits of reducing effort have been largely studied, but little empirical works evaluate how vessels react and redistribute after an MPA is created. The economic and ecological implications of displacing fishing effort are not yet fully understood. We use identification of fishing activity via Automatic Identification Systems (AIS) and causal inference techniques to provide the first analysis of behavioral changes and spatial redistribution of tuna purse seiners due to the implementation of a Large Scale Marine Protected Area in the Pacific Ocean. Our work provides three main findings: 1) aggregate fishing effort remains relatively unaffected; 2) vessels that fished inside the protected area redistribute to adjacent waters; and 3) we observe a crowding effect for the first months after implementation. Our results not only provide an impact evaluation of the effect of LSMPAs on fishing activity, but provide insights into vessel redistribution dynamics, which may have ecological and economic implications for marine conservation and fisheries management. As countries continue to implement LSMPAs as a way to reach the stated 10% target of ocean protection, managers should consider how fishing effort will change in space and through time to ensure that fishing effort is not just displaced elsewhere, leading to overfishing in adjacent waters. While LSMPAs can provide a wide range of benefits, their implementation must be accompanied with traditional fisheries management to maximize their effectiveness.

We do not observe an increase in daily fishing hours for vessels that used to fish within PIPA, compared to a control group of vessels. Thus, for vessels that have chosen to continue to operate following the closure, we do not observe evidence that they are being forced to exert more effort as a result of the closure. Unfortunately, without detailed data on vessel catch, we cannot conclude whether vessels are better or worse off financially following the closure.

Our results suggest that PIPA has displaced some vessels to the high seas just outside the EEZ of Kiribati. This might imply a drop in demand for Kiribati fishing rights and a decline in revenue. We also uncover evidence that there is more congestion immediately following the closure, but this is declining over time.

*Bren School of Environmental Science & Management, University of California at Santa Barbara, Santa Barbara, CA

[†]Department of Economics, University of Hawaii at Manoa, Honolulu, HI

33 Next steps: I think we need to make the buffers around the EEZs 5 degrees instead of 1 degree
34 (redistribution of effort section) Might change the specification of the redistribution regression

35 1 Introduction

36 Marine Protected Areas (MPAs) are intended to safeguard parts of the ocean from fishing and other
37 extractive activities. Current international goals aim to protect 10% of the ocean environment by
38 2020. In an effort to meet this target, there has been a rapid increase recently in MPA coverage
39 (Wood et al., 2008; Sala et al., 2018), largely driven by a small number of Large Scale Marine
40 Protected Areas (LSMPAs).¹). Today, a small number of LSMPAs make up at least 80% of the
41 managed areas in the ocean (Toonen et al., 2013).

42 Given the relatively recent establishment of most LSMPAs, very little is known about their human
43 dimensions and implication for fisheries (Gray et al., 2017). As with smaller MPAs, it is important
44 that we understand the socioeconomic implications of management interventions. One issue of
45 particular importance is that of the displacement or redistribution of fishing effort, which may
46 influence the outcomes of an MPA (Smith and Wilen, 2003). Theoretical models make different
47 assumptions about the ways in which fishers will reallocate fishing effort after an area closure,
48 which often determine whether or not MPAs produce negative or positive impacts on fisheries.
49 The existing empirical literature focuses on small MPAs and has been criticized for confounding
50 correlation with causation (?).

51 Recent advances in satellite tracking technologies and near real-time identification of fishing activity
52 provide us with an opportunity to ask, how do fishers respond to the implementation of an LSMPA?
53 Does effort increase or decrease and where does it go? We use identification of fishing activity via
54 Automatic Identification Systems (AIS) and causal inference techniques to describe the behavioral
55 changes and spatial redistribution of the industrial tuna purse seine fleet due to the implementation
56 of the Phoenix Islands Protected Area. We use the same data to hypothesize what might be the
57 impacts of the proposed Palau National Marine Sanctuary.

58 Our work is novel in the sense that it provides empirical evidence of the effect of Large Scale Marine
59 Protected Areas on fishing behavior and distribution and can help guide future interventions.
60 Understanding how effort is displaced from LSMPAs might provide insights into how distant water
61 fishing fleets would react to a high seas closure.

62 The paper is organized as follows: Section 2 provides more information on Large Scale Marine
63 Protected Areas and some background on empirical studies of effort redistribution. An overview
64 of the Nauru Agreement and associated countries, a description of the fleet that operates in the
65 region, and a brief history of PIPA and PNMS is also included. Section 3 describes our data and
66 identification strategy. Section 4 presents our results, and Section 7 discusses our results.

¹See, for example, Game et al. (2009), Singleton and Roberts (2014), Boonzaier and Pauly (2016), McCauley et al. (2016), and Alger and Dauvergne (2017).

67 2 Background

68 2.1 Large Scale Marine Protected Areas

69 The cutoff at which an MPA is considered to be a LSMPA ranges from areas larger than 30,000
70 km² as defined by De Santo (2013) or areas larger than 250,000 km², as defined by (Toonen et al.,
71 2013). Figure 1 shows LSMPAs that meet the latter condition, and are also fully no-take. LSMPAs
72 are often implemented in the pelagic environment, where the dominant human activity is industrial
73 fishing (Gray et al., 2017; Kroodsma et al., 2018). The early literature on LSMPAs focused on
74 the inherent challenges and difficulties that come with a pelagic environment. Kaplan et al. (2010)
75 claimed that very large MPAs would result in excessive opportunity costs and that these would be
76 difficult to enforce. Game et al. (2009) suggested that most of the challenges could be overcome
77 with the incorporation of technology, in what then became known as Dynamic Ocean Management
78 (Maxwell et al., 2015).²

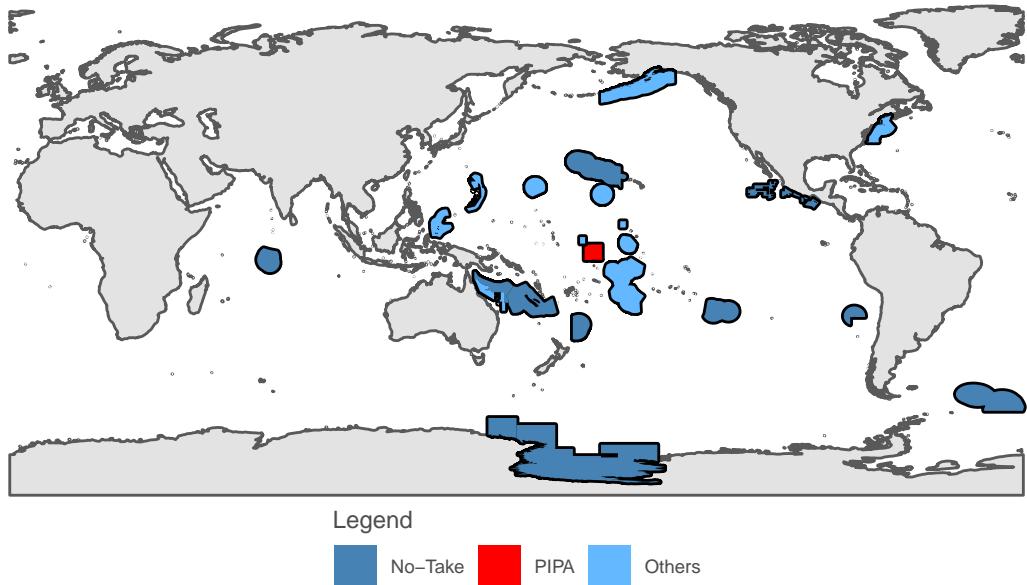


Figure 1: Large Scale Marine Protected Areas. The map shows all areas larger than 250,000 Km². Areas in dark blue meet at least one of these conditions: reported no take area is greater than 0, are categorized as IUCN Ia or Ib, their designated English name is ‘Protected Area’.

79 LSMPAs were erroneously assumed to have little social implications due to their remoteness. How-
80 ever, there have been calls to incorporate the human dimensions into LSMPAs management and
81 evaluation (Agardy et al., 2011; Gray et al., 2017). Most research incorporating these dimensions
82 has focused on governance and enforcement of LSMPAs (*i.e.* Alger and Dauvergne (2017); Christie
83 et al. (2017)), but they are yet to be the focus of economic analyses (Gray et al., 2017). Overall,
84 there has been little empirical work regarding LSMPAs. Recent technological advances in vessel-
85 detection systems allows for the discovery and advancement of many important facets of LSMPAs.

²See Singleton and Roberts (2014), who provide an objective discussion of the pros and cons of LSMPAs.

86 For example, McDermott et al. (2018) show that the anticipation of a LSMPA can lead to preemptive overfishing, which can erode or delay the expected benefits of the intervention. White et al.
87 (2017) combine shark tags and vessel-tracking data to demonstrate that the fairly large Palmyra
88 Atoll National Wildlife Refuge (54,000 Km²) protects two thirds of the tagged grey reef sharks
89 by effectively excluding fishing effort. More recently, (Bradley et al., 2018) use similar data to
90 highlight cases of potential illegal shark fishing *inside* a 2 million km² shark sanctuary. To date,
91 no studies have evaluated the displacement of fishing effort due to LSMPA implementation.

92 Spatial closures of this magnitude are likely to induce changes in fishers' behavior. Theoretical
93 models of fishing effort redistribution range from the simplistic assumption that effort inside the
94 bounded region disappears, to spatially explicit models that reallocate fishing effort based on habitat
95 characteristics, presence of other vessels, and expected returns (Smith and Wilen, 2003; Hilborn
96 et al., 2006). However, these focus on the long term optimal equilibrium, and redistribution of
97 fishing effort may not always be optimal, especially over the first few years (Stevenson et al., 2013).

98 The empirical research that has been done in smaller sized MPAs suggests that resource users
99 may show idiosyncratic responses. For example, Stevenson et al. (2013) show that a network of
100 MPAs displaced fishing effort farther away from ports, resulting in higher *perceived* costs, and
101 increases in catch per unit effort. Cabral et al. (2017) analyze the redistribution of fishing and
102 non-fishing vessels following the implementation of a network of MPAs in California, and find that
103 commercial dive boats (???????) follow a fishing-the-line pattern, while some fishing boats follow
104 an ideal free distribution. More recently Elahi et al. (2018) used satellite tracking data to show
105 that a temporal spatial closure caused trawlers to maintain effort but apply it more intensively
106 elsewhere, particularly along the borders and closer to shore. The way in which fishers react to a
107 spatial closure can have major implications for its outcome (Smith and Wilen, 2003; Hilborn et al.,
108 2006), highlighting the need to understand how fishers react to the implementation of LSMPAs,
109 how fishing effort changes, and how it is spatially redistributed. All these closures took place
110 within Exclusive Economic Zones, where other regulations exist. This may not always be the case
111 for LSMPAs, where often the entire EEZ is converted into a LSMPA, leaving fishers with the option
112 of moving to the high seas or other countries' EEZs.

114 2.2 Nauru agreement and the Phoenix Islands Protected Area

115 The Nauru Agreement was established in 1982 by a select group of Pacific island nations to manage
116 their important tuna resources more effectively. Parties to the Nauru Agreement (PNA) Members
117 include the Federated States of Micronesia, Kiribati, the Marshall Islands, Nauru, Palau, Papua
118 New Guinea, the Solomon Islands, and Tuvalu. The Nauru Agreement regulated access of foreign
119 vessels (*i.e.* those from non-PNA countries). Holding ~80% of the historical purse seining grounds
120 within their Exclusive Economic Zones, PNA countries gained greater bargaining power when
121 providing fishing access to foreign fleets (Havice, 2010).

122 The cooperation that emerged under the Nauru Agreement allowed for subsequent agreements
123 that strengthened fisheries management, like the Palau Agreement, which limited the number
124 of purse seiners at 205 vessels from 1995-2007³. However, the most notable regulation is their
125 approach to manage fishing effort: a Vessel Day Scheme (VDS) implemented in 2007 (Havice, 2013).
126 This effectively modified how fishing effort was managed, from number of vessels under the Palau
127 Agreement to fishing hours. The VDS works as follows: Each year, scientific advisers recommend

³See Havice (2010) for a detailed description of the Nauru, Palau, and Federal States of Micronesia agreements.

128 a total number of fishing vessel-days per year. Hours are allocated to each PNA country based on
129 catch history, and they then use or sell fishing rights to other non-PNA countries (Aqorau et al.,
130 2018). While the effectiveness of this scheme has been debated in terms of meeting their fishery
131 management and conservation objectives, the licensing significantly contributes to the economy of
132 these island nations (Havice, 2010).

