

Marine Conservation In Fishing Effort Markets

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Our oceans are becoming increasingly crowded and privatized as different activities compete for space. Marine Protected Areas have the primary objective of conserving bounded waters by displacing or limiting fishing effort. With global conservation targets seeking to protect 10% of the world's ocean by 2020, it is important that the spatial displacement of economic activities is fully considered. A spatial closure in a vessel-day-managed fishery can potentially result in large costs to the implementing country, which might deter conservation efforts. We show that effort markets may reduce up to 70% of the costs, and that allocation rules based on biomass (rather than effort) may incentivize conservation. We then use identification of fishing activity via Automatic Identification Systems to evaluate the effect that the Phoenix Islands Protected Area (PIPA, 397,447 km², implemented in 2015 in Kiribati) had on vessel distribution and behavior. Aggregate fishing effort remains relatively constant following the closure, but is significantly reduced within Kiribati, especially by vessels that historically fished within PIPA before its implementa-

tion. We find evidence that Kiribati may be trading its surplus of vessel-days, which would significantly reduce the costs of displacing fishing effort. We use these insights predict the implications of a proposed Large-Scale Marine Protected Area (LSMPA) in Palau (another PNA member) and estimate potential losses to range from \$2.5 to \$11 million annually. Our work highlights the role of effort markets in enabling conservation, and the importance of allocation rules that praise, not punish, conservation. The Parties to the Nauru Agreement have successfully managed their fishery and developed a management scheme that allows for conservation interventions to come at a low cost.

1 Introduction

Humans are increasingly utilizing the oceans. Multiple ocean uses such as off-shore aquaculture, conservation, energy harvesting, deep-sea mining, and fisheries are likely to compete for space. As we move forward with blue growth, we must understand the potential effects of activities displacing each other and establish causal links between past management interventions and their outcomes (1). One of the most notable spatial interventions is the creation of no-take Marine Protected Areas (MPAs), which seek to conserve the environment by eliminating fishing effort within their waters.

Today, a small number of Large-scale Marine Protected Areas (LSMPAs; areas larger than 30,000 km² *sensu* (2)) represent at least 80% of the protected area in the ocean (Fig. 1; (3)). However, very little is known about their human dimensions and implications for fisheries (4). Furthermore, research on LSMPAs has focused on their potential ecological benefits, but have left aside the economic implications. One issue of particular importance is that of the displacement or redistribution of fishing effort, which may influence the outcomes of a spatial closure and represent large opportunity costs (5, 6).

Displacement or elimination of fishing effort within the protected area might come at a cost when fisheries are managed under an effort market. For example, tuna purse seine fisheries in the Pacific are collectively managed under a Vessel-Day Scheme (VDS) by nine countries commonly referred to as the Parties to the Nauru Agreement (PNA; Figure S1). The agreement regulates the access of foreign fishing vessels to PNA waters, and control the allocation and sales of 45,000 vessel-days. Holding 80% of historical skipjack tuna purse seine grounds within their joint Exclusive Economic Zones (EEZ; with surface area of 14.6 million km²), PNA countries have achieved greater bargaining power (7); vessel-day price rose from \$5,000 USD in 2012 to at least \$9,000 USD in 2016, driven by increasing price floors. The revenue from access fees may represent up to 50% of government revenue for some of the parties.

The Phoenix Islands Protected Area (PIPA) in Kiribati is one of the most notable Large-Scale Marine Protected Areas on earth. Implemented on January 1st of 2015, PIPA closed an area of 397,447 km², about 11% of Kiribati's EEZ or 2.7% of total PNA area, effectively displacing all fishing effort within it (8, 9). A pressing question is that of the costs associated to the displacement of fishing effort and possible revenue losses from the fishing license sales. While no studies have assessed the implications of PIPA, other PNA members have pledged the implementation of LSMPAs, like Palau's 80% closure to go into effect in 2020 (10).

We develop a bioeconomic model of a fishery operating under a vessel-day scheme with spatial closures to explore the effects of effort redistribution resulting from MPA implementation. We characterize the likely outcomes with and without vessel-day trading between countries, as well as the effect of different allocation rules. We then empirically evaluate the behavioral responses and spatial redistribution of the industrial tuna purse seine fleet operating within PNA waters after the implementation of the Phoenix Islands Protected Area, and quantify its economic ramifications and implications to Kiribati. We use the same data and insights from our model to hypothesize what might be the impacts of the proposed Palau National Marine Sanc-

tuary. These are two of the largest protected areas on the planet and both are controlled by PNA countries.

2 Results

2.1 Fishing effort markets and marine conservation

We begin by simulating a spatial closure in a world where vessel-days can not be traded between countries. A spatial closure in Patch 1 always results in a loss or no-change in revenue from vessel-days, even when the stock moves freely between the protected and non-protected portions of the Patch (Fig. 2A). The losses are caused by a drop in vessel-day prices for Patch 1, which results from a lower portion of the stock being available for harvest (Fig. S2). Costs increase with reserve size (R), and decrease as within-patch movement (θ) increases. Given the non-linear costs ($\beta > 1$), the closure-to-cost ratio can be greater than 1:1 if within-patch movement is low (*i.e.* $\theta < 0.2$). The spatial closure in Patch 1 results in moderate increases in revenue for Patches 2 - 9 (Fig. S3). When countries do not trade between them, the costs of conservation are large and incurred by Patch 1, but the benefits perceived by the eight other patches. Importantly, the increases in revenue to other patches do not compensate for the total losses to Patch 1.

The objective of a VDS is to allow for trading between parties and achieve the optimal distribution of effort in relation to fish distribution. In the case of the PNA, for example, spatial distribution and longitudinal shifts of tuna -particularly skipjack- have been linked to ENSO events (11). Vessels redistribute across PNA countries as they track fish, and tradable permits reduce costs to the parties (12). We explore the effect of a spatial closure in Patch 1, but now allow for vessel-days to be traded between Patches. Patch-level demand for vessel days is governed by the fishable stock size (*i.e.* biomass not within the reserve). Trading occurs until vessel-day prices are the same across patches, given a total allowable effort (*i.e.* 45,000 vessel-

days) is reached. Here, costs from a closure in Patch 1 are significantly reduced by trading (Fig. 2B-C). In this case, surplus of vessel-days in Patch 1 are sold to Patches 2 - 9. Vessel-day prices are constant across patches and lower than when there is no reserve, but still higher than when no trading is allowed (Fig. S4). By optimally trading vessel-days, the cost to the implementing country can be reduced as much as 97%.

Our previous analyses estimate loss in revenue from effort displacement. Annual vessel-days are often allocated based on a combination of historical within-patch effort and biomass. In the PNA, for example, sixty percent of the allocation is calculated based on EEZ effort over the last seven years and 40% is calculated based on the 10-year average of each countrys share of estimated skipjack and yellowfin biomass within its EEZ.¹ Trading vessel-days to other countries would imply that historical within-patch effort declines through time. The allocated days to the Patch with the spatial closure would eventually be reduced to the 40% depending on biomass.

We explore the effect that allocation rules have through time when trading is allowed by projecting the fishery 50 years into the future with. We used a range of allocation rules that weight effort (α) and biomass ($1 - \alpha$) differently to update annual allocations based on a seven-year running mean of these measures. Costs are greater when allocation is based only on patch-level historical effort (*i.e.* $\alpha = 1$; Fig. 3). However, costs can be significantly reduced when trading is allowed if allocation is based on historical biomass instead. When there is no trading, patch-level effort remains unchanged, and since any growth in stock is equally distributed across patches, allocation rules do not affect costs in a case without trading.

¹This is explained in more detail in Article 12.5 of the 2012 Amendment to the Palau Agreement and in (13)

2.2 Phoenix Islands Protected Area

We use identification of fishing activity via Automatic Identification Systems (AIS) to track 313 tuna purse seine vessels that fished in PNA waters between 2012 and 2018 (9, 14). We track the spatial redistribution of the fleet and explore behavioral changes following the 2015 implementation of PIPA in Kiribati. We continuously observe 92 vessels for the 2012 - 2018 period. Of these, 64 vessels fished within PIPA at least once prior to its implementation, and we refer to them as the “displaced vessels”. The remaining 28 vessels never fished in PIPA waters, and we refer to them as “non-displaced vessels”. The group with the remaining 221 vessels contains vessels that were not continuously observed before and after the implementation of PIPA, and we refer to these as “other vessels”. Our analyses incorporate the monthly NINO4 anomaly index (Fig. S5) to control for the association that fish and vessels might have to environmental variation (11, 12, 14).

A market for fishing effort and allocation rules can reduce the costs of displacing fishing effort, but a spatial closure is still likely to cause behavioral changes as vessels redistribute. We find that between 2015 and 2016, displaced vessels spent 3,916 and 2,249 less vessel-days in Kiribati and PNA waters, respectively (Figs. 4). Over the same period, non-displaced and other vessels spent 1,278 less days in Kiribati, but spent 9,853 more days in PNA waters overall. These result in a net loss of 5,195 vessel-days for Kiribati, and a net gain of 7,600 vessel-days at the PNA-level (*i.e.* the other 8 countries). At a vessel-day price of \$9,000 USD per day, these would represent \$46.7 and \$68 million USD, respectively.

Annual vessel-days in Kiribati continued to decrease to just 7,479 in 2018, mainly due to displaced vessels allocating less time to Kiribati (Fig. 4B); Fig. S10). Current effort in Kiribati waters are well below their 11,000 vessel-day allocation (15). However, we observe a relatively constant 45,000 vessel-days at the PNA-level, suggesting that Kiribati is, in fact, able to trade vessel-days and reduce the costs of displacement. The spatial redistribution patterns of

displaced vessels relative to non-displaced vessels are shown in Figure S6.