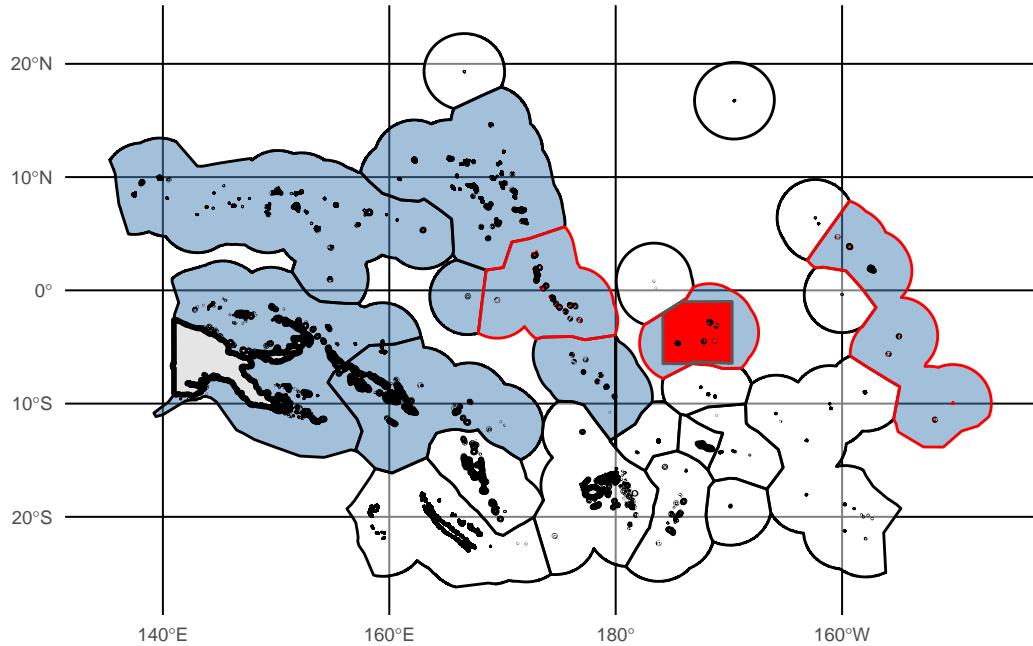


Figure 2: Map of the Exclusive Economic Zones (EEZs) of the region of interest. Countries that belong to the PNA are shown in blue, while empty polygons indicates all others. A red line indicates the Kiribati EEZ, and a solid red polygon delineates PIPA.

133 The main tuna species caught in the region are skipjack (*Katsuwonus pelamis*), yellowfin (*Thunnus*
134 *albacares*), albacore (*Thunnus alalunga*) and bigeye (*Thunnus obesus*). From these, the first two are
135 amongst the top-10 species represented in global fisheries production statistics, with 2016 catches
136 increasing relative to the 2005-2014 average (FAO, 2018). This region of the Pacific has historically
137 accounted for a large portion of tuna catches (Aqorau and Bergin, 1997). Today, the PNA controls
138 close to 50% of the global skipjack tuna production (PNA, 2018). A large portion of these catches
139 derive from purse seine vessels licensed under the VDS. Fishing vessels from Australia, New Zealand,
140 China, France, Korea, Japan, the Philippines, Taiwan, and the United States participate in the
141 purse-seining VDS.

142 One of the most notable and recent management interventions in the region is the implementation
143 of the Phoenix Islands Protected Area (PIPA) by the government of Kiribati. PIPA was first
144 declared in 2006, and established in 2008 with only 4% of the area declared as no-take. On January
145 1st, 2015, the no-take area within PIPA was expanded to a total area of 397,447 km², roughly 1.5
146 times the size of Ecuador. Figure 2 shows a map of the PNA countries and the Phoenix Islands
147 Protected Area.

148 The closure of such a large area in one of the most important fishing regions in the world provides
149 a unique opportunity to evaluate the behavioral responses and redistribution of fishing effort by

vessels that used to fish there. PIPA has been the focus of previous research showing that fishing effort is effectively reduced after implementation (McCauley et al., 2016; McDermott et al., 2018). To this, we pose two questions: How did individual vessels respond to the sudden exclusion of such a big area? Where did all the vessels go? And what does this imply for revenues from selling VDS for both Kiribati and the PNA as a whole?

A spatial closure might cause fishers to modify their behavior as they adapt to a new state of the world. For example, some may have to travel further distances to find new fishing grounds, increasing their fuel costs. If fishers had developed experience for fishing in particular sites, being excluded might impose a learning cost on them, as they identify new fishing grounds. This might result in increased search times. However, if the area of knowledge of a vessel is significantly larger than that of the spatial closure, they might already know other places to fish. In the next sections we describe the data and methods used to answer these questions.

3 Methods

This section is divided into two main parts. First, we provide a general description of AIS data and the process of identification of vessel-level fishing events performed by Global Fishing Watch.⁴ Alongside, we describe the subset of data used in our analysis. We also point out possible shortcomings in the data, or factors that must be considered in the analysis. We then move on to explain our empirical strategy for the identification of behavioral changes and the redistribution of fishing effort.

3.1 Data

Automatic Identification Systems (AIS) are on-board devices that provide at-sea safety and prevent ship collisions by broadcasting vessel position, course, and activity to surrounding vessels. These broadcast messages can be received by satellites and land-based antennas. GFW then uses machine learning techniques (convolutional neural networks) on the broadcast messages to infer what type of fishing is taking place and where it is taking place, thus allowing the estimation of near real-time fishing events globally (Kroodsma et al., 2018).

The amount of data gathered by GFW is dependent on the number of antennas and satellites that can receive signals. The total satellite count increased from 3 to 6 on June 1st 2014 , and then from 6 to 10 on January 1st 2016. This causes an increase in the number of *received* AIS messages (*i.e.* points), and therefore an apparent increase in the number of vessels and fishing hours. However, the addition of new satellites affects all vessels in the same way.

Our analysis focuses on purse seine vessels, the most important fishery for PNA countries.⁵ We identify a total of 103 purse seiners that fished in PNA waters at least once before 2015⁶. These vessels represent over 26 million individual observations for the 2012 - 2017 period. We identify 65 vessels that have fished inside PIPA at least once since 2012. From these, 62 did so before the announcement (*i.e.* 09/01/2014 *sensu* (McDermott et al., 2018)) but we are only able to track

⁴Global Fishing Watch: globalfishingwatch.org

⁵We perform some of the same analyses for longliners and include them in the Appendix

⁶New vessels that entered the fishery after 2015 were not exposed to the policy intervention in the pre-treatment period and are therefore excluded from our analyses.

¹⁸⁶ 61 for the complete period of study before and after its implementation.⁷ On the other hand, 38
¹⁸⁷ vessels never fished inside PIPA but we only have data for 26 of these vessels both before and after
¹⁸⁸ MPA implementation.

¹⁸⁹ Therefore, our treatment group contains all purse seiners ($n = 61$) that fished within PIPA at
¹⁹⁰ least once before the announcement, and that continued to fish elsewhere after the January 2015
¹⁹¹ implementation. Vessels in the control group meet the following two conditions: i) never fished
¹⁹² within PIPA waters from 2012-2015, and ii) vessels have fished in surrounding areas (*i.e.* PNA-
¹⁹³ countries' EEZ) before and after PIPA closure ($n = 26$).

¹⁹⁴ We include three additional control groups as a robustness check. The first group contains only
¹⁹⁵ vessels that belong to PNA countries ($n = 7$). The second group excludes Chinese vessels ($n =$
¹⁹⁶ 21). Our third control is made up of Japanese purse seiners that fish in the Pacific but have never
¹⁹⁷ fished inside PIPA ($n = 27$).⁸ Our main definition of treatment and control groups leaves us with
¹⁹⁸ 61 treated and 26 control vessels, which have just over 22 million observations with about 22% of
¹⁹⁹ these observations identified as fishing events by the neural network classification.

²⁰⁰ For each vessel, we calculate total daily fishing hours and build a daily panel with 37,800 observa-
²⁰¹ tions. Table 1 shows the number of vessels following a Before-After-Control-Impact (BACI) design,
²⁰² as well as the fishing hours, before and after PIPA. Fig. S2 shows that mean fishing hours for purse
²⁰³ seiners have an abrupt increase, just before January 1st, 2015. This trend is observed for both
²⁰⁴ treated and control groups. Across all measures, the treatment and control vessels follow similar
²⁰⁵ patterns, confirming our claim that the control group provides a plausible counterfactual. Figure
²⁰⁶ S1 in the Appendix provides a visual representation of the vessel-level fishing events that make up
²⁰⁷ each group through time. We use this panel data to answer our key research questions: Are the
²⁰⁸ 61 purse seiners that used to fish within PIPA affected by the closure? If so, how? What are they
²⁰⁹ doing differently and where are they going that they didn't go to before? The following section
²¹⁰ describes our empirical identification strategy.

Table 1: Number of fishing vessels and mean daily fishing hours by group before and after PIPA implementation.

Group	n	Before	After	Change (A / B)
Control	26	11.74	20.13	1.71
Treatment	61	10.67	18.36	1.72

²¹¹ 3.2 Analyses

²¹² Our first analysis focuses on identifying the response of vessels to the PIPA closure. We use
²¹³ daily fishing hours, daily proportion of fishing vs. non-fishing hours, daily distance traveled (km),
²¹⁴ distance from shore (km) and distance from home port(km) as our main outcomes of interest.
²¹⁵ We compare these outcomes before and after the implementation of PIPA using a Difference-in-
²¹⁶ Differences approach. Our main specification is the following:

$$y_{i,t} = \alpha + \beta_1 Post_t + \beta_2 Treat_i + \beta_3 Post_t \times Treat_i + \phi_t + \gamma_i + \theta_t + \epsilon_{i,t},$$

⁷1 vessel never fished again after August 18, 2013.

⁸We also considered a control group of Taiwanese flagged vessels but there are only 4 vessels with pre-2015 data.

217 where $y_{i,t}$ is the outcome of interest for vessel i on day t . A dummy variable $Post_t$ takes the
 218 value of 0 for all dates prior to PIPA implementation and a value of 1 for all dates following PIPA
 219 implementation. $Treat_i$ is a dummy variable indicating whether a vessel belongs to the treatment
 220 ($Treat_i = 1$) or control ($Treat_i = 0$) group. α is the standard intercept term, β_1 captures the
 221 temporal trend, β_2 captures the initial difference between treated and control groups, and β_3 is
 222 our parameter of interest: the Difference-in-Differences estimate capturing the treatment effect.
 223 Finally, ϕ_t and γ_i represent month and flag dummies that account for seasonality or country-level
 224 management interventions. θ_t are time dummies to account for the addition of more satellites over
 225 time.⁹

226 Our second part of the analysis focuses on the redistribution of fishing effort. The role of institutions
 227 may play an important role. As stated before, LSMPAs often span the entirety of an EEZ. In the
 228 case of purse seining in the PNA, vessels purchase access to the fishery at the beginning of the season
 229 (QUESTION FOR JC: WHEN IS THIS? And wasn't the closure expected?). If a vessel decides to
 230 fish within the PNA, they've already made the decision to pay for access, expecting higher returns
 231 from fishing there than in the high seas. In the particular case of PIPA, one would expect that a
 232 vessel previously holding a permit to fish in Kiribati waters would most likely continue to purchase
 233 access to PNA waters (either in Kiribati's remaining open areas or a nearby EEZ). Alternatively,
 234 if a vessel fishes illegally in Kiribati waters before the implementation, one would expect them to
 235 reallocate to other regions with equal or less enforcement. In this case, they might then choose to
 236 move to other Kiribati waters, waters of countries that have similar enforcement levels, or to the
 237 high seas.

238 We discretize spatial units by creating a polygon for PIPA and distinct spatial units for each
 239 geographically separate EEZ of each country. Some vessels might shift from EEZs into the high
 240 seas, but we are interested in knowing *where* in the high seas, so we incorporate additional regions
 241 by using a 1 degree buffer of the high seas around each of the EEZ regions. The rest of the high
 242 seas are merged into a single spatial unit. For example, if we were to do this only for Kiribati, we
 243 would have 8 spatial units: PIPA, three EEZ units, three 1-degree buffers of high seas around each
 244 EEZ, and the rest of the high seas. Whenever the buffer units overlap, we randomly clip one on to
 245 the other (QUESTION FOR JC: WHAT DOES THIS MEAN?).

246 We then take these spatial polygons and rasterize them to a 1-degree grid (see Figure S11 in the
 247 Appendix), which we then use to rasterize points of fishing activity. For each cell, we calculate the
 248 total monthly fishing hours by treated and control vessels (see Figure 5). Since we are interested
 249 in the redistribution of fishing effort, we use this gridded data to calculate the proportion of fishing
 250 hours that each region receives each month relative to the total fishing hours observed across all
 251 regions and all cells in that month.

252 We use this measure to evaluate the change in effort allocation by regressing it on the interaction
 253 between a year dummy and a dummy for regions:

$$y_{i,t} = \alpha + \beta_1 Year_t + \beta_{2,i} Region_i + \beta_{3,i} Year_t \times Region_i + \epsilon_{i,t}$$

254 Our variable of interest, $y_{i,t}$ represents the proportion of fishing hours that region i receives at time
 255 t . Years are modeled as factors ($Year_t$), using 2012 as the reference level. $Region$ is a dummy
 256 variable for regions defined above. Our parameter (QUESTION FOR JC: VECTOR?) of interest is

⁹We test for other more complex specifications that interact a quarterly dummy or year-month dummies with the treatment group and find qualitatively the same results.

257 $\beta_{3,i}$, which captures the yearly by-country change in proportional allocation of fishing effort relative
258 to 2012.

259 All regression coefficients were estimated via ordinary least squares, and heteroskedasticity-robust
260 standard errors were calculated. All analyses were performed in R version 3.5.1 (R Core Team,
261 2018). Raw data and code used in this work are available on github.

262 4 Results

263 Regression coefficients for our DiD analysis are shown in Table 2. Columns 1 -3 use our primary
264 control group. Columns 4 - 6 use only PNA vessels as controls, and columns 7 - 9 exclude Chinese
265 vessels. Columns 10 and 11 use Japanese vessels in the Pacific as controls, and don't include flag
266 fixed-effects to avoid perfect collinearity between the Japanese flag and control group status. Our
267 DiD analysis shows an overall increase in purse seine fishing hours, even after accounting for the
268 introduction of new satellites (Table 2).¹⁰ This *post* coefficient estimate is consistent across different
269 model specifications and across controls, and effectively represents the patterns observed in Figure
270 S2.