We complement our analysis of change in observed vessel-days by looking at country-level data reported by the Pacific Islands Forum Fisheries Agency (FFA). We find that Kiribati's revenue went from \$148.8 million USD in 2015 to \$118.3 million USD in 2016, representing a decrease of \$30.5 million USD. Total PNA revenues showed a net increase of \$28 million USD (Fig. S13). The largest decrease was observed for Kiribati, while the largest increase was also observed for Papua New Guinea.

Total fishing effort in the PNA increased between 2015 and 2016, but fishing area decreased due to PIPA implementation (Fig. 4). We inspect the crowding effects that may arise by applying more (or the same amount of) fishing effort over less fishing area with two indices of spatial overlap between displaced and non-displaced vessels: 1) the number of cells that had fishing activity from both groups for each month and 2) the correlation of presence/absence of fishing activity between both groups over one month. The two fleets significantly interact more with each other after the implementation of PIPA (Table S1 Fig. 5). The number of cells with presence from both fleets and spatial correlation increase by a factor of four and three, respectively. Crowding might be driven by environmental variation concentrating fish at specific locations, but the NINO4 anomaly alone explained just 3% and 7% of the variation in our crowding measures, compared to the 70% variation explained when accounting for PIPA implementation. We observe similar patterns when repeating the crowding exercise for Kiribati's EEZ only (Table S2, Fig S7).

Effort displacement and crowding are likely to induce behavioral changes in the fishing fleet. We calculate nine vessel-level measures that could capture responses to spatial closures: daily fishing hours, daily non-fishing at-sea hours, the proportion of fishing to non-fishing at-sea hours, daily distance traveled, daily mean distance from shore of fishing events (km), daily mean distance from port of fishing events (km), as well as monthly hours spent in PNA waters, Kiribati

waters, and the High Seas (Fig. S8). We find no evidence of displaced vessels fishing more after PIPA implementation, and in fact observe a 27.5% decrease ($p < 0.01$; Table S3) relative to non-displaced vessels. Likewise, we observe a 3.4% decrease of fishing hours relative to total at-sea hours ($p < 0.01$; Table S3). Treated vessels traveled 23.3% less distance, and fishing events occurred 32.9% and 16.9% closer to shore and to port, respectively. These changes in distance from shore and port are likely caused by redistribution, as we observe that displaced vessels fished 79.67% and 49.6% less in Kiribati and PNA waters respectively, compared to the trend observed for non-displaced vessels ($p < 0.01$). We do not observe a statistically significant increase in non-fishing at-sea hours (*i.e.* a proxy for search time) and fishing hours on the High Seas. We repeat this analysis for sample restrictions where we exclude all Chinese vessels (Table S4), all PNA-owned vessels (Table S5) and all Taiwanese and USA vessels (Table S6) and find qualitatively the same responses.

2.3 Palau National Marine Sanctuary

On October 28, 2015, the President of Palau signed into law the Palau National Marine Sanctuary (PNMS) Act. Starting in December 2020, this Act will close 500,000 km² (about 80 percent of Palau's EEZ) to commercial fishing activities, creating the 14th largest protected area in the world, effectively displacing all effort within its waters (Fig. S16 shows longline and purse seine fishing effort within PNMS for 2018). As part of the Nauru Agreement, Palau receives 700 purse seine vessel-days each year. They also use a vessel-day scheme to manage the longline fishery (10,500 vessel-days per year). Unlike the purse seine vessel days, longline vessel-days are not tradable within countries and are worth a fraction, at about \$200 USD each. Our previous analyses suggest that Palau should be able to lessen the costs of the closure by trading purse seine vessel-days (\$5.65 - \$8.75 million USD) with other PNA members, but might lose more than 50% of the revenues from the longline vessel-days (\$2.1 million USD).

Table 1 presents estimates of the potential revenue losses following full enactment of the PNMS under four different scenarios. In Scenario 1, Palau retains its current allotment of purse seine vessel-days and is able to sell them for a similar price to what it is currently selling them to the United States under the South Pacific Tuna Treaty (also known as the Multilateral Treaty on Fisheries; \$12,500/day). In Scenario 2, Palau is able to keep its current allotment of purse seine VDS to transfer to other PNA countries at the current benchmark price (\$8,000/day). Scenario 2 is likely if Palau retains its current allocation, but the US no longer purchases days. It should be noted that if allocation continues to be calculated based on effort and biomass, and if Palau continues to be allocated vessel days, its allocation will decrease as purse seine effort in its EEZ eventually reaches zero. In Scenarios 3 and 4, Palau loses all of its purse seine VDS, at \$8,000/day and \$12,500/day, respectively. In all scenarios, all longline vessel days and export tax revenues are lost, since longline vessel days are currently not tradable and Palau is planning on banning the export of fish. The longline vessel day loss is calculated using an average value of \$200 for 10,500 days. The export tax loss is calculated given the average tax revenue from 2012-2014 (\$482,236 from (16)).

Table 1: Estimated revenue losses under different scenarios of PNMS (in USD)

Scenario	PS VDS	LL VDS	Export tax	Total revenue loss
1	0	-2,100,000	-482,236	-2,582,236
2	-3,150,000	-2,100,000	-482,236	-5,732,236
3	-5,600,000	-2,100,000	-482,236	-8,182,236
4	-8,750,000	-2,100,000	-482,236	-11,332,236

3 Discussion

Our findings provide insights into the effect that LSMPAs can have on redistribution of fishing effort and change in behavior. Our analyses predicts losses in revenue to countries that implement a spatial closure. However, these losses can be significantly reduced by market instruments and biomass-based allocation rules. Using vessel tracking data, we see that vessels that fished inside PIPA before the implementation redistribute to other areas within Kiribati, but also other PNA countries and the high seas. Surprisingly, there is no drop off in fishing effort in Kiribati in 2015 but a noticeable drop from 2016 onward. Redistribution lead to a crowding effect, and the spatial closure had little effect on the *total* fishing effort exerted by purse seiners, but vessel-level responses were heterogeneous. Our analysis suggests that losses would be expected for Palau’s Marine Sanctuary, but that these can be minimized if effort is traded freely and allocations are based in biomass, not effort. Here, we discuss the implications of our findings and possible shortcomings in our analysis.

We observe 50% discrepancies in our estimates of lost revenues when using tracking data and FFA data. Vessel-day usage through time matches the 45,000 vessel-days for the PNA and the 11,000 vessel-days allocated to Kiribati (15). Moreover, the trends through time are certainly similar, and all indicate a reduction of usage of Kiribati’s waters. A possible explanation is that we used the reported mean annual vessel-day prices in our calculations, but there is certainly a spread around these. Furthermore, vessels can report their activity as “transiting to port” or “solving mechanical issues” and this time does not count towards a vessel’s usage of VDS, which would result in an overestimation from tracking data.

ENSO events are known to drive the location of fish and behavior of fishing vessels, especially in PNA waters (11, 12, 14). We do our best to control for these environmental changes by incorporating the NINO4 anomaly index in our analyses, and by tracking the non-displaced

vessels as a “control” group that is equally affected by the environmental variation. For example, we observe that both displaced and non-displaced vessels shifted their effort post-PIPA to the Western margin of the PNA region, namely Kiribati’s Gilbert Islands and Tuvalu. However, displaced vessels redistribute a greater proportion of fishing effort to these areas, as well as the High Seas (Fig S6). Our analysis shows that sea surface temperature variation does have an effect on our crowding and behavioral methods, but that by itself it does not explain the observed patterns. While environmental variation certainly influences fishing behavior, policy and management interventions such as moratoriums and spatial closures can have even larger effects (14).

A major shortcoming of our analysis is that we do not observe catch or revenue at the vessel level, which ultimately guides the decision-making process of vessel captains. Therefore, it is difficult to know whether the small change in fishing hours and redistribution represents a positive or negative impact. Likewise, the available data from the FFA does not cover the 2017 and 2018 period, and we do not observe vessel-day transactions or between countries or detailed vessel-day prices.

Our work suggests that the implementation of a Large-Scale Marine Protected Area had little impact on total fishing effort, but that it may result in revenue losses for countries that sell fishing rights. The closure caused a crowding effect, which causes some vessels to redistribute to areas close by. The Parties to the Nauru Agreement have shown that multinational cooperation in management of transboundary resources can have management and economic benefits (12, 17). Overlaying conservation interventions over existing management schemes must be accompanied by similar dialogue and international cooperation, so that the benefits of conservation are captured by those incurring the costs. Furthermore implementation of Large-Scale Marine Protected Areas must be accompanied by traditional fisheries management to maximize effectiveness, consider the opportunity costs of such closures (6), and identify sustainable financ-

ing mechanisms (18) that would compensate losses and incentivize marine conservation in the long term.

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4 Figures and tables

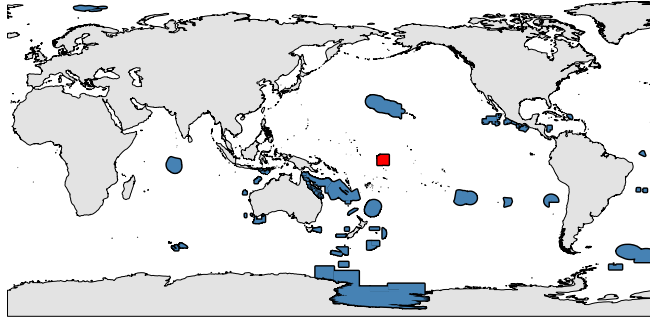


Figure 1: Large Scale Marine Protected Areas. The map shows all areas larger than 30,000 Km². The Phoenix Islands Protected Area is shown in red.

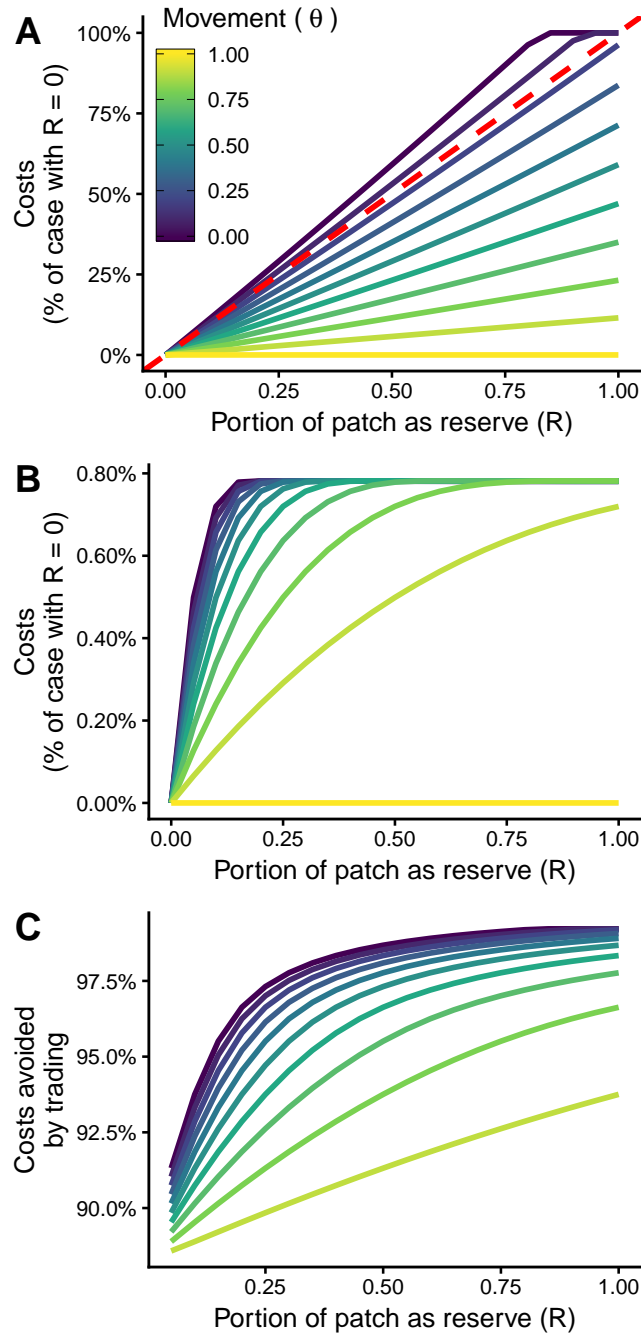


Figure 2: Cost of spatial closures in a vessel-day fishery (vertical axis) as a function of reserve size (R ; horizontal axis) and movement (θ ; line color). Costs are shown for Patch 1 when there is no trading (A) and when trading is allowed (B). Costs avoided by trading are shown in (C).

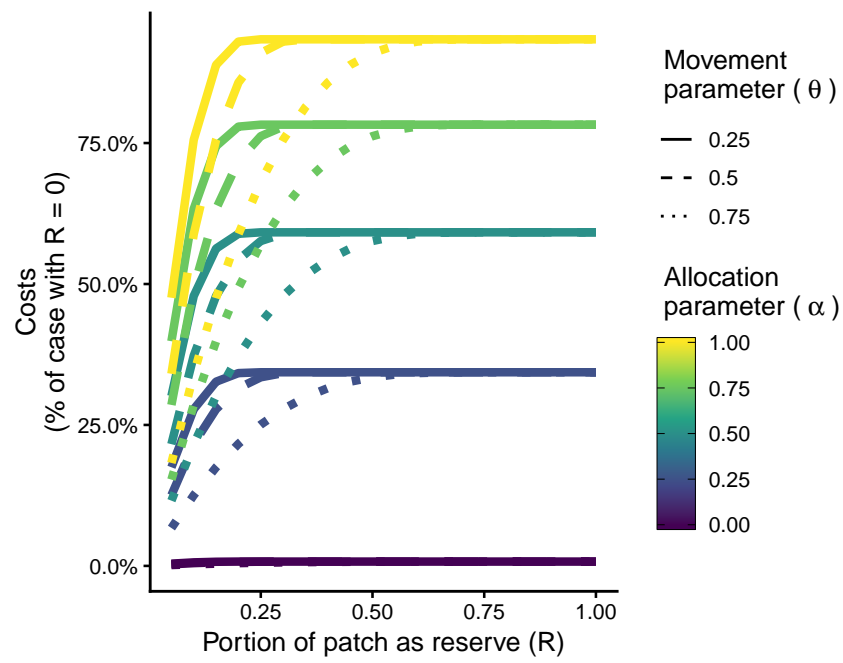


Figure 3: Cost (vertical axis) of different allocation rules (line colors) as a function of reserve size (R ; horizontal axis) and movement (θ ; line type). Costs are shown for Patch 1 only.

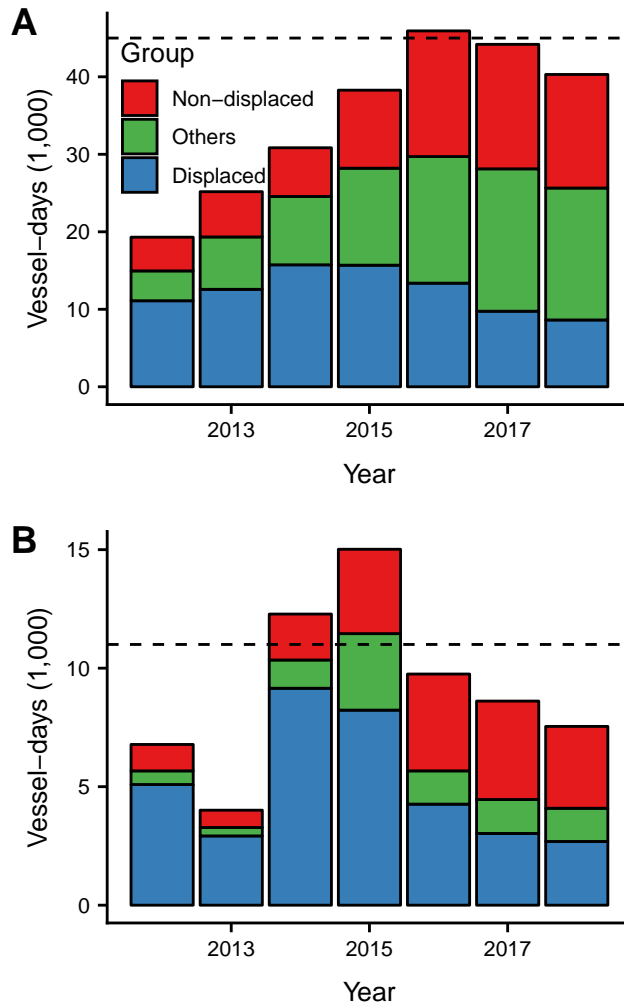


Figure 4: Observed vessel-days for all PNA countries (A) and Kiribati (B) by displaced, non-displaced, and other vessels. Vessels displaced from PIPA fish less within the PNA, but non-displaced and other vessels fish more, resulting in a net increase. Vessels that used to fish within PIPA account for the largest effort reduction to Kiribati. The dashed horizontal lines represent the total allowable effort in the PNA (45,000 days) and Kiribati (11,000 days).

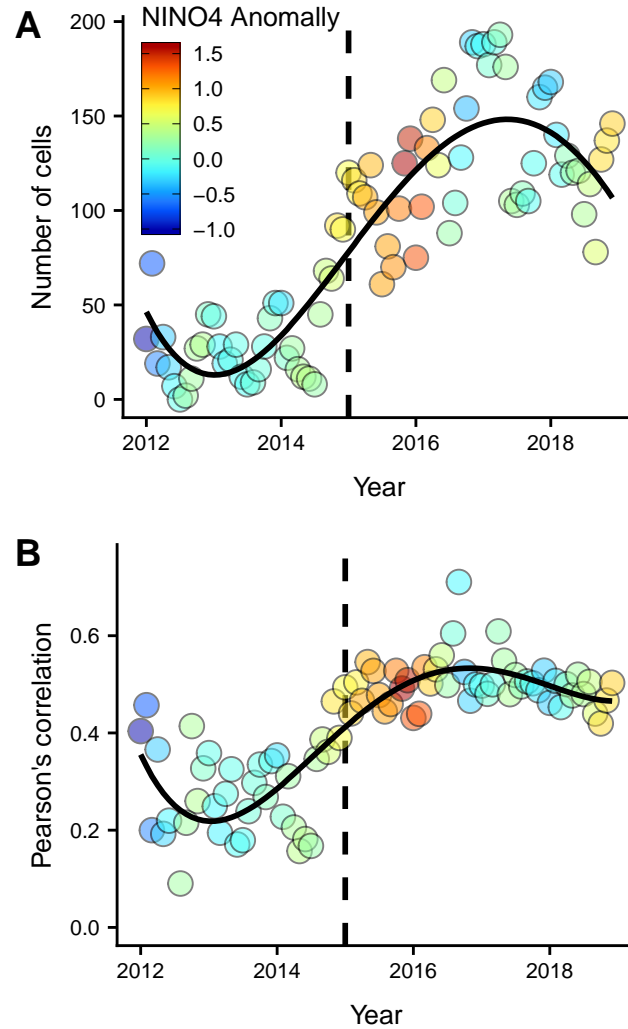


Figure 5: Number of cells that had displaced and non-displaced vessels (A) and spatial correlation in the presence-absence of each group per cell (B). The solid lines represent the 4th degree polynomial fit reported in S1. Note that the late 2016 and early 2017 showed negative or neutral NINO4 anomalies similar to those in the pre-PIPA period, but crowding effect does not decline to pre-implementation level and seems to stabilize.