271 The β_3 coefficient indicating our treatment effect suggests that, relative to the control, treated
272 vessels fish less: to the order of 0.5-3.7 hours per day. In 6 out of 11 specifications, there is a
273 statistically significant decrease in daily fishing hours. Another way to interpret this is that the
274 increase in fishing effort by treated vessels has occurred at a lower rate than control vessels. These
275 values are equivalent to a reduction of 15 - 111 fishing hours per month. However, this result is only
276 robust and significant ($p < 0.01$) in the simplest specification of the Diff-in-Diff. When reducing
277 the linear structure of the *Pre* \times *Post* design and instead interacting the treatment dummy with
278 quarterly or year-month combinations, we are not able to reject the null hypothesis of no change,
279 suggesting that the total fishing hours remains similar for the displaced vessels (Figs. S3-S10).
280 Since vessels don't seem to be fishing a lot less, an obvious question arises: where are they fishing
281 now?

282 Figure 5 provides a spatial representation of these changes. Note that the 2013 effort distribution
283 is roughly similar across vessels. Then, in 2014, there is a sharp increase in fishing hours by treated
284 vessels inside PIPA (the so-called Blue Paradox). In 2015, treated vessels then allocate more effort
285 to the easternmost Kiribati EEZ and the high seas. In 2016, the spatial distribution of fishing effort
286 is again similar across groups. It is evident that the increase in relative fishing effort is greater for
287 regions closer to PIPA.

288 Along a similar line, Figure 6 shows a detailed temporal evolution of the relative allocation of fishing
289 effort.¹¹ It shows the change in fishing effort inside PIPA, including the preemptive fishing and
290 immediate reduction previously reported (McDermott et al., 2018). Kiribati waters in KIR 1 and
291 KIR 2 show an increase in fishing effort for 2015; the same pattern is observed for their respective
292 High Seas buffers, and the general High Seas. Other regions farther away from PIPA don't show
293 a clear pattern under visual inspection. This suggests fishing effort inside PIPA has moved to the
294 high seas around Kiribati instead of paying to fish within alternative Kiribati EEZ waters.

295 We then move on to our regression results, presented as a figure showing the year-region interaction
296 coefficient estimates for each region as a stream of changes through time (Figure 7 and Table 3).
297 As noted before, the marginal change in the relative allocation of fishing effort relative to 2012
298 increases in 2014 for PIPA. Most coefficients are not statistically significant, but follow the same
299 trends previously discussed, with a tendency for regions closer to PIPA to show an increase in
300 the relative allocation of fishing effort in the post-implementation period. The largest increase is
301 observed for the Kiribati EEZ 2, where the redistribution of treated vessels caused a 15% increase
302 in the *relative* allocation of fishing effort within Kiribati waters. This displacement is likely causing

¹⁰Results of the same analysis is shown for longliners in Table S1

¹¹Recall that to evaluate the redistribution of fishing effort we only track fishing vessels that belong to the treated group.

303 a crowding effect after PIPA-fishing vessels redistribute to other areas.
 304 We use the rasterized effort to evaluate if the PIPA closure increased crowding outside the LSMPA.
 305 We count the number of raster cells that had both treated and control vessels each year (Fig. 8A).
 306 The pattern observed here could just be an artifact of satellite detections increasing though time.
 307 Therefore we also calculate a spatial correlation of presence/absence and observe similar patterns
 308 (Fig. 8B). While this is at the year level and cannot assure that vessels were in effect at the same
 309 time in the same place, it provides a of what we would expect to find if our claims of displacement
 310 are true. Spatial overlap increases in 2015 as vessels from PIPA are excluded, but then starts to
 311 decline. [QUESTIONS FOR JC: Is this significant? Can we do a daily measure? What about
 312 mean or shortest distance between vessels. Treated/Control/Combined]

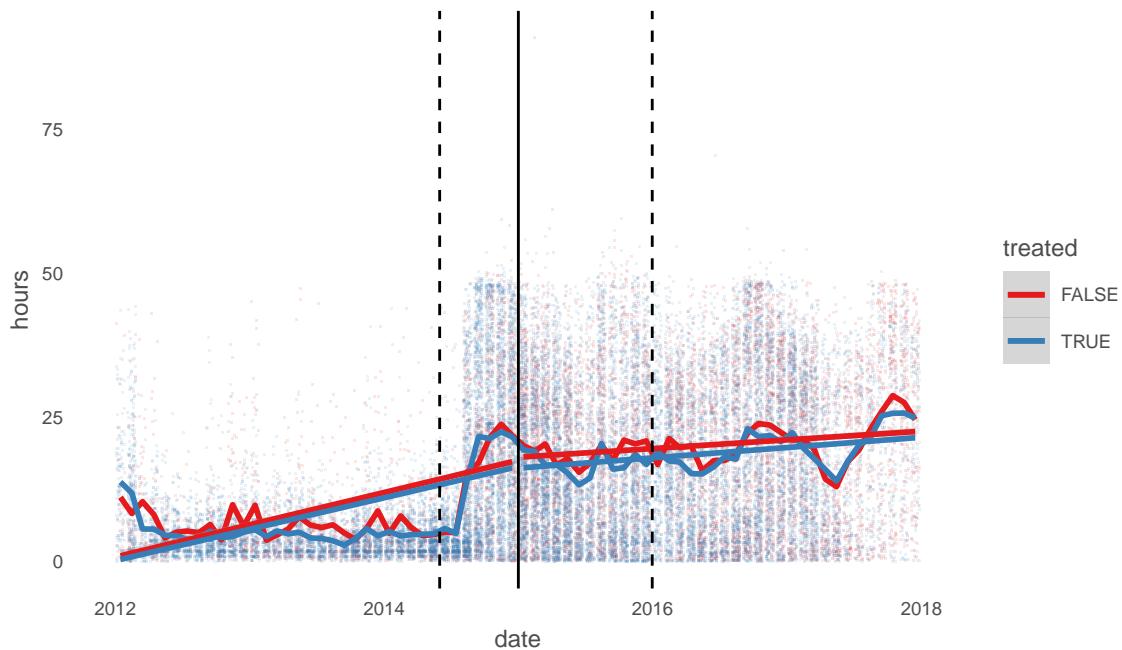


Figure 3: Daily fishing hours for all vessels in our main treatment-control groups. Solid straight lines show a linear trend by period (pre-post) and treatment. The other red and blue lines show monthly averages. Vertical dashed lines indicate dates when satellites were added, solid line indicates PIPA closure.

Table 2: Difference-in-differences estimates for 3 different controls and 3 different specifications. The first three columns use our primary control group. Columns 4 - 6 use only PNA vessels as controls, and columns 7 - 9 exclude Chinese vessels. Columns 10 and 11 use Japanese vessels in the Pacific as controls, and hence don't include flag FEs. Numbers in parentheses are heteroskedasticity-robust standard errors.

	Dependent variable: hours										
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Constant	6.345*** (0.186)	7.863*** (0.274)	10.322*** (0.404)	7.124*** (0.265)	9.007*** (0.341)	9.933*** (0.620)	6.057*** (0.193)	7.584*** (0.279)	10.194*** (0.538)	22.625*** (0.629)	24.731*** (0.683)
post	-0.052 (0.303)	1.134*** (0.304)	1.115*** (0.339)	-1.199*** (0.435)	-0.279 (0.429)	0.124 (0.436)	-0.284 (0.318)	1.013*** (0.318)	1.166*** (0.357)	2.330*** (0.890)	3.991*** (0.877)
treated	-1.142*** (0.214)	-0.839*** (0.209)	0.010 (0.281)	-1.977*** (0.284)	-1.675*** (0.274)	0.626 (0.416)	-0.877*** (0.221)	-0.536** (0.215)	0.150 (0.297)	-18.190*** (0.673)	-18.257*** (0.661)
sate2	12.579*** (0.197)	11.589*** (0.199)	11.340*** (0.232)	12.709*** (0.212)	11.850*** (0.213)	11.749*** (0.246)	12.631*** (0.199)	11.599*** (0.201)	11.349*** (0.235)	14.346*** (0.303)	12.967*** (0.309)
sate3	14.675*** (0.260)	13.587*** (0.262)	13.328*** (0.307)	14.799*** (0.286)	13.894*** (0.287)	13.795*** (0.329)	14.958*** (0.264)	13.804*** (0.266)	13.566*** (0.315)	15.187*** (0.402)	13.719*** (0.410)
post:treated	-0.515* (0.281)	-0.833*** (0.276)	-0.934*** (0.310)	0.562 (0.413)	0.373 (0.402)	-0.176 (0.407)	-0.439 (0.293)	-0.814*** (0.287)	-1.084*** (0.326)	-3.209*** (0.811)	-3.715*** (0.799)
Control	All	All	All	PNA	PNA	PNA	-CHN	-CHN	-CHN	JPN	JPN
Month FE	No	Yes	Yes	No	Yes	Yes	No	Yes	Yes	No	Yes
Flag FE	No	No	Yes	No	No	Yes	No	No	Yes	No	No
Observations	37,840	37,840	30,359	30,583	30,583	25,034	36,415	36,415	28,934	34,047	34,047
R ²	0.171	0.200	0.208	0.178	0.208	0.215	0.173	0.203	0.211	0.260	0.280

Note:

*p<0.1; **p<0.05; ***p<0.01

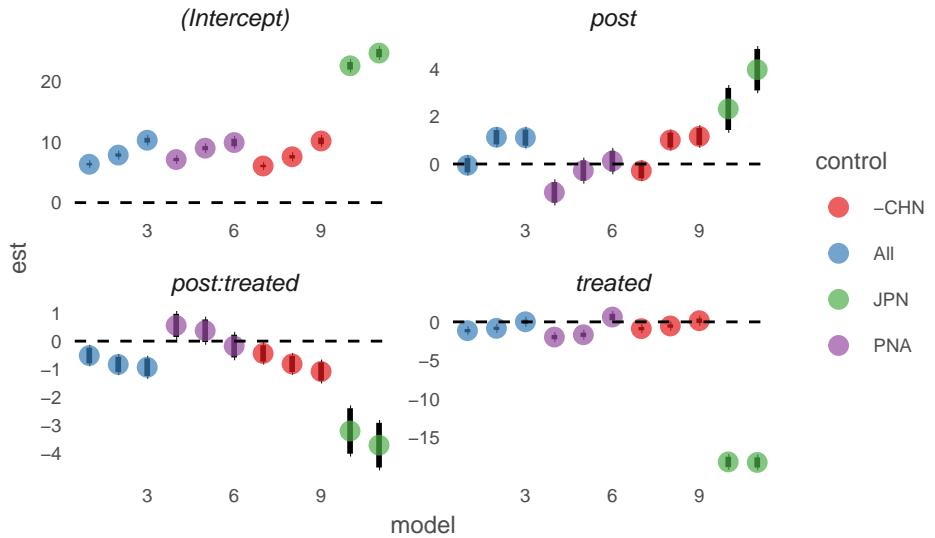


Figure 4: Coefficient estimates for each model. Top panel indicates variable, x-axis represents model specification, and y-axis coefficient estimate. [QUESTION FOR JC: delete the intercept panel.]

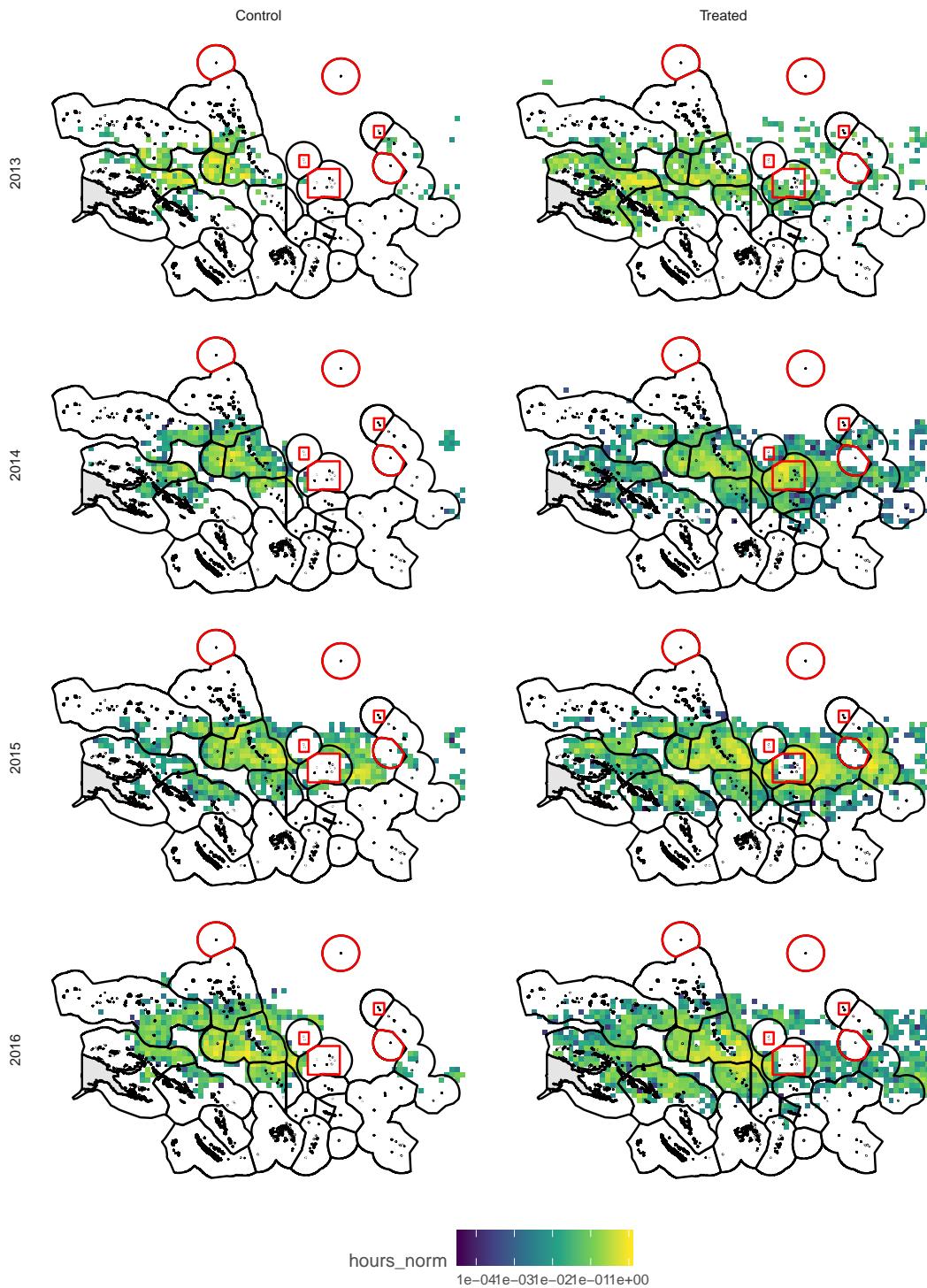


Figure 5: Yearly spatial distribution of fishing effort by treated and control vessels. Colors have been adjusted relative to the maximum observed by group and year. Red polygons show LGMPAs in the region. For visualization purposes only some years are presented, the full 2012 - 2017 period is shown in Figure S12. [QUESTION FOR JC: we should include all years in the main text. Is there a way to normalize this, e.g. color corresponds to % of total fishing effort, so it's easier to compare?]