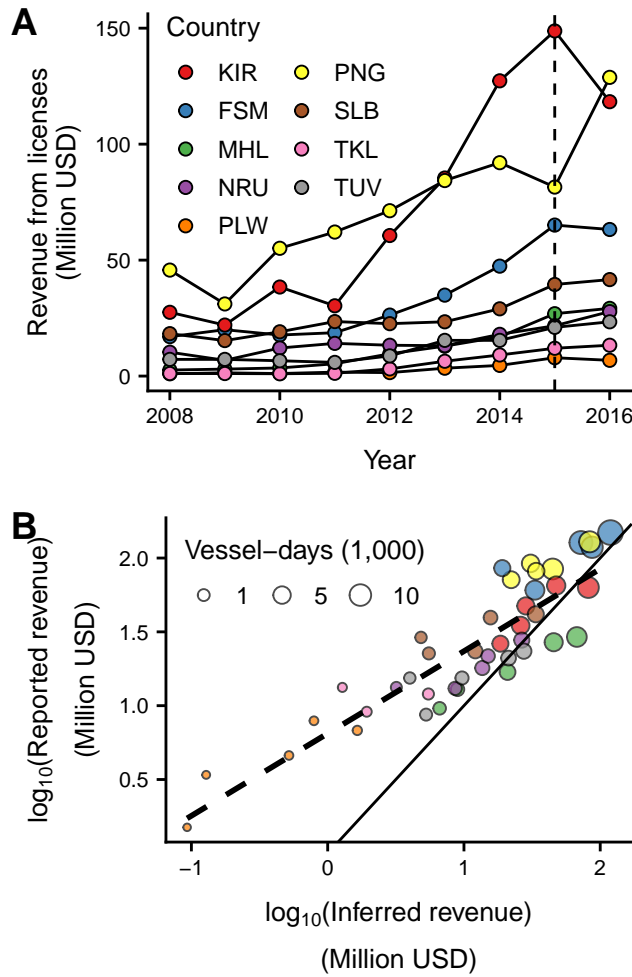


Figure 6: License revenue for PNA countries. A) Annual revenue from fishing license fees by country and year (2008 - 2016) B) \log_{10} -transformed FFA-reported revenues vs. the revenues inferred from vessel activity observations (2012 - 2016). The dashed line represents line of best fit, solid line represents 1:1 line. The same graph using absolute values is shown in S12.

5 Methods

5.1 Model for marine conservation with effort markets

We model a ten-patch discrete-time meta-population system, where Patch 1 is considering a spatial closure. Patches 1 - 9 operate under a vessel-day scheme, and Patch 10 represents the high seas and other areas not managed under a VDS. The stock of fish in each country is relatively stationary within a single fishing season, but redistributes across all patches annually. The price of fish is p , and catchability is given by q . These parameters are held constant across patches.

5.1.1 Fishery dynamics

In the absence of a reserve, the revenue for vessels in patch i is given by pqE_iX_i , where E_i and X_i are effort (vessel-days) and stock size in patch i at the beginning of a period. The cost of fishing in patch i is given by cE_i^β , where $\beta = 1.3$ matches commonly-used cost functions.

Patch 1 considers a spatial closure by implementing a reserve as a fraction R of the total patch ($R \in [0, 1)$). Fish move within a patch based on θ , where $\theta = 0$ implies no movement within the patch, and $\theta = 1$ implies that fish within the patch are well mixed during the fishing season. In this patch, revenues are given by $pqE_1X_1(\theta + (1 - \theta)(1 - R))$. The parameterization of movement and reserve size imply that profit from fishing Patch 1 is given by:

$$\Pi_1(E_1, X_1, R) = pqE_1X_1\Omega_i - cE_1^\beta$$

With $\Omega_i = (\theta + (1 - \theta)(1 - R))$ being a parameterization that combines reserve size as a proportion of patch ($R \in [0, 1)$) and within-patch fish movement (θ). Under this parameterization, $\Omega_{i \neq 1} = 1$ since only Patch 1 implements a reserve.

Profits from fishing in each Patch are:

$$\Pi_i(E_i, X_i) = pqE_iX_i\Omega_i - cE_i^\beta$$

The above equations imply that the marginal profit from the last unit of effort in a patch are given by:

$$\pi_1(E_1) = pqX_1\Omega_i - \beta cE_1^{\beta-1}$$

In practice, the effort levels in each Patch are allocated by management (so E_1, E_2, \dots, E_9 are given) and the effort level on the high seas (E_{10}) is a result of open access dynamics. Therefore, we assume that effort continues to enter Patch 10 until the profit from the last unit of effort is exactly zero, indicating that E_{10} is the value for which $\pi_{10}(E_{10}) = 0$. Setting Equation 5.1.1 for $i = 10$ equal to zero and removing $\Omega_{10} = 1$ for simplicity, we can solve for E_{10} :

$$E_{10} = \left(\frac{pqX_{10}}{\beta c} \right)^{\frac{1}{(\beta-1)}}$$

Under vds-operated patches, however, profits from marginal effort must equate the cost of fishing in the patch. Therefore vessel-day price for patches under vds ($i = (1, 9)$) is given by:

$$\pi(E_i) = pqX_i\Omega_i - \beta cE_i^{\beta-1}$$

We can solve for E_i and obtain:

$$\begin{aligned} \pi_i + \beta cE_i^{\beta-1} &= pqX_i\Omega_i \\ \beta cE_i^{\beta-1} &= pqX_i\Omega_i - \pi_i \\ E_i^{\beta-1} &= \frac{pqX_i\Omega_i - \pi_i}{\beta c} \\ E_i &= \left(\frac{pqX_i\Omega_i - \pi_i}{\beta c} \right)^{\frac{1}{\beta-1}} \end{aligned} \tag{1}$$

Therefore, total allowable effort in the fishery is given by:

$$\bar{E} = \sum_{i=1}^9 \left(\frac{pqX_i\Omega_i - \pi}{\beta c} \right)^{\frac{1}{\beta-1}}$$

5.1.2 Stock dynamics

Patch-level harvest is then determined by effort and stock size:

$$H_i = qE_iX_i\Omega_i$$

Therefore, escapement in patch i is the difference between initial stock size and harvests given by $e_{it} = X_{it} - H_{it}$ and total escapement is $e_t = \sum_{i=1}^{10} e_{it}$. The entire stock then grows logistically according to:

$$X_{t+1} = e_t + e^{r \times \frac{e}{K}}$$

Where r and K are species-specific intrinsic growth rate and carrying capacity.

After the stock grows, a constant and patch-specific fraction f_i of the total stock redistributes to patch i , so:

$$X_{it+1} = f_i X_{t+1}$$

5.1.3 Vessel-day revenues

The vessel-day price that a country charges is given by π_i from Eqn 5.1.1. Therefore, patch-level license revenues are given by:

$$\omega = \pi_i E_i$$

Equation 5.1.2 shows that low values of θ and $R > 0$ would increase escapement, which would lead to an increase in stock size (Equation 5.1.2) and a benefit to all the other patches. But this would also cause the stock in the high seas (X_{10}) to increase, leading to an increased effort being allocated to the high seas (Equation 5.1.1) and a loss of these potential rents. Thus, the spillover benefits of increasing R are never completely captured.

5.1.4 Simulations and parametrization

We inform our model to loosely match the fishery dynamics observed for the VDS operated by the PNA. The table below contains the values used to parameterize the model.

Parameter	Value	Source
MSY	1.875600e+06	50th percentile from MSY in table 8 of Stock Assessment
B_{msy}	1.628000e+06	50th percentile from MSY in table 8 of Stock Assessment
K	6.876526e+06	50th percentile from MSY in table 8 of Stock Assessment
B_c/B_{msy}	5.100000e-01	50th percentile from MSY in table 8 of Stock Assessment
C_{now}	1.679444e+06	Catches from Stock Assessment
B_{now}	3.507028e+06	Current Biomass (2012 - 2015 average)
r	5.700000e-01	From fishbase: Prior $r = 0.57$, 95 CL = 0.41 - 0.78
β	1.300000e+00	Standard
p	1.100000e+03	Mean between Thailand and Japan values (Value of WCPFC-CA tuna fisheries)
q	3.420000e-05	Estimated so that efforts match catches given biomass and vessel-day prices
c	1.800000e+02	-
f	1.000000e-01	Biomass is equally distributed between patches ($f_i = 0.1$)

We use the model above run three different simulations. We test the model across a range reserve sizes and movement parameters. We use two main scenarios. The first scenario does not allow trading. In this case, total allowable effort (\bar{E}) and biomass B_{now} are known and equally distributed among patches 1-9. For patch 10, we solve for Eq 5.1.1 until biomass converges to match B_{now} . We then proceed to “close” a portion of Patch 1, and calculate the vessel-day price in Patch 1 given that only $X_i\Omega_i$ biomass is available for harvest. We compare vessel-day revenues of each scenario to a case with no reserve ($R = 0$). This produces a measure of the cost of implementing a spatial closure of size R in patch 1.

The second scenario allows trading. We start again by solving for the high seas to obtain total effort. Since a closure is not in effect, this equilibrium is the same as the first step above. We then implement a spatial closure in patch 1. This essentially lowers the price fishers would be willing to pay to fish in a patch with biomass $X_i\Omega_i$, lowering demand for vessel-days in patch 1. Patches 2 - 9 have a higher demand for vessel days, and therefore a portion of vessel-days from patch 1 are sold to patches 2 - 9. This increases effort in these patches, which reduces escapement and therefore biomass. This reduction in biomass in turn would modify the demand curves (Eq 1). We therefore iterate this process until biomass stabilizes, therefore finding the system's equilibrium. Like before, we calculate vessel-day revenues to each patch and compare them to a case with no reserve in patch 1.

5.2 Allocation rules

In the second scenario above, effort from a patch with conservation is traded to other patches. This means that its allocation will decrease as purse seine effort in its EEZ is reduced. After solving for the new equilibrium for each combination of R and θ , we project the fishery 50 years in time. At the end of every time, vessel-days are allocated to each patch based on the following rule:

$$E_i^* = \alpha \left(\frac{\sum_{\tau=0}^{\hat{\tau}} E_{i,t-\tau}}{\bar{E}^{\hat{\tau}}} \right) + (1 - \alpha) \left(\frac{\sum_{\tau=0}^{\hat{\tau}} X_{i,t-\tau}}{\bar{X}^{\hat{\tau}}} \right)$$

Where α is a weight on historical effort (E_i) and $1 - \alpha$ is the weight in historical biomass (B_i). We use $\hat{\tau} = 7$ to obtain a moving mean of 7 years for these measures. The difference between allocated days (E_i^*) and used days (determined by Eq: 1) for patch 1 are the sales. We then calculate vessel-day revenues to each country over the 50-year time horizon and compare them to a case where there is no reserve and allocations are based solely on biomass ($\alpha = 0$).

5.3 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 algorithms (convolutional neural networks) on the broadcast messages to infer type and location of fishing events (14).

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 vessel activity. The addition of new satellites affects all vessels in the same way.

Our displaced group contains all purse seiners ($n = 64$) that fished within PIPA at least once before the announcement, and that continued to fish elsewhere after the January 2015 implementation. Vessels in the non-displaced group meet the following two conditions: i) never fished within PIPA waters from 2012-2015, and ii) vessels have fished in surrounding areas (*i.e.* PNA-countries' EEZ) before and after PIPA closure ($n = 28$). Together, these vessels represent more than 20 million geo-referenced positions for which we know activity (fishing or not fishing). We perform three sample restrictions as a robustness check. The first restriction excludes all Chinese vessels, the second excludes all PNA vessels, and the third excludes US and Taiwanese vessels.

5.4 Analysis

5.4.1 Crowding effect

$$y_t = \alpha + \beta_1 M_t + \beta_2 M_t^2 + \beta_3 M_t^3 + \beta_4 M_t^4 + \sigma_s + \mu N_t + \epsilon_t$$

We test for a crowding effect using the specification in Equation [5.4.1]. We have two different outcome variables: 1) the number of cells that had fishing activity from displaced and non-displaced vessels per month and 2) the correlation of presence/absence of fishing events between both groups over one month. We allow for the possibility of three inflection points: 1) initial crowding due to MPA implementation, 2) When the crowding has reached its peak and starts to decrease, and 3) when this decrease potentially levels off. For this reason, we fit a 4th degree polynomial to our monthly indices. We do so by centering our time series of crowding indices on the day of implementation. Our explanatory variable is therefore the number of months (M) before or after the implementation. For example, since PIPA was implemented in January 1st of 2015, December of 2014 has a value of -1 and Feb of 2015 would receive a value of 1. Note that we restrict the sample to our displaced and non-displaced vessels (vessels that show up in the dataset before PIPA implementation) to try to minimize bias from more and more vessels using AIS over time. We also include controls (σ_s) that captures the effect of additional satellites receiving AIS signals, which occurred on April 1st, 2014 and December 31st, 2015. The μ coefficient captures the effect of the NINO4 anomaly.

5.4.2 Behavioral changes

We attempt to identify the response of vessels to the PIPA closure. We use daily fishing and non-fishing hours, daily proportion of fishing vs. non-fishing hours, daily distance traveled (km), distance from shore (km) and distance from home port (km) for fishing events, and proportion of total fishing hours allocated to Kiribati waters and PNA waters as our main outcomes of interest. We compare these outcomes before and after the implementation of PIPA using a

Difference-in-Differences approach.

Our main specification is the following:

$$\log(y_{i,t}) = \alpha + \beta_1 P_t + \beta_2 D_i + \beta_3 P_t \times D_i + \phi_t + \gamma_i + \epsilon_{i,t}$$

where $\log(y_{i,t})$ is the hyperbolic-sine transformation² of the outcome of interest for vessel i on period t . A dummy variable P_t takes the value of 0 for all dates prior to PIPA implementation and a value of 1 for all dates following PIPA implementation. D_i is a dummy variable indicating whether a vessel belongs to the displaced ($D_i = 1$) or non-displaced ($D_i = 0$) group. α is the standard intercept term, β_1 captures the temporal trend, β_2 captures the initial difference between displaced and non-displaced groups, and β_3 is our parameter of interest: the difference-in-differences estimate capturing the treatment effect. Finally, ϕ_t and γ_i represent month and flag dummies that account for seasonality or country-level management interventions.

All regression coefficients were estimated via ordinary least squares, and heteroskedasticity-robust standard errors were calculated. All analyses were performed in R version 3.5.1 (19). Raw data and code used in this work are available on github.

5.4.3 Revenues and catches

We obtained information on revenues from the Pacific Islands Forum Fisheries Agency *Tuna Development Indicators 2016* report. Specifically, we use data compiled by the Pacific Islands Forum Fisheries Agency (FFA³) where annual revenues from license fees (for VDS and other access programs) are reported for each country (2008 - 2016; Fig. 6A). For countries in the PNA, these revenues show a combination of vessel-day license fees as well as joint-venture operations.

² $\ln(y + \sqrt{1 + y^2}) \rightarrow \ln(2y)$ the transformation was not applied to the proportion of fishing to non-fishing hours.

³<https://www.ffa.int/node/2050>

Total purse seine catch for each country's EEZ for the 1997 - 2016 period were also obtained from the FFA (Fig. S14). Catches in Kiribati waters decreased from 24,051 to 12,894 tonnes between 2015 and 2016 (46.3% decrease). Similar decreases were observed for The Federated States of Micronesia (60.9%), Papua New Guinea (43.4%) and the Solomon Islands (58.5%). In contrast, Tokelau (due south of PIPA) showed a 22.3% increase in purse seine catch over the same period.

6 Supplementary tables and figures

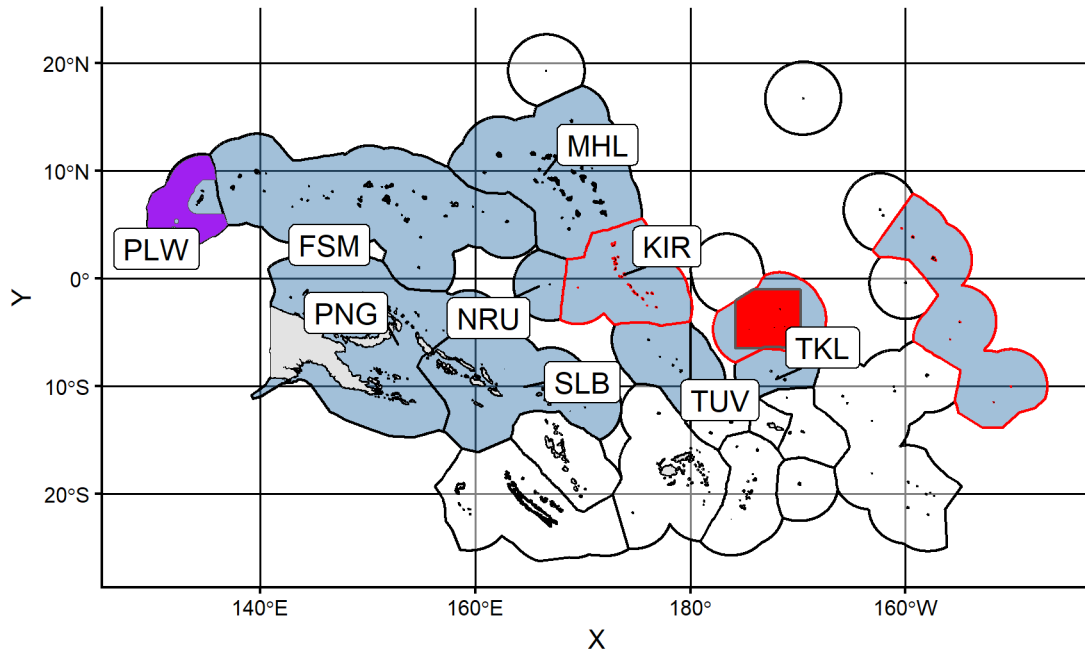


Figure S1: Map of the Exclusive Economic Zones (EEZs) of the region of interest. Countries that belong to the PNA are shown in blue, while empty polygons indicate all others. A red line indicates the Kiribati EEZ, and a solid red polygon shows PIPA. A solid purple polygon shows PNMS. Land masses are shown in gray. Labels indicate ISO3 country codes (PLW: Palau, PNG: Papua New Guinea, FSM: Federal States of Micronesia, SLB: Solomon Islands, NRU: Nauru, MHL: Marshall Islands, KIR: Kiribati, TUV: Tuvalu, TKL: Tokelau).