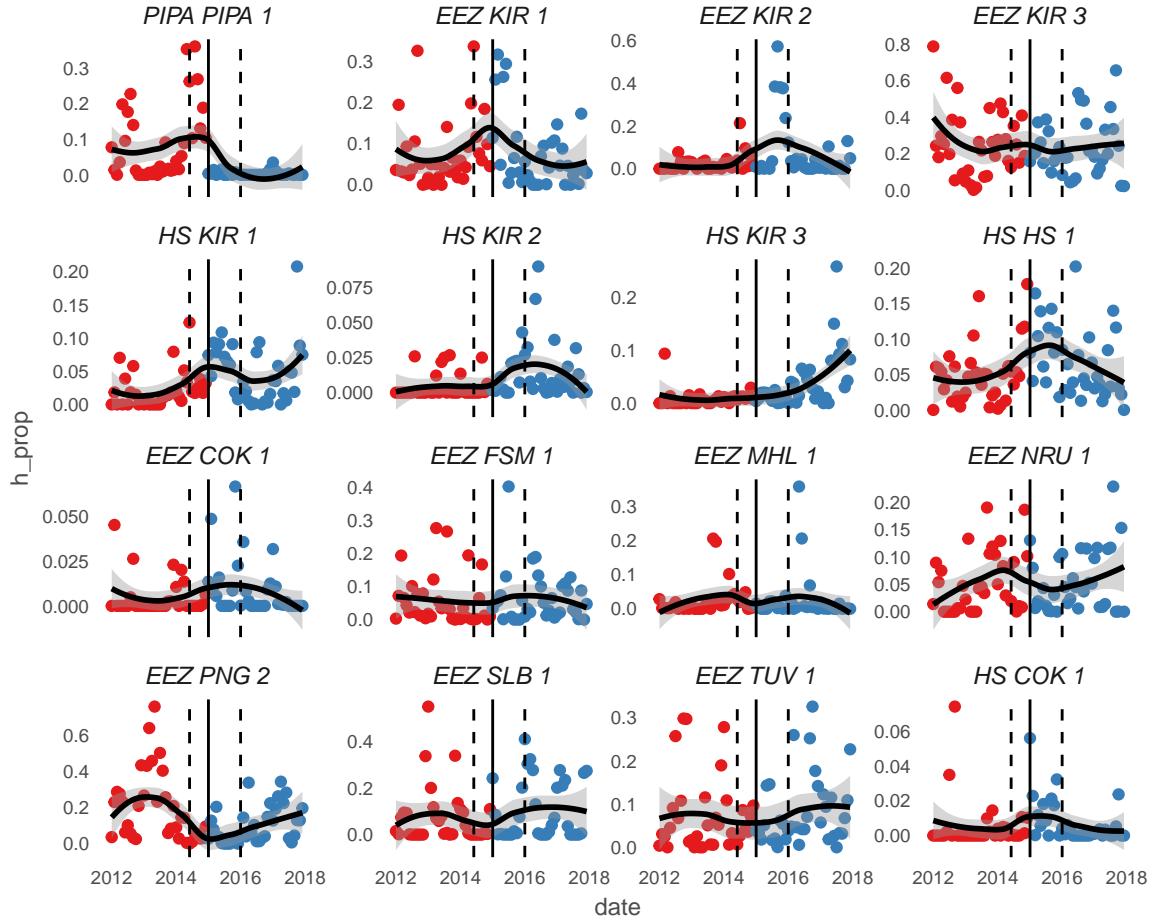


Figure 6: Monthly relative allocation of fishing effort by PIPA-fishing vessels before and after PIPA by region. Colors indicate the pre- and post- periods. Solid line across points shows local mean with a region-specific loess smoother. Vertical dashed lines indicates dates when satellites were added, solid vertical line indicates PIPA closure. Note that inflection points of the loess smoother are not affected by the addition of satellites. [QUESTION FOR JC: this figure needs a lot of work. subplot titles need to be more intuitive. I would combine all EEZ KIR into one group and all HS KIR into one group.]

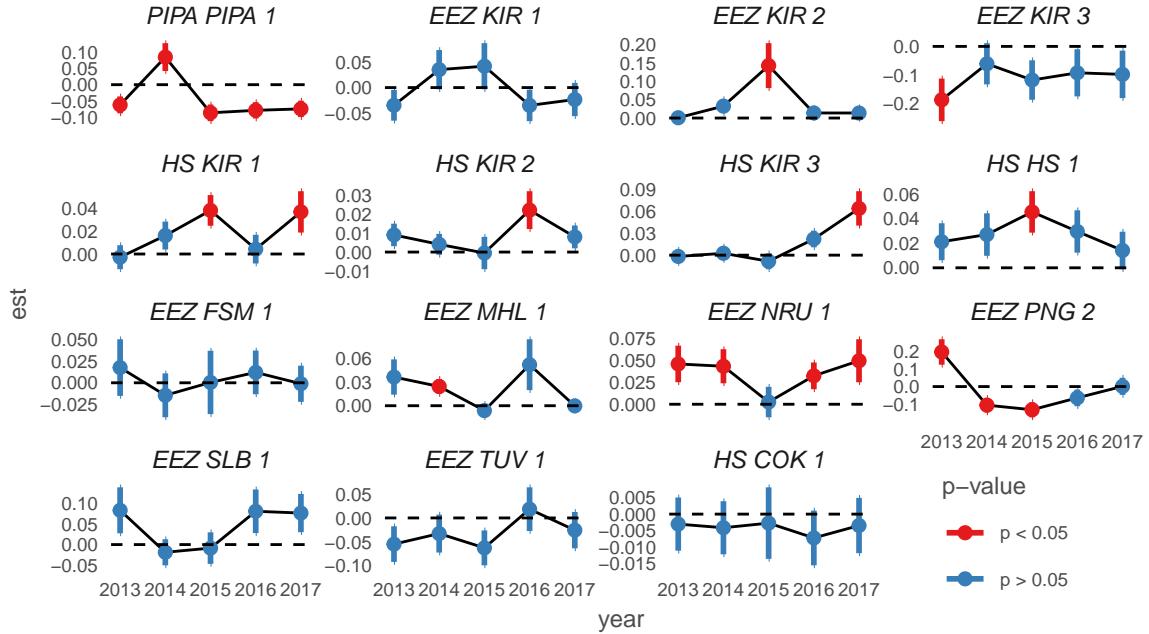


Figure 7: Coefficient estimates for the redistribution regression. Each panel shows the region-specific coefficients, with heteroskedasticity-robust standard errors as error bars. The horizontal dashed line represents 0 change relative to the 2012 region-specific levels. [QUESTION FOR JC: just do 95% confidence intervals.]

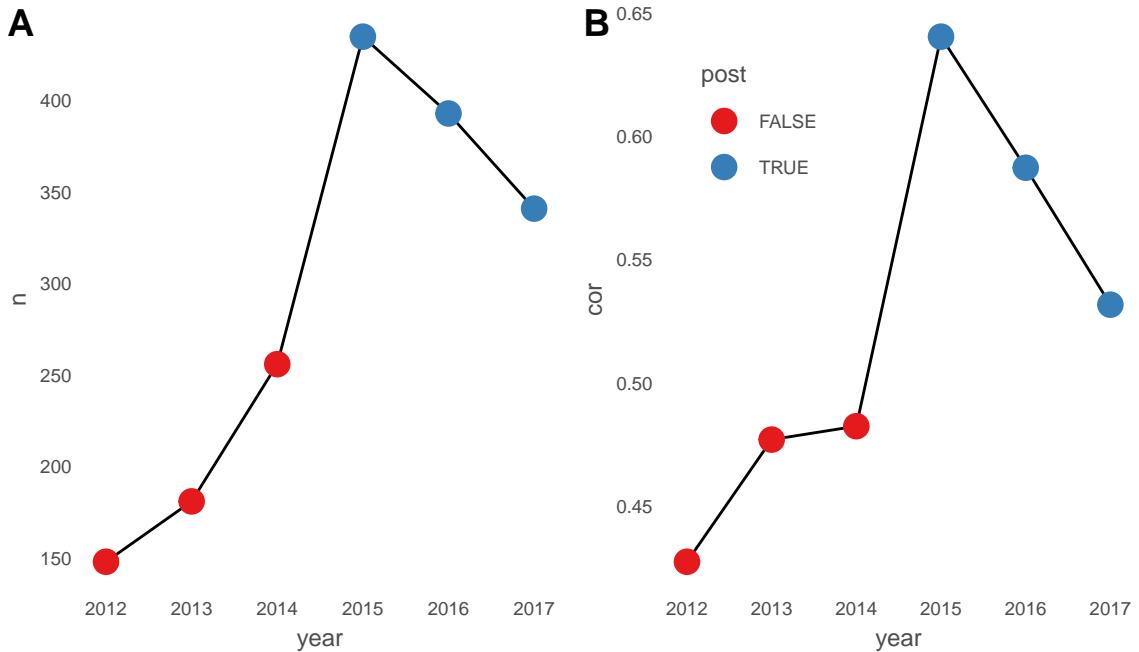


Figure 8: Number of cells that had treated and control vessels (A) and spatial correlation in the presence-absence of treated and control vessels per cell (B).

Table 3: Coefficient estimates for the interaction of year and region. Each row represents a region, each column a year. Numbers in parentheses are heteroskedasticity-robust standard errors. R2 = 0.47 (F(95, 1056) = 9.99; p < 0.001). * < 0.1; ** p < 0.05; ***p < 0.01.

term	2013	2014	2015	2016	2017
PIPA PIPA 1	-0.062 (0.026)**	0.085 (0.043)**	-0.086 (0.026)***	-0.079 (0.025)***	-0.074 (0.025)***
EEZ KIR 1	-0.034 (0.030)	0.035 (0.038)	0.042 (0.045)	-0.035 (0.031)	-0.023 (0.032)
EEZ KIR 2	0.001 (0.009)	0.032 (0.019)*	0.142 (0.061)**	0.014 (0.013)	0.014 (0.015)
EEZ KIR 3	-0.188 (0.075)**	-0.061 (0.072)	-0.118 (0.069)*	-0.093 (0.083)	-0.098 (0.083)
HS KIR 1	-0.003 (0.011)	0.016 (0.012)	0.038 (0.013)***	0.004 (0.013)	0.037 (0.018)**
HS KIR 2	0.009 (0.006)	0.004 (0.006)	-0.000 (0.009)	0.022 (0.010)**	0.008 (0.006)
HS KIR 3	-0.002 (0.009)	0.003 (0.009)	-0.008 (0.011)	0.022 (0.011)*	0.064 (0.024)***
HS HS 1	0.021 (0.015)	0.027 (0.017)	0.045 (0.017)***	0.030 (0.017)*	0.014 (0.015)
EEZ FSM 1	0.018 (0.033)	-0.015 (0.025)	0.000 (0.037)	0.012 (0.025)	-0.001 (0.021)
EEZ MHL 1	0.037 (0.022)	0.025 (0.010)**	-0.006 (0.009)	0.052 (0.032)	-0.000 (0.006)
EEZ NRU 1	0.046 (0.021)**	0.044 (0.020)**	0.003 (0.017)	0.033 (0.015)**	0.050 (0.025)**
EEZ PNG 2	0.196 (0.072)***	-0.106 (0.041)***	-0.132 (0.043)***	-0.064 (0.048)	0.001 (0.049)
EEZ SLB 1	0.083 (0.056)	-0.019 (0.032)	-0.009 (0.038)	0.081 (0.053)	0.076 (0.046)*
EEZ TUV 1	-0.055 (0.037)	-0.033 (0.040)	-0.063 (0.037)*	0.018 (0.046)	-0.026 (0.038)
HS COK 1	-0.003 (0.008)	-0.004 (0.008)	-0.003 (0.011)	-0.007 (0.008)	-0.003 (0.008)

313 5 Palau National Marine Sanctuary

314 On October 28, 2015 the President of Palau, H.E. Tommy E. Remengesau Jr., signed into law the
315 Palau National Marine Sanctuary Act. Starting in December 2020, this will close 193,000 square
316 miles to commercial fishing activities, creating the 14th largest protected area in the world.

317 5.1 Background on Commercial Fishing in Palau

318 Foreign tuna fishing in Palau's EEZ began before WWI. Pole and line was the initial gear used to
319 target skipjack tuna by the Japanese. Foreign fishing activities stopped during WWII and started
320 again in the 1960s when the Japanese returned and the locally based Van Camp Seafood Company
321 carried out fishing with boats crewed by Okinawans. The Japanese also began purse seine fishing
322 in the 1960s and 1970s and limited longline fishing for yellowfin tuna during the same time. During
323 the 1980s, longline fishing began to set their lines deeper, shifting their targets to bigeye tuna.
324 During the 1980s Korean and Taiwanese vessels also began longline fishing in Palauan waters for
325 export to Japan (?). There are currently three transshipment companies operating in Palau.

326 5.1.1 Purse seine fishing

327 Purse seine vessels target schools of skipjack tuna that are used for canning. Compared to other
328 EEZs in the Pacific, Palau's EEZ is not known to be preferred for purse seining (Sisior, pers.
329 comm.). Currently all purse seine boats fishing in Palau's EEZ are Japanese and do not land their
330 catches in Palau Table ??.

331 5.1.2 Longline fishing

332 All longline vessels fishing in Palau's EEZ are foreign owned (Table 2). Japanese longline vessels do
333 not land their catches in Palau. Locally based, foreign owned longline boats are mostly Taiwanese
334 vessels that are contracted by one of the two main fishing companies (Palau International Traders
335 Incorporated (PITI), Kuniyoshi Fishing Company (KFC)) (Table 2). Ninety-five percent of the
336 catch from PITI and KFC contracted boats is exported to Japan, upon landing in Koror. Most of
337 the discards, or portion of the catch that is of too low quality to export, is either sold or donated in
338 Palau. A very small portion of the discards is frozen and transported to Taiwan when the vessels
339 return to their home ports.

340 5.1.3 Access fees

341 An agreement between Palau and four Japanese fishing associations¹² covers three methods of
342 fishing (longline, purse seine, pole and line) and has allowed Japan to have up to 290 vessels in
343 Palau waters with no limits on catch. Japan pays 4-5% of catch returns, but there was no way to
344 validate their catch (?). Between 2010 and 2014, Japan has paid Palau between \$196,100- \$867,120

¹²This is a single agreement between Palau and four associations—1) Federation of Japan Tuna Fisheries Cooperative Associations, 2) National Offshore Tuna Fisheries Association of Japan, 3) Japan Far Seas Purse Seine Fishing Association, 4) Federation of North Pacific District Purse Seine Fisheries Cooperative Association of Japan (Performance Audit Report on Managing Sustainable Fisheries (Tuna) 2013).

345 annually in access fees (?). Currently, these access fees have been replaced by the vessel day schemes
346 (described in section 3).

347 **5.1.4 Uniform Longline Agreement for locally based vessels (ULA)**

348 Agreements between the Republic of Palau and foreign owned, local based companies (PITI and
349 KFC) (?). Vessels under these agreements are internationally owned, locally based vessels that
350 have payed for annual licenses per vessel based on the size of the vessel. These are the vessels
351 in Table ?? (excluding the Japanese vessels). So, for example, 38 vessels in 2016. Between 2010
352 and 2014, total licensing fee revenue was between \$219,000-\$284,600 per year (?). Currently, these
353 access fees have been replaced by the vessel day scheme (described in section 3).

354 **5.1.5 US Multilateral Tuna Treaty and Federated States of Micronesia Arrangement**

355 These two multilateral treaties grant preferential access to the US and FSM flagged purse seine
356 boats, respectively. The US Treaty was renegotiated in 2016 with stipulations on minimum vessel
357 day fees (more in section 3). Very little purse seine fishing is conducted by FSM and US vessels in
358 Palau's EEZ, but Palau still receives a portion of these treaty funds. Some Pacific Island countries
359 treat this money as foreign aid, not for fishing access (?).

360 **5.1.6 Other sources of revenue from pelagic fisheries**

361 The export tax for all commercial tuna and billfish, either fresh or frozen is \$0.35/kg. Average
362 revenues from export taxes was approximately \$500,000 annually from 2011-2014 (?). Otherwise
363 there is little additional revenue from pelagic fisheries, especially considering that very few jobs are
364 created by the fishery. Between 80-90 people are employed by offshore fisheries, of these only 20%
365 are Palauans and average wages are 50% average wages in the country (?).

366 **5.1.7 Parties to the Nauru Agreement and the Vessel Day Scheme**

367 The Parties to the Nauru Agreement (PNA) is an agreement between Federated States of Micronesia,
368 Kiribati, Marshall Islands, Nauru, Palau, Papua New Guinea, Solomon Islands and Tuvalu
369 to manage the largest tuna fishery in the world. The Purse Seine Vessel Day Scheme (PS VDS),
370 established in 2006 with the Palau Agreement, is the current method used by the PNA to manage
371 effort in the fishery. Although not a PNA member, Tokelau joined the PS VDS in 2014, with 1,000
372 transferrable days (?).

373 In the PS VDS, a total number of annual vessel days for the entire fishery is agreed upon by
374 the parties. Each party's allowable effort (PAE) is then calculated based on historic effort and
375 biomass within each party's EEZ. Sixty percent of the PAE is calculated based on the party's effort
376 over the last seven years and 40% of the PAE is calculated based on the 10 year average of the
377 party's share of estimated skipjack and yellowfin biomass within its EEZ (as explained in Article
378 12.5 of the 2012 Amendment to the Palau Agreement and in Hagrannsoknir sf 2014). A minimum
379 benchmark fee is set for purse seine vessel days (PS VD) which each party can transfer (i.e., sell
380 to another PNA member) or sell to the highest bidder (i.e., sell to fishing company). Vessel days
381 can be transferred to fish in any PNA party's EEZ, without penalty to the transferred parties

382 VD allocation (?). The process of transferring days between parties is described in Article 7 of
383 the Management Scheme (?). The mean value of a vessel day has steadily increased since 2007
384 (Havice, 2013). A minimum benchmark fee of \$8,000/day was set from 2015 (?). Detailed records
385 on sales or transfers of vessel days are not publicly available (Havice, 2013; ?). ? summarize the
386 PNA ‘implementing arrangements’ as follows: “foreign vessels [are required] to be registered and
387 licensed, report catches, maintain log books, allow observers on board and maintain transparency
388 over their fishing activities.” Further, vessels registered with the Vessel Day Scheme are required to
389 have Automatic Location Communicators (ALC) or Mobile Transceiver Units (MTU) that transmit
390 their locations at least once per hour while they are within the VDS Management Area (?). Every
391 entire day that a vessel is within the VDS Management Area is counted as a vessel day used, unless
392 the vessel reports a “no fishing day” (e.g., travel, maintenance, etc.) and periods of less than 24
393 hours are counted as partial days (?). As described in the Palau Arrangement (?), fishing days
394 by vessels shorter than 50m or longer than 80m count as 0.5 and 1.5 vessel days, respectively.
395 Parties are responsible for ensuring registered vessels comply with the implementing arrangements
396 and that they stay within the PAE. Parties that exceed their PAE should reportedly be penalized
397 by reductions to the following year’s PAE (?). However, there are criticisms of the PNA PS
398 VDS claiming that there is a lack of monitoring, compliance, and transparency that could hinder
399 the future success of the SP VDS (??). In the case of Palau, the VDS is run by the Bureau of
400 Marine Resources, within the Ministry of Natural Resources, Environment and Tourism. Palau’s
401 currently PS VD allotment is reportedly 700 vessel days annually (Sisior, pers. comm., ?), though
402 historically, this number has been lower (average of 580 days between 2008-2011 (?). Palau sells
403 most of its vessel days to the US for \$12,500 each (as negotiated in the US treaty), regardless
404 of whether US vessels actually fish in Palau’s EEZ. The remainder of Palau’s PS vessel days are
405 sold for between \$9-10,000 each. When vessel days are purchased to transfer to another EEZ,
406 Palau charges a transfer fee (approximately \$500), for administration costs and future opportunity
407 costs (since vessel day allotments are based partly on historical effort within the EEZ). Though
408 records are confidential, it has been publicly reported that one PNA member party that Palau has
409 transferred PS VD to is Papua New Guinea (?) and this year, PS VD have been sold to a fishing
410 company in the Philippines for the first time (?). This year, Palau has brought in approximately
411 \$9 million USD in PS VDS revenue (Sisior pers. comm.).

412 The longline VDS (LL VDS) is in its infancy and has not yet been fully implemented at the PNA-
413 scale, although several countries are now implementing at the country-scale. Palau was the first
414 party to implement the VDS for the longline fishery in 2017. In 2014, longline boats fishing in
415 Palau’s EEZ fished 10,500 days (Sisior, pers. comm.). Since 2016, the number of allowed longline
416 VDs has been a fraction of the 2014 days, reducing each year as stipulated in the PNMS Act. Palau
417 sells its LL VD for \$150-\$250 (Sisior pers. comm.). Because the LL VDS is still new and has yet to
418 be fully implemented by all parties, LL VD are not currently transferable between PNA member
419 EEZs.

420 5.1.8 The Palau National Marine Sanctuary Act

421 In 2015, Palau enacted a policy to protect 500 thousand square kilometers of its ocean, representing
422 80% of its exclusive economic zone (EEZ), by 2020. The policy’s goal is to preserve and manage the
423 stocks, health, and beauty of Palau’s waters and natural resources by limiting fishing in its deep
424 ocean waters. State waters hugging the coasts of the archipelago are not affected by the policy.
425 Commercial fishing will still be allowed within the open 20% that surrounds the main islands of

⁴²⁶ Koror and Babeldaob, but all catch must be landed in Palau and exports of pelagic fish will be
⁴²⁷ prohibited. Alongside these protections, the Palau National Marine Sanctuary (PNMS) Act will
⁴²⁸ support a domestic pelagic fishery.

⁴²⁹ 5.1.9 Future of pelagic fisheries under the PNMS

⁴³⁰ To prepare for full enactment of the PNMS, the act stipulates a “winding down” period, in which
⁴³¹ baseline vessel days (i.e., the number of vessel days used in 2014) were reduced by 20% in 2016; and
⁴³² an additional 10% from baseline in each subsequent year until full enactment in 2020. This appears
⁴³³ to be occurring for the LL VDS, but it is uncertain whether this is being followed for the PS VDS,
⁴³⁴ because PS VD can be transferred for use in other EEZs. There has been no official statement on
⁴³⁵ what will happen to Palau’s vessel days upon full implementation of PNMS. However, there is a
⁴³⁶ sense that Palau will be able to keep its allotment come 2020 (Hanich, pers. comm.). In the 2015
⁴³⁷ Micronesian Presidents Summit, a letter was drafted by heads of state, calling on PNA members
⁴³⁸ to be supportive of Palau as they moved forward with the PNMS Act (?). Further, other PNA
⁴³⁹ members have not been penalized for other protected area closures (e.g., Phoenix Islands Protected
⁴⁴⁰ Area in Kiribati) (Hanich, pers. comm.). Table 3 estimates the potential losses of revenue under
⁴⁴¹ full enactment of the PNMS under four scenarios. In Scenario 1, Palau is able to keep its current
⁴⁴² allotment of purse seine vessel days (700) to transfer to other parties at the current benchmark
⁴⁴³ price (\$8,000/day). Scenario 1 is likely if Palau retains its PAE, but the US no longer purchases
⁴⁴⁴ its days (note, US is currently purchasing days at \$12,500). Scenario 2 is the same as the first
⁴⁴⁵ scenario, except instead of being able to transfer its days at the current benchmark price, it is able
⁴⁴⁶ to transfer its days for \$12,000 (to account for current price to US of \$12,500, which will increase
⁴⁴⁷ in accordance with the US Treaty, and also the general increasing trend in the value of PS vessel
⁴⁴⁸ days). It should be noted that if PAE continues to be calculated based on effort and biomass, and
⁴⁴⁹ if Palau continues to be allocated vessel days, its PAE will decrease as effort in its EEZ reaches
⁴⁵⁰ zero. In Scenario 3 and 4, Palau loses all of its PS vessel days, at \$8,000/day and \$12,000/day,
⁴⁵¹ respectively. In all scenarios, all longline vessel day and export tax revenue are lost. Longline vessel
⁴⁵² day loss is calculated using an average value of \$200 for 10,500 days. Export tax loss is calculated
⁴⁵³ given the average tax revenue from 2012-2014 (\$482,236 from ?).

⁴⁵⁴ 5.1.10 Plans for a domestic pelagic fishery

⁴⁵⁵ In the recent past, a single domestic pole and line vessel supplied Palau with an estimated 100mt
⁴⁵⁶ of catch Plans for domestic fishery (?). The vessel ceased operations in recent years when the
⁴⁵⁷ captain became ill and passed away, but there is talk that it is gearing up to start operations again
⁴⁵⁸ under new leadership (Sisior, pers. comm.). Otherwise, there are no domestic vessels dedicated
⁴⁵⁹ to full-time offshore fishing. A number of fishers occasionally troll offshore for tuna and other
⁴⁶⁰ pelagics, but there are no official estimates of volumes (although Phil James with SPC has results
⁴⁶¹ that should come out soon). Local NGOs have sponsored short trainings on small-scale canning
⁴⁶² and proper techniques for processing high grade tuna to encourage existing fishers to devote more
⁴⁶³ time to offshore fishing. There are several possible scenarios for domestic fishery (including those
⁴⁶⁴ featured in a feasibility report by FFA (?)), but there has yet to be any official action towards a
⁴⁶⁵ particular scenario.