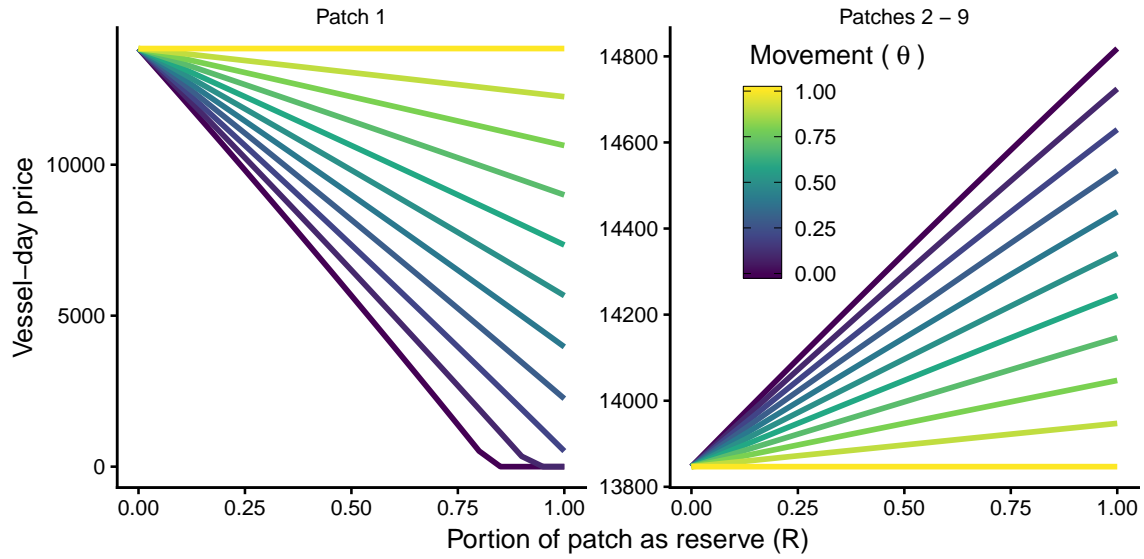


Figure S2: Vessel-day prices (vertical axis) for a combination of reserve sizes (R in the horizontal-axis) and different within-patch movement (θ) for the patch with spatial closure and other patches (left - right, respectively) when there is no trading.

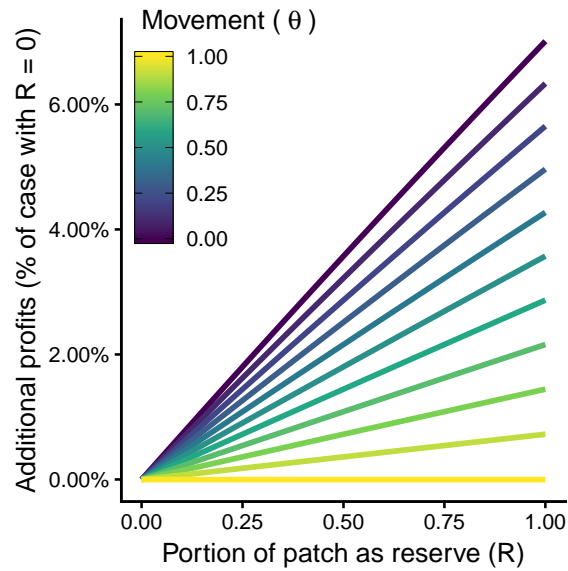


Figure S3: Relative change in revenue for patches 2 - 9 (vertical axis) for a combination of reserve sizes (R in the horizontal-axis) and different within-patch movement (θ) when there is no trading.

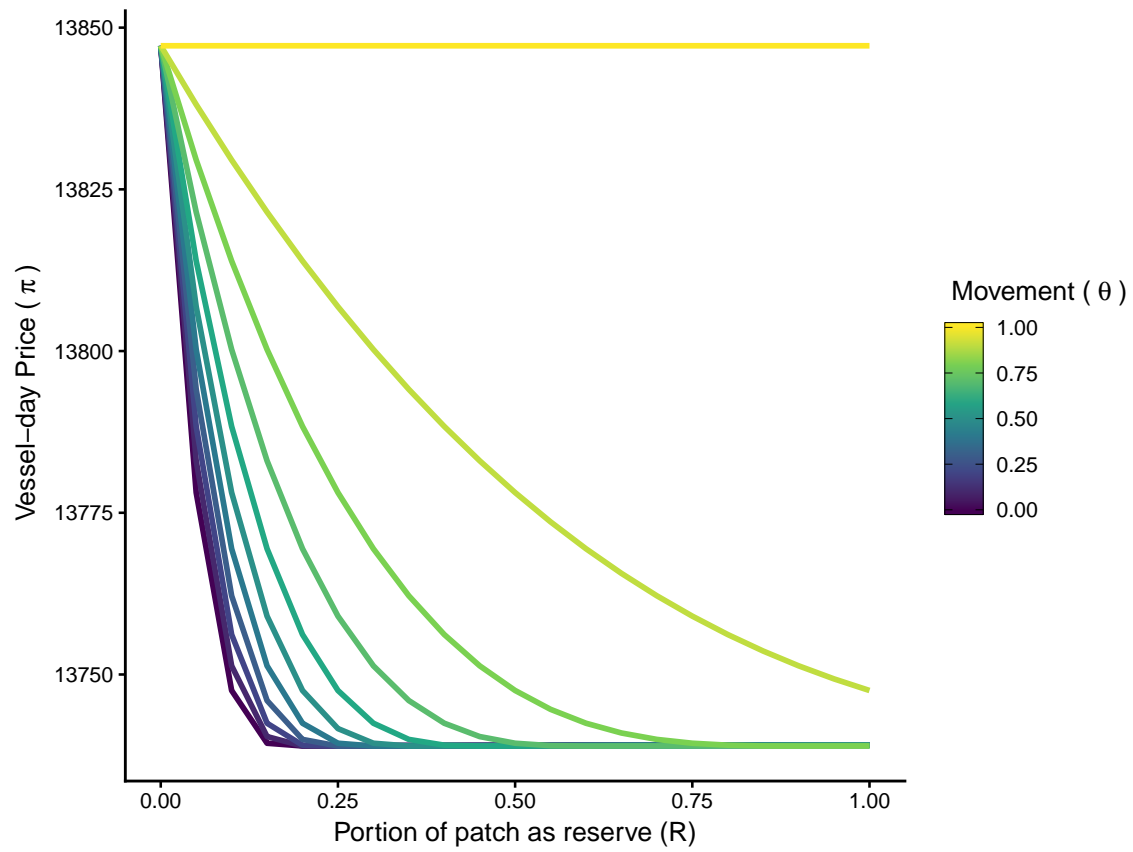


Figure S4: Vessel-day prices (vertical axis) for a combination of reserve sizes (R in the horizontal-axis) and different within-patch movement (θ) for the patch with spatial closure and other patches (left - right, respectively) when there is no trading.

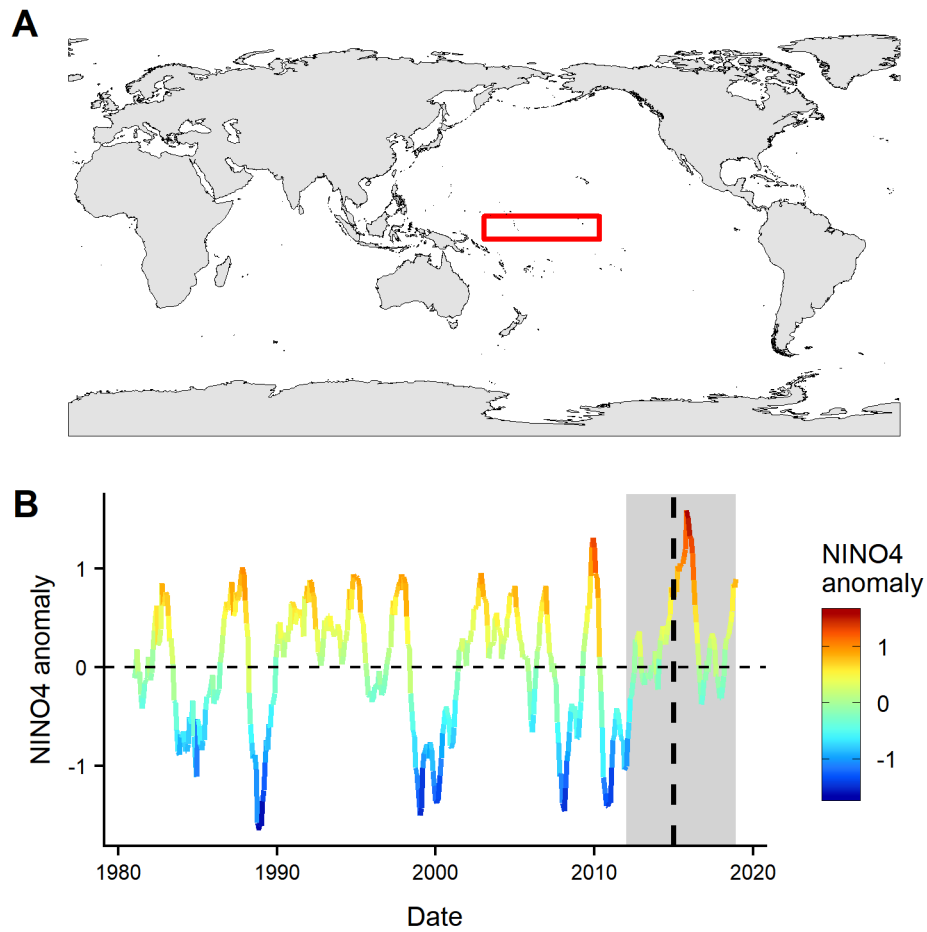


Figure S5: NINO4 anomaly index. A) Map of the NINO4 region (5S-5N and 160E-150W). B) Timeseries of NINO4 anomaly from January, 1980 to December, 2018.