⁴⁶⁶ We will use aggregated GFW data at the 1km * 1km level to estimate what % of fishing effort
⁴⁶⁷ under the PNA is currently taking place in the proposed sanctuary? This will estimate the impact

⁴⁶⁸ on foreign-flagged fleets that used to fish in Palau waters. We are currently in possession of this
⁴⁶⁹ data. In terms of the impact on PNA payments, it is our understanding that the allocation will
⁴⁷⁰ not be reduced (in the near future) but perhaps the price received per VDS will be lower. But we
⁴⁷¹ do not have good information on prices paid for VDS. We also don't yet have good data on local
⁴⁷² fishers in Palau that are likely to be positively impacted since they will be allowed to fish in the
⁴⁷³ sanctuary.

⁴⁷⁴ **6 Conclusion**

Figure 9: Fishing Effort in Palau Relative to the Total PNA Area

Figure 10: Fishing Effort in Palau Relative to the Total PNA Area By Distant Water Fishing Fleet

475 **7 Discussion**

476 Our findings provide insights into the effect that LSMPAs can have on vessel behavior and the
477 redistribution of fishing effort. These results show that the implementation of PIPA had little effect
478 on the total fishing effort exerted by purse seiners. However, regions adjacent to PIPA show an
479 increase in the relative amount of fishing effort, suggesting that effort is simply displaced to nearby
480 areas. This does not imply that there is more fishing effort exerted by treated vessels, but rather
481 that each region receives a greater portion of the post-PIPA fishing effort of these same vessels,
482 which is lower than pre-PIPA levels. Additionally, we show that post-PIPA there is a concentration
483 of effort leading to crowding. In this section, we discuss the implications of vessel-level reductions
484 in fishing effort and the increase in relative allocation of the remaining effort through space. We
485 also provide plausible explanations as to why purse seiners seem to be more reactive to the spatial
486 closure.

487 Previous studies on protected areas around Pacific islands suggest that vessels move to distant
488 places, which might be translated as increased costs (Stevenson et al., 2013). Others have used
489 similar satellite-tracking systems to show that fishing effort accumulates near the edges of spatial
490 closures, yielding greater catches over time (Murawski et al., 2005). But these vessel tracks do not
491 cover the pre-reserve period, making it difficult to identify the contribution of spatial closures to
492 the observed spatial distribution of fishing vessels. Recent work by Elahi et al. (2018) identified
493 that total fishing effort in a focal region where a short-term MPA was implemented showed little
494 change, likely indicating that fishers redistributed fishing effort to compensate for the reduction in
495 available space. Our data, which is assembled in a similar way, allows us to make similar inferences
496 about the unobserved change in aggregate fishing effort and its spatial redistribution.

497 The role of the institutions at play is evident. Vessels that fished inside PIPA before the imple-
498 mentation were likely fishing in Kiribati waters under a VDS license. Upon implementation, some
499 vessels redistribute to other areas of the Kiribati EEZ and the adjacent high seas for the year im-
500 mediately after the closure. As stated before, it is reasonable for a vessel holding a license to fish in
501 Kiribati waters to simply reallocate their effort to other parts of that EEZ. However, the shift to the
502 high seas may not follow the same logic. Instead, it is possible that vessels that redistribute to the
503 high seas were operating in Kiribati illegally, and decide to reallocate in the high seas to continue
504 fishing without a license. Or they are now not willing to pay Kiribati to fish in the remaining open
505 areas.

506 A major shortcoming of our analysis is that we do not observe catches or revenues, which ultimately
507 are the factors that guide the decision making process of profit-maximizing agents. Therefore, it is
508 difficult to know whether the small change in fishing hours represents a positive or negative impact.
509 An additional factor that we are yet to test is the change in non-fishing hours spent at sea. It is
510 plausible that fishing hours remain constant, but vessels have to increase search time. Upon being
511 relocated, fishers may not identify the best fishing grounds as easily as before, and therefore invest
512 a greater proportion of their time searching for their catch. Further analysis of temporal trends
513 in non-fishing hours, as well as distance traveled should provide us with insights as to why fishers
514 reduced fishing hours. However, the widespread footprint of the treated fleet may suggest that they
515 are well acquainted with the region (see Figure 5), and it's unlikely that they would need to invest
516 much time to identify new fishing grounds.

517 Our work suggests that the implementation of LSMPAs has little impact on total fishing effort. We
518 also show that fishing effort is redistributed to areas close by, and that this leads to a potentially

negative crowding effect. A growing body of literature suggests that closing the high seas to all fishing could increase fishery yields and profitability of fisheries, with negligible costs to food security (White and Costello, 2014; Sumaila et al., 2015; Sala et al., 2018; Schiller et al., 2018). Such management interventions should consider how fishing effort will change in space and through time, and the ecological implications of this redistribution to ensure that fishing effort is not just displaced elsewhere, leading to overfishing in adjacent waters. While LSMPAs can provide a wide range of benefits, their implementation must be accompanied with traditional fisheries management to maximize effectiveness.

8 Further work

- Distance traveled
- Non-fishing hours at sea
- Monthly crowdnes (same approach as now, but monthly instead of yearly)
- Number of vessels per cell as another measure of crowdness (this is a more real one)
- Repeat raster exercise for Revillagigedo and other recently implemented LSMPAs. It would be great to have a pannel figure of 3 or 4 other LSMPAs that show wher effort was and where it went to generalize our results.
- John talked about this being relevant to Palau (their upcoming LSMPA). We might want to talk about where the vessels that are there would go.

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624 9 Appendix

625 9.1 Stream of data by vessel, group, and period

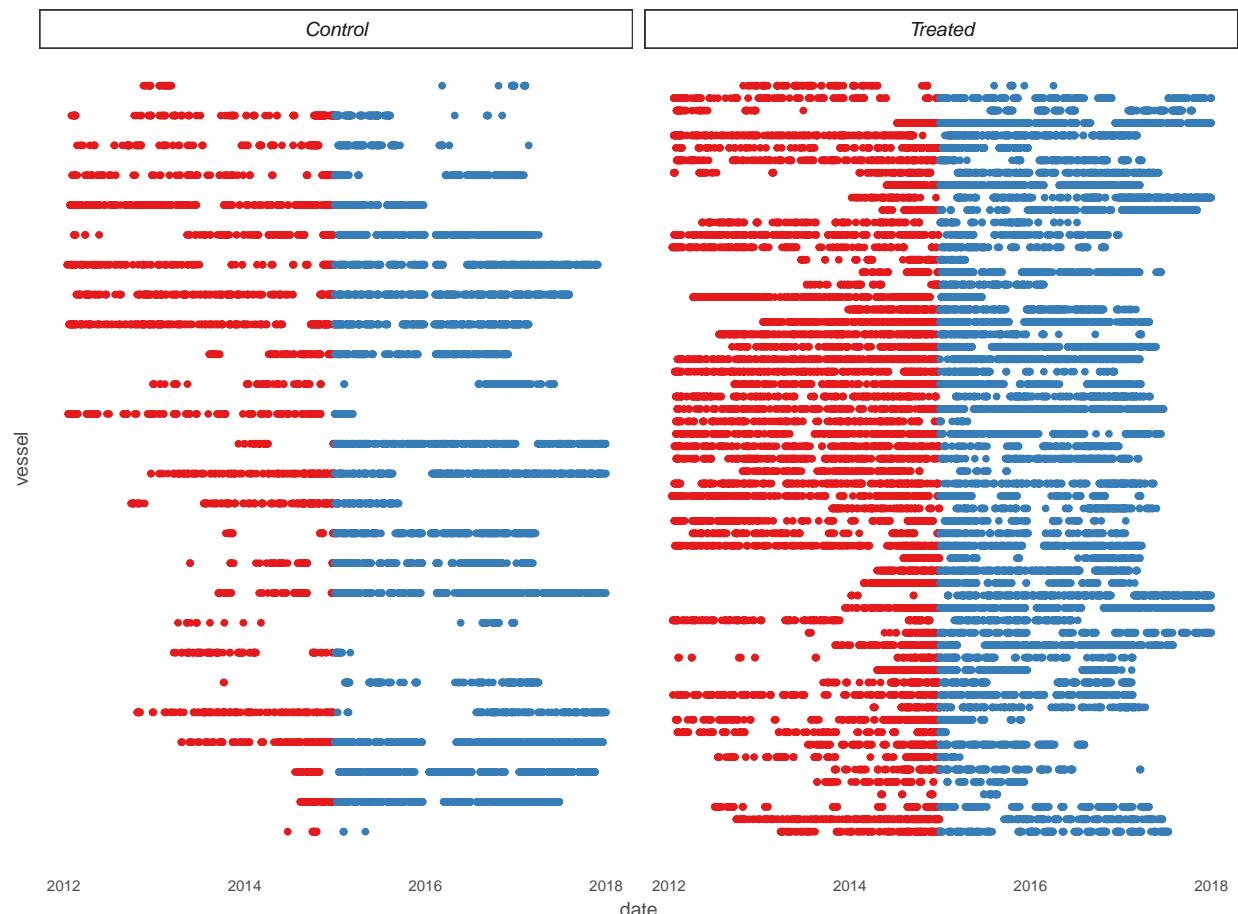


Figure S1: Stream of fishing events by vessels through time. Each line represents a vessel, with dots indicating months with fishing activity and colors indicating the pre and post periods.

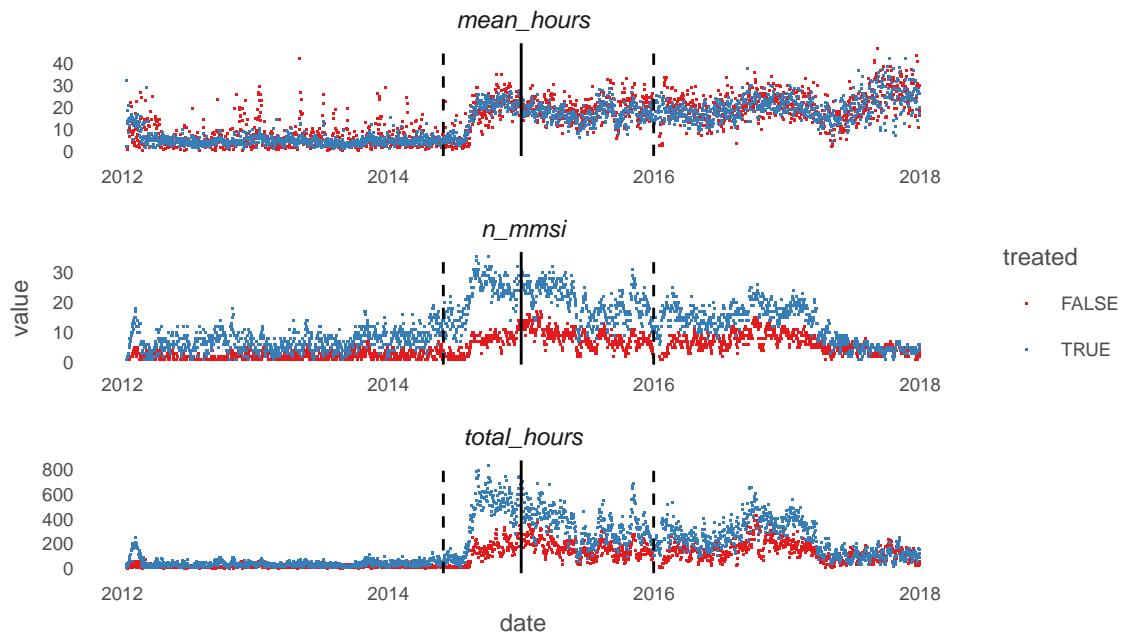


Figure S2: Fishing hours and number of vessels by month for all vessels. Vertical dashed lines indicates dates when satellites were added, solid line indicates PIPA closure.

626 9.2 Longliners

627 Longliners show a similar pattern of effort reduction. However, the magnitude of the β_3 coefficient
 628 is smaller and not significant unless including month and flag FEs for the general treatment-control
 629 groups, or when excluding Chinese vessels. This, along with higher standard error values suggest
 630 that longliners have a smaller and more variable response to the implementation of LSMPAs.

Table S1: Difference-in-differences estimates for 2 different controls and 3 different specifications. The first three columns use all data for controls. Columns 4 - 6 exclude Chinese vessels. Numbers in parentheses are heteroskedastic-robust standard errors.

	<i>Dependent variable:</i>					
	hours					
	(1)	(2)	(3)	(4)	(5)	(6)
Constant	39.192*** (0.095)	38.411*** (0.132)	36.520*** (0.169)	39.402*** (0.102)	38.636*** (0.140)	34.402*** (0.236)
post	-1.014*** (0.143)	-0.189 (0.145)	0.710*** (0.158)	-1.400*** (0.163)	-0.574*** (0.165)	0.896*** (0.186)
treated	2.544*** (0.101)	2.570*** (0.101)	2.724*** (0.121)	2.349*** (0.110)	2.391*** (0.110)	3.299*** (0.135)
sate2	-0.495*** (0.099)	-1.488*** (0.105)	-0.898*** (0.110)	-0.561*** (0.104)	-1.497*** (0.110)	-1.015*** (0.114)
sate3	-0.313** (0.129)	-1.150*** (0.132)	0.050 (0.143)	-0.375*** (0.134)	-1.129*** (0.138)	0.001 (0.149)
post:treated	0.098 (0.134)	0.056 (0.134)	-1.125*** (0.148)	0.532*** (0.149)	0.427*** (0.149)	-1.280*** (0.171)
Control	All	All	All	-CHN	-CHN	-CHN
Month FE	No	Yes	Yes	No	Yes	Yes
Flag FE	No	No	Yes	No	No	Yes
Observations	227,873	227,873	217,467	209,135	209,135	198,729
R ²	0.010	0.016	0.022	0.010	0.016	0.023

Note:

*p<0.1; **p<0.05; ***p<0.01

631 9.2.1 Quarterly DID interactions

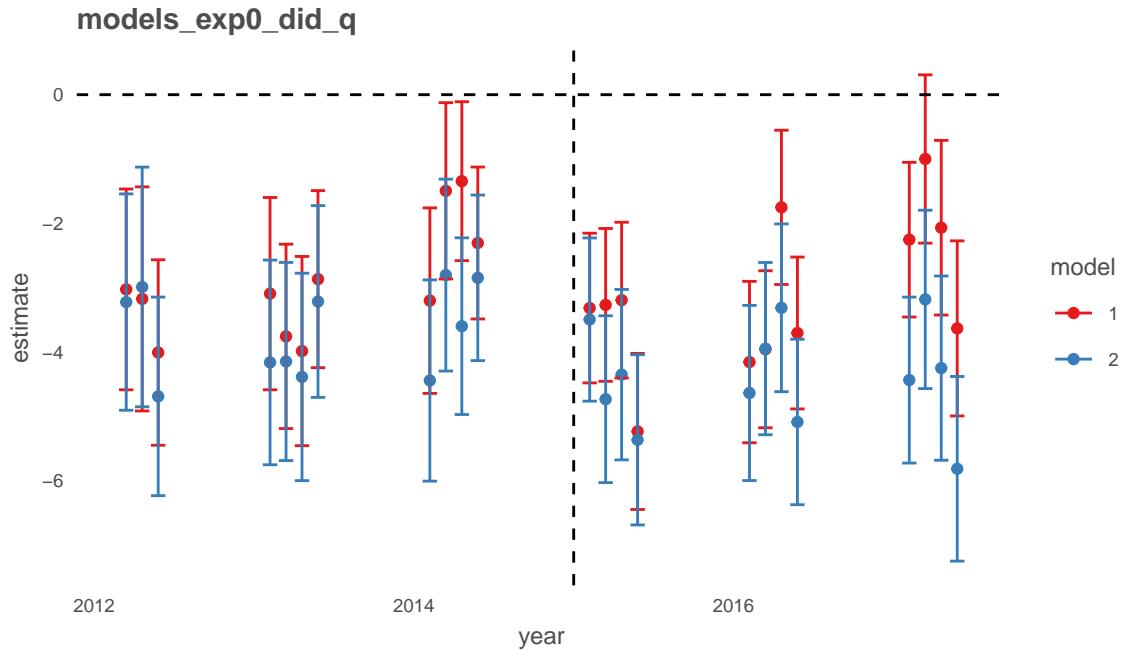


Figure S3: Interaction of quarters and treatment. Control group is all vessels.

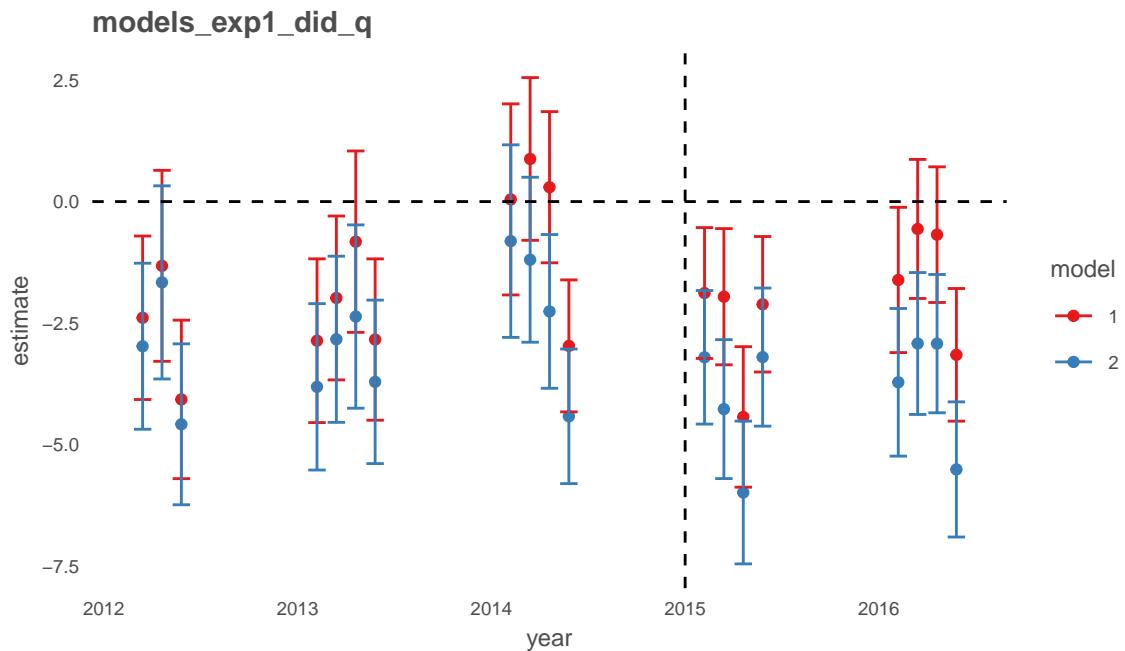


Figure S4: Interaction of quarters and treatment. Control group is vessels from PNA.

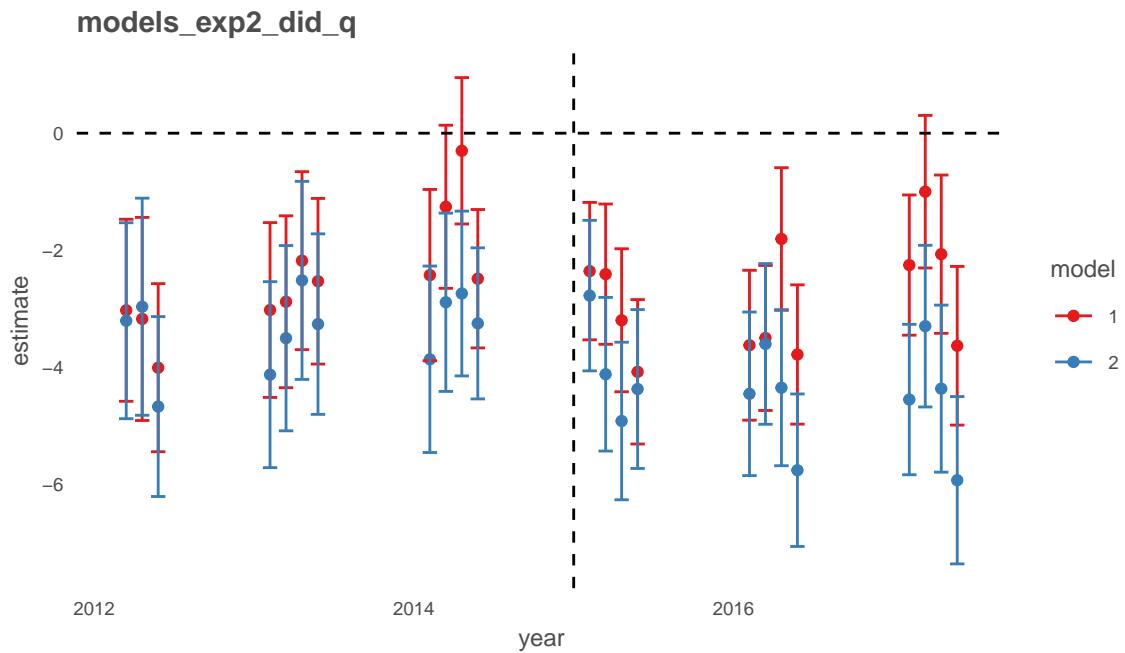


Figure S5: Interaction of quarters and treatment. Control group excludes Chinese vessels.

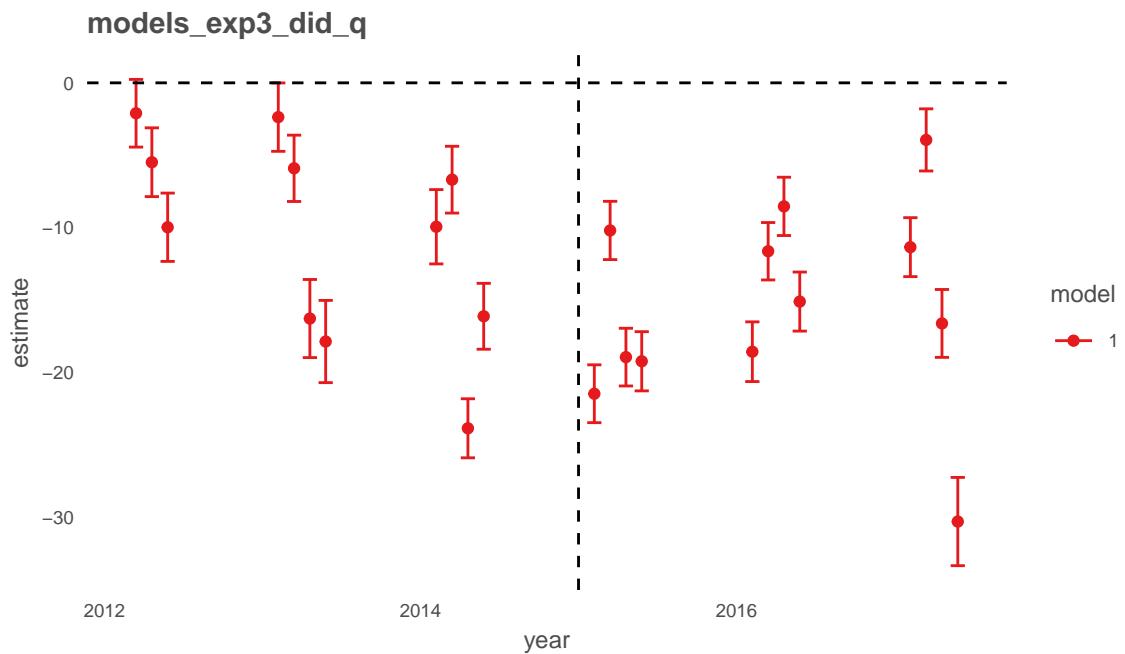


Figure S6: Interaction of quarters and treatment. Control group is Japanese vessels.

632 9.2.2 Year-month DID interactions

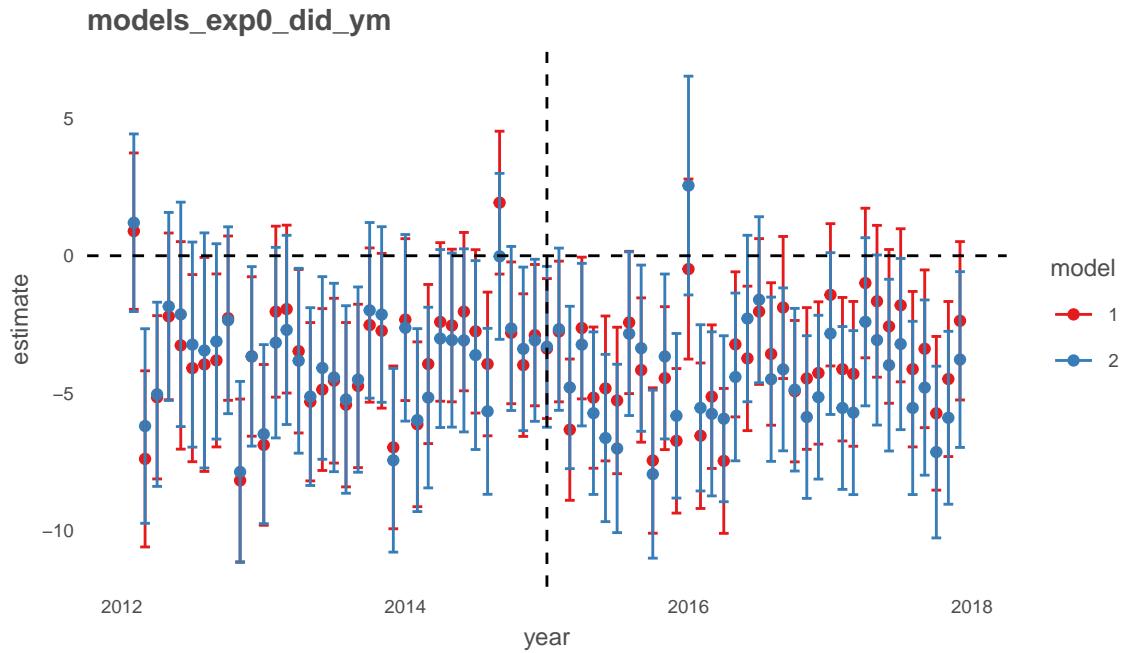


Figure S7: Interaction of year-month and treatment. Control group is all vessels.