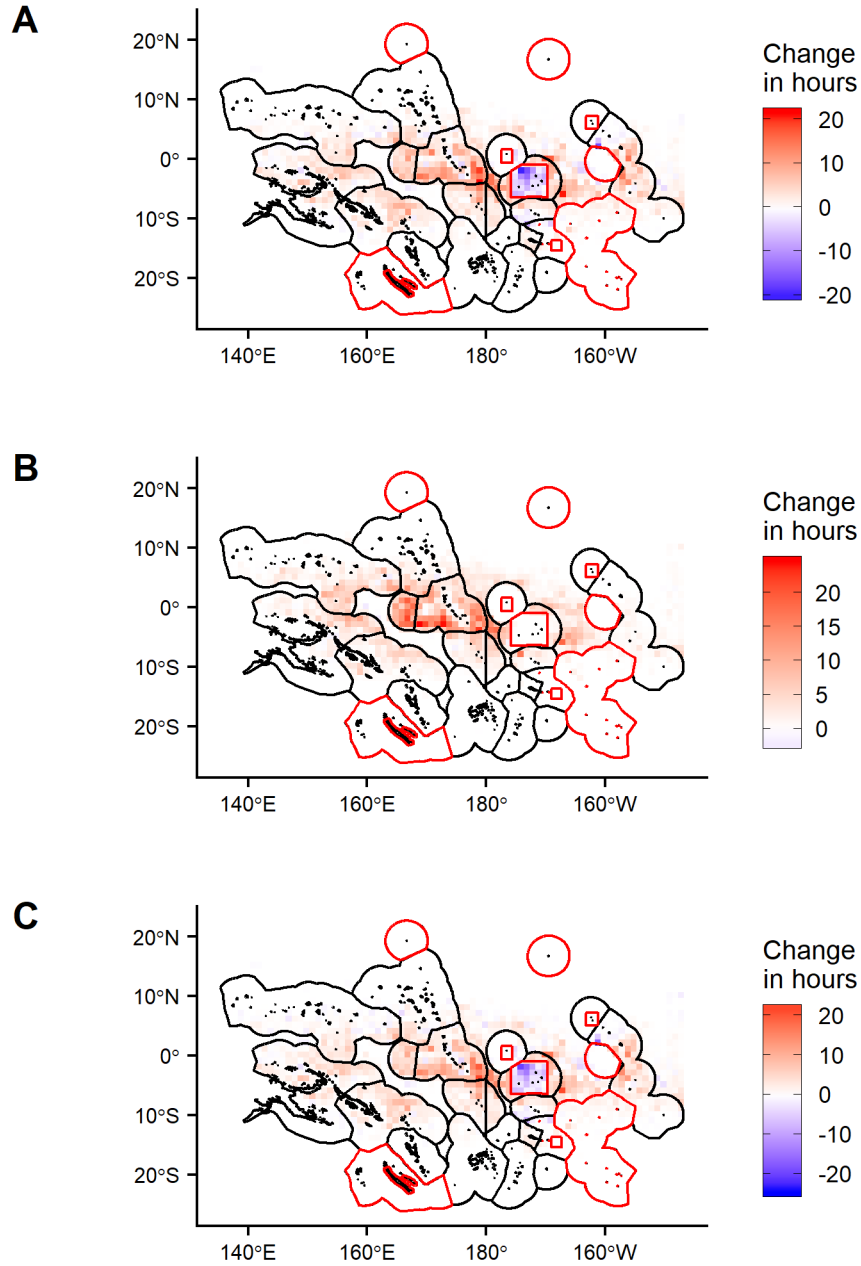


Figure S6: Change in spatial footprint of analyzed vessels. Black lines show Exclusive Economic Zone (EEZ), red lines show existing Marine Protected Areas. Panels A and B show the change through time (after - before) for displaced (A) and non-displaced vessels (B). Panel C shows the difference between A and B (displaced - non-displaced), highlighting areas where displaced vessels redistributed to, relative to non-displaced vessels. Note that displaced vessels allocate more hours to the Gilbert Islands and Line islands EEZs, but also Tuvalu and the high seas surrounding PIPA and Kiribati's EEZ.

Table S1: Coefficient estimates for a fourth-degree polynomial fit to the measures of crowding for all PNA waters. The first five columns represent different specifications for number of cells with presence of both fleets. Columns 6 - 10 show coefficients for the spatial correlation for presence / absence of displaced and non-displaced vessels. The explanatory variable is the number of months before or after implementation of PIPA. Numbers in parentheses are heteroskedastic-robust standard errors. The last column of each group presents fits with only NINO4 anomaly index as an explanatory variable.

	Number of cells					Pearson's correlation coefficient				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Constant	78.04*** (5.44)	84.64*** (8.99)	60.96*** (15.60)	66.73*** (15.99)	80.62*** (7.96)	0.41*** (0.01)	0.42*** (0.03)	0.37*** (0.06)	0.38*** (0.07)	0.38*** (0.02)
M	3.94*** (0.30)	4.07*** (0.35)	3.07*** (0.97)	3.14*** (1.04)		0.01*** (0.001)	0.01*** (0.001)	0.01** (0.003)	0.01** (0.004)	
M ²	-0.005 (0.02)	-0.02 (0.03)	0.01 (0.03)	-0.01 (0.03)		-0.0001*** (0.0000)	-0.0001 (0.0001)	-0.0001 (0.0001)	-0.0001 (0.0001)	
M ³	-0.002*** (0.0003)	-0.002*** (0.0003)	-0.002*** (0.001)	-0.002*** (0.001)		-0.0000*** (0.0000)	-0.0000*** (0.0000)	-0.0000*** (0.0000)	-0.0000*** (0.0000)	
M ⁴	0.0000 (0.0000)	0.0000 (0.0000)	0.0000 (0.0000)	0.0000 (0.0000)		0.0000*** (0.0000)	0.0000** (0.0000)	0.0000 (0.0000)	0.0000 (0.0000)	
NINO4		-8.09 (8.29)		-10.39 (9.48)	19.76** (9.41)		-0.01 (0.03)		-0.01 (0.03)	0.07*** (0.02)
σ_1			21.32 (19.49)	25.10 (22.49)				0.06 (0.08)	0.06 (0.08)	
σ_2			5.30 (18.87)	3.19 (18.67)				-0.01 (0.04)	-0.02 (0.03)	
NINO4	No	Yes	No	Yes	Yes	No	Yes	No	Yes	Yes
Satellites	No	No	Yes	Yes	No					
AIC	796.937	797.991	799.477	799.97	919.583	-187.208	-185.269	-184.916	-183.233	-97.386
Observations	84	84	84	84	84	84	84	84	84	84
R ²	0.79	0.79	0.79	0.80	0.03	0.70	0.70	0.71	0.71	0.07

Note:

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

Table S2: Coefficient estimates for a fourth-degree polynomial fit to the measures of crowding for Kiribati EEZ only. The first five columns represent different specifications for number of cells with presence of both fleets. Columns 6 - 10 show coefficients for the spatial correlation for presence / absence of displaced and non-displaced vessels. The explanatory variable is the number of months before or after implementation of PIPA. Numbers in parentheses are heteroskedastic-robust standard errors. The last column of each group presents fits with only NINO4 anomaly index as an explanatory variable.