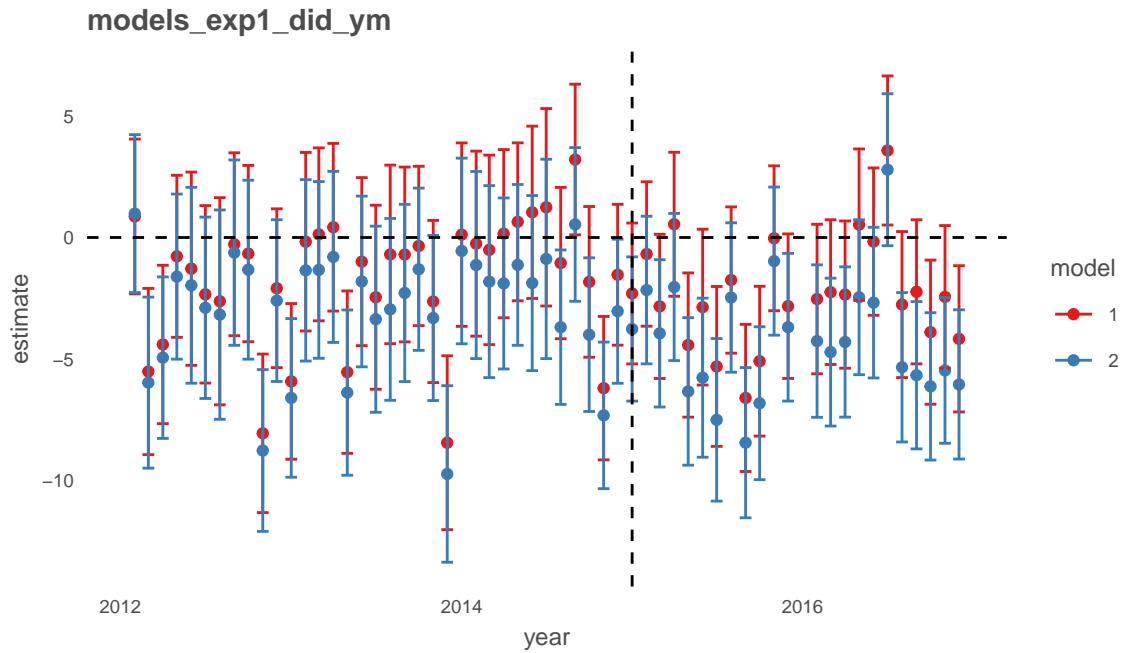


Figure S8: Interaction of year-month and treatment. Control group is vessels from PNA.

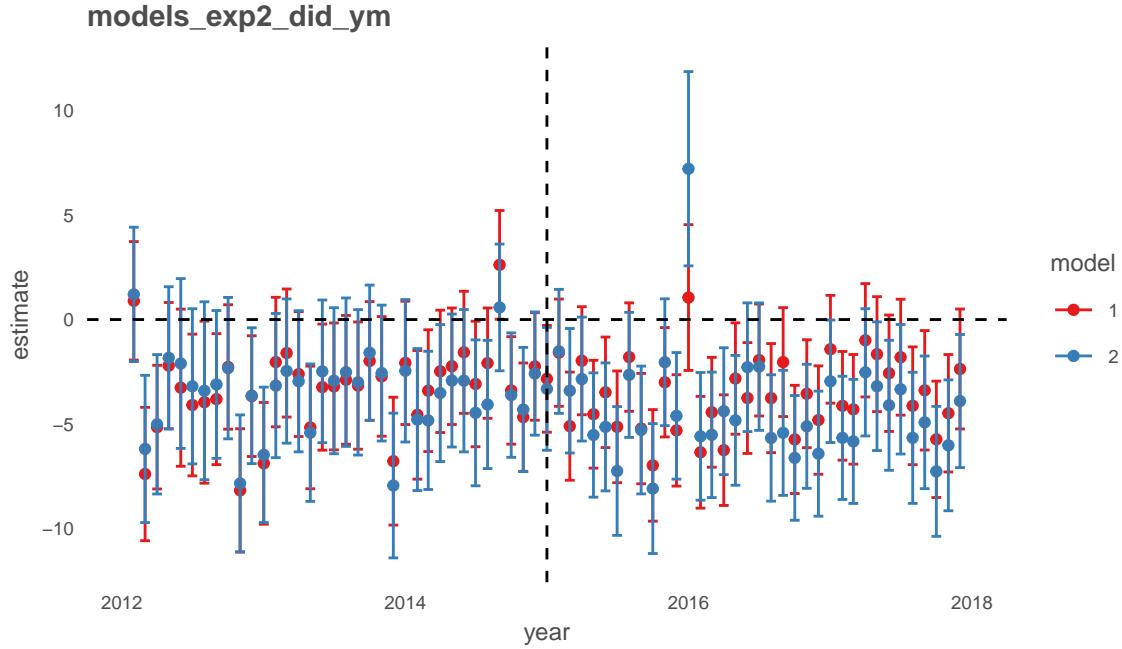


Figure S9: Interaction of year-month and treatment. Control group excludes Chinese vessels.

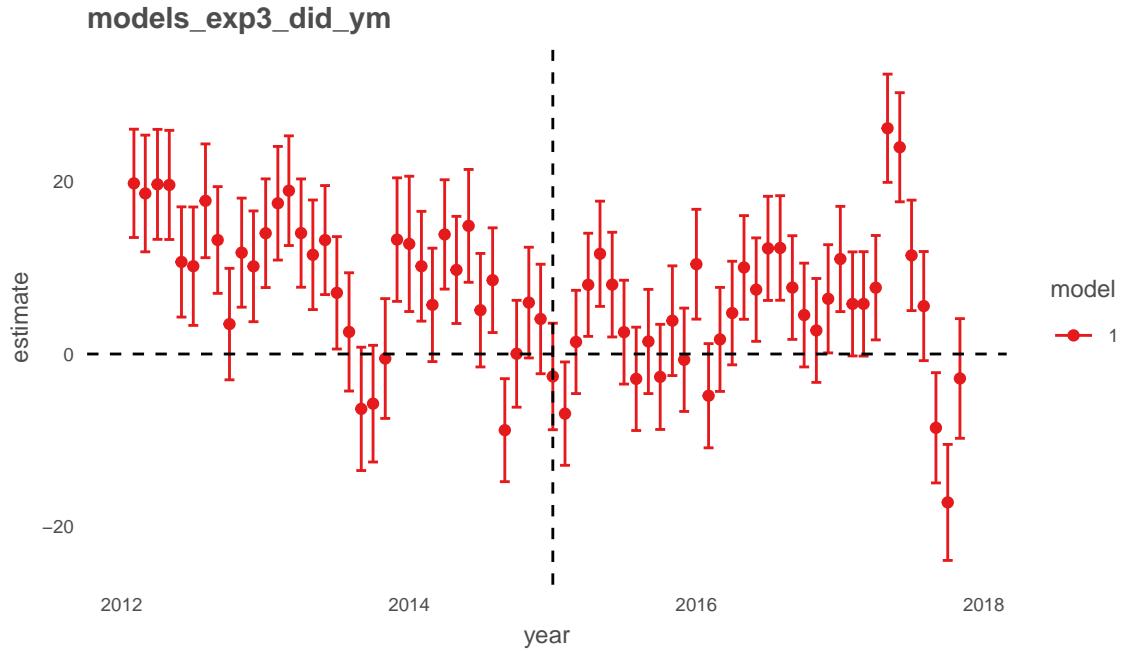


Figure S10: Interaction of year-month and treatment. Control group is Japanese vessels.

633 9.3 Other measures of change in behavior

634 Global Fishing Watch data includes two variables of interest that are readily available: distance
 635 from shore and distance from port. We use these as variables of interest and perform the following

⁶³⁶ regression:

$$y = \beta_1 Year + \beta_2 Treated + \beta_3 Year \cdot Treated + \epsilon$$

⁶³⁷ Where y is our variable of interest, $Year$ represent year dummies, with 2012 as the reference level,

⁶³⁸ and $Treated$ indicates whether a vessel belongs to the treated or control group.

⁶³⁹ **9.3.1 Distance from shore**

Table S2: Regression results for distance from shore. Numbers in parentheses are heteroskedastic-robust standard errors.

	<i>Dependent variable:</i>
	distance_from_shore
Constant	233,057.900*** (1,558.294)
year_c2013	−818.044 (2,623.263)
year_c2014	14,220.260*** (1,701.637)
year_c2015	140,980.200*** (1,668.140)
year_c2016	32,557.750*** (1,584.182)
year_c2017	168,283.100*** (1,653.912)
treated	−11,371.280*** (1,688.879)
year_c2013:treated	62,899.900*** (3,274.205)
year_c2014:treated	−17,499.020*** (1,834.090)
year_c2015:treated	−34,356.840*** (1,821.587)
year_c2016:treated	72,512.980*** (1,739.976)
year_c2017:treated	−62,350.120*** (1,805.322)
Observations	4,937,666
R ²	0.032
Adjusted R ²	0.032
Residual Std. Error	297,128.600 (df = 4937654)
F Statistic	14,649.360*** (df = 11; 4937654)

Note:

*p<0.1; **p<0.05; ***p<0.01

640 **9.3.2 Distance from ports**

641 **9.4 Effort redistribution**

642 **9.4.1 Rasterized regions**



Figure S11: Rasterized regions. Each color (number) indicates a distinct region.

Table S3: Regression results for distance from shore

	<i>Dependent variable:</i>
	distance_from_port
Constant	362.707*** (1.971)
year_c2013	9.082*** (3.198)
year_c2014	105.414*** (2.148)
year_c2015	357.746*** (2.113)
year_c2016	112.441*** (2.009)
year_c2017	356.506*** (2.099)
treated	149.653*** (2.249)
year_c2013:treated	26.366*** (4.125)
year_c2014:treated	50.808*** (2.453)
year_c2015:treated	−151.268*** (2.422)
year_c2016:treated	−49.445*** (2.311)
year_c2017:treated	−252.334*** (2.390)
Observations	4,937,666
R ²	0.047
Adjusted R ²	0.047
Residual Std. Error	391.757 (df = 4937654)
F Statistic	22,351.480*** (df = 11; 4937654)

Note:

*p<0.1; **p<0.05; ***p<0.01

643 9.4.2 Fishing raster

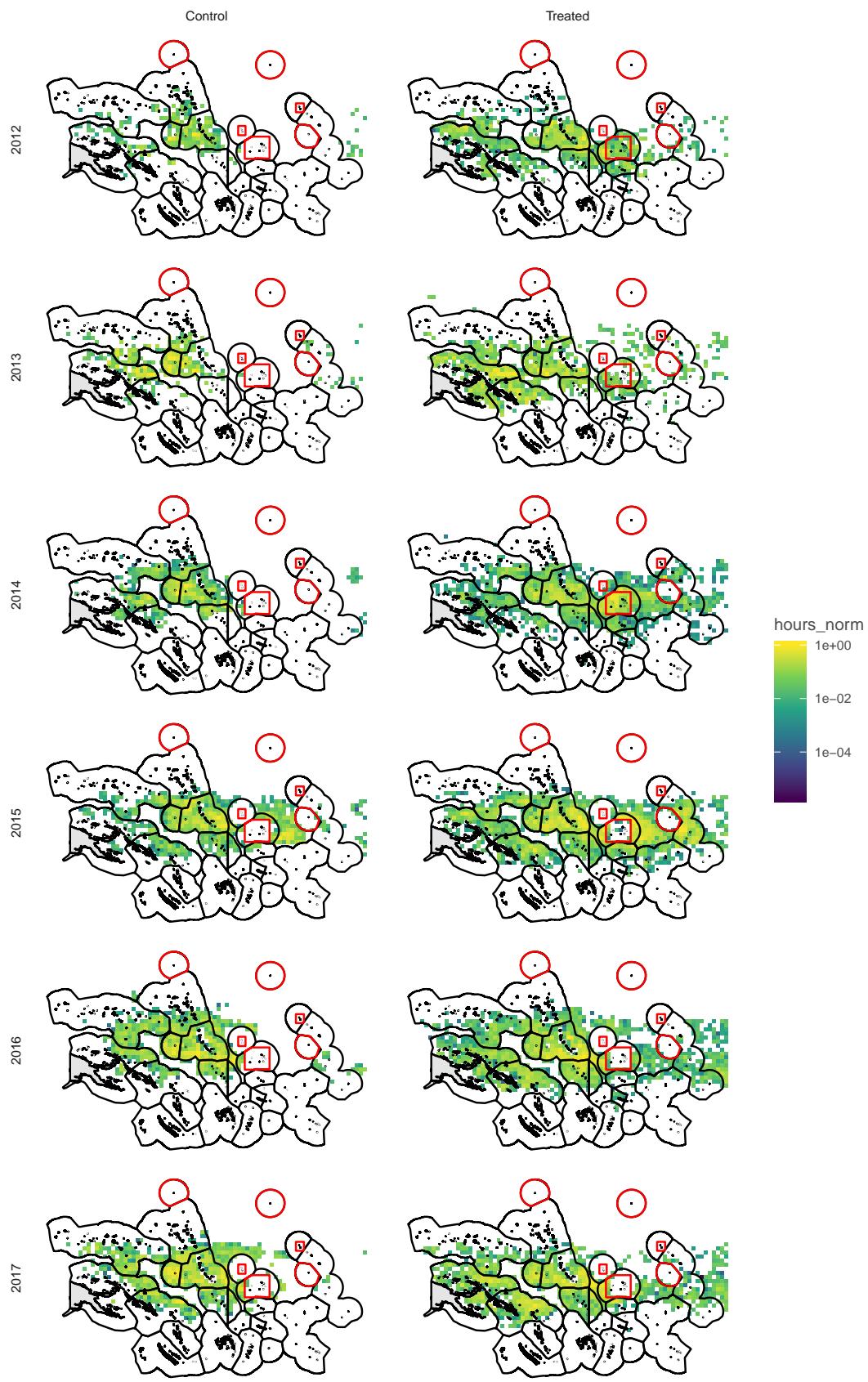


Figure S12: Yearly rasters of fishing hours for each group.