	Number of cells					Pearson's correlation coefficient				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Constant	30.24*** (2.43)	32.77*** (4.16)	23.30*** (4.59)	26.14*** (5.49)	16.23*** (2.37)	0.43*** (0.03)	0.43*** (0.05)	0.33*** (0.07)	0.34*** (0.08)	0.38*** (0.03)
M	1.29*** (0.14)	1.34*** (0.17)	1.00*** (0.27)	1.06*** (0.30)		0.01*** (0.002)	0.01*** (0.002)	0.01* (0.01)	0.01 (0.01)	
M ²	-0.03*** (0.01)	-0.04*** (0.01)	-0.02** (0.01)	-0.03*** (0.01)		0.0000 (0.0002)	0.0000 (0.0002)	0.0001 (0.0002)	0.0001 (0.0002)	
M ³	-0.001*** (0.0001)	-0.001*** (0.0001)	-0.001*** (0.0002)	-0.001*** (0.0002)		-0.0000*** (0.0000)	-0.0000*** (0.0000)	-0.0000** (0.0000)	-0.0000** (0.0000)	
M ⁴	0.0000*** (0.0000)	0.0000*** (0.0000)	0.0000** (0.0000)	0.0000*** (0.0000)		-0.0000 (0.0000)	-0.0000 (0.0000)	-0.0000 (0.0000)	-0.0000 (0.0000)	
NINO4		-3.10 (3.97)		-4.19 (4.16)	9.93*** (3.33)		-0.001 (0.05)		-0.02 (0.06)	0.11*** (0.03)
σ_1			9.40 (5.92)	10.33 (6.35)				0.12 (0.09)	0.13 (0.09)	
σ_2			-2.46 (5.70)	-3.84 (5.77)				0.02 (0.10)	0.01 (0.11)	
NINO4	No	Yes	No	Yes	Yes	No	Yes	No	Yes	Yes
Satellites	No	No	Yes	Yes	No					
AIC	654.294	655.536	655.482	656.125	724.58	-65.872	-63.873	-63.481	-61.572	-35.895
Observations	84	84	84	84	84	75	75	75	75	75
R ²	0.63	0.63	0.64	0.65	0.08	0.44	0.44	0.45	0.45	0.09

Note:

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

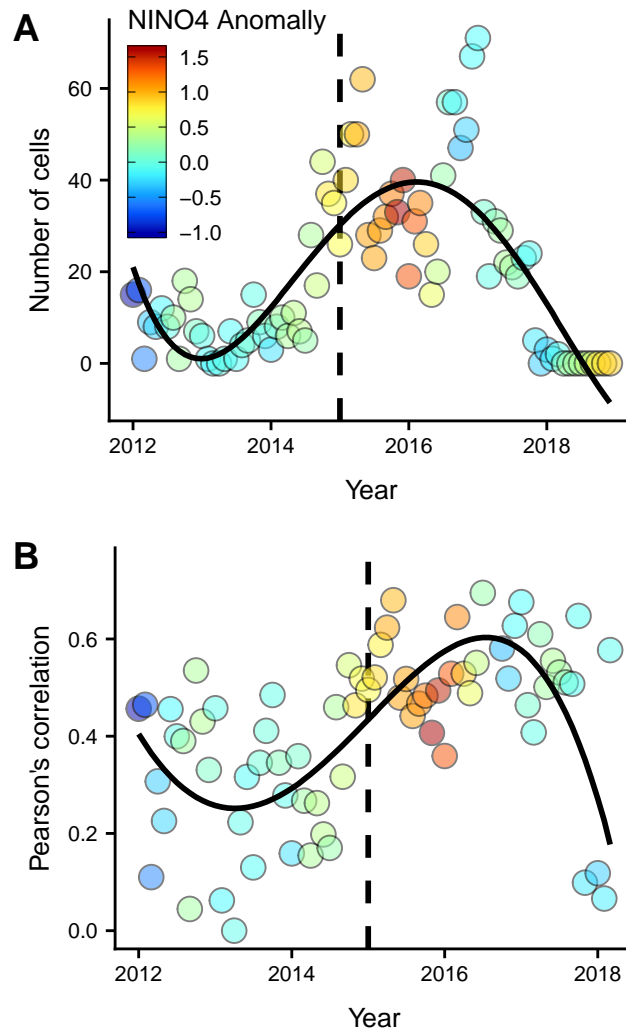


Figure S7: Number of cells that had displaced and non-displaced vessels (A) and spatial correlation in the presence-absence of each group per cell (B) for Kiribati's EEZ only. The solid lines represent the 4th degree polynomial fit reported in S1. Note that the late 2016 and early 2017 showed negative or neutral NINO4 anomalies, similar to those in the pre-PIPA period.

Table S3: Difference-in-differences estimates for our 9 variables of interest: 1) Daily fishing hours, 2) Daily non-fishing at-sea hours, 3) Daily proportion of fishing hours to total at-sea hours, 4) Daily distance traveled, 5) Daily mean distance from port for fishing events, 6) Daily mean distance from shore for fishing events, 7) Monthly fishing hours spent in Kiribati waters, 8) Monthly fishing hours spent in PNA waters, and 9) Monthly fishing hours in the high seas. Numbers in parentheses are heteroskedastic-robust standard errors.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Constant	0.495*** (0.022)	3.607*** (0.012)	0.075*** (0.004)	4.440*** (0.042)	12.998*** (0.021)	12.462*** (0.019)	3.709*** (0.195)	4.456*** (0.151)	2.429*** (0.415)
Post	0.846*** (0.018)	−0.227*** (0.009)	0.138*** (0.003)	0.112*** (0.031)	0.271*** (0.014)	0.275*** (0.014)	0.943*** (0.141)	1.129*** (0.110)	0.709** (0.284)
Displaced	0.136*** (0.013)	0.014** (0.007)	0.015*** (0.002)	0.255*** (0.029)	0.225*** (0.016)	0.117*** (0.016)	0.549*** (0.148)	0.153 (0.118)	−0.280 (0.236)
NINO4	−0.014 (0.011)	−0.001 (0.005)	−0.001 (0.002)	−0.411*** (0.017)	0.167*** (0.008)	0.064*** (0.007)	0.357*** (0.068)	0.137** (0.056)	0.484*** (0.122)
Post × Displaced	−0.243*** (0.019)	0.013 (0.009)	−0.034*** (0.003)	−0.210*** (0.036)	−0.285*** (0.017)	−0.157*** (0.017)	−0.586*** (0.161)	−0.403*** (0.127)	0.338 (0.285)
Month FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Flag FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	83,052	83,052	83,051	79,669	32,055	32,055	1,814	2,588	684
R ²	0.102	0.072	0.107	0.017	0.075	0.082	0.126	0.200	0.252

Note:

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

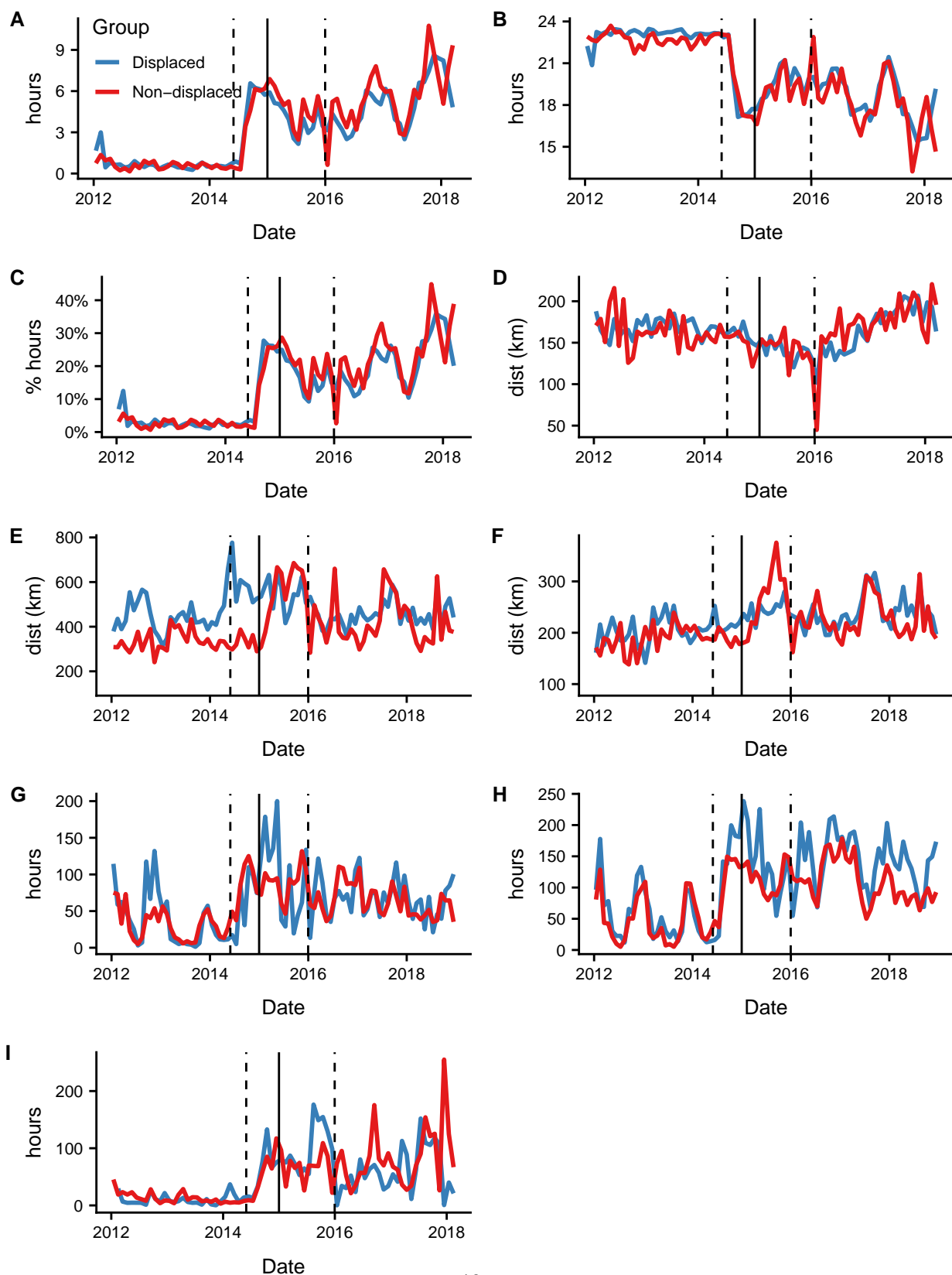


Figure S8: Time series showing monthly averages for our nine variables of interest: A) Fishing hours, B) Non-fishing hours at-sea, C) Proportion of fishing hours to total hours at-sea, D) Distance traveled, E) Mean distance from port for fishing events, F) Mean distance from shore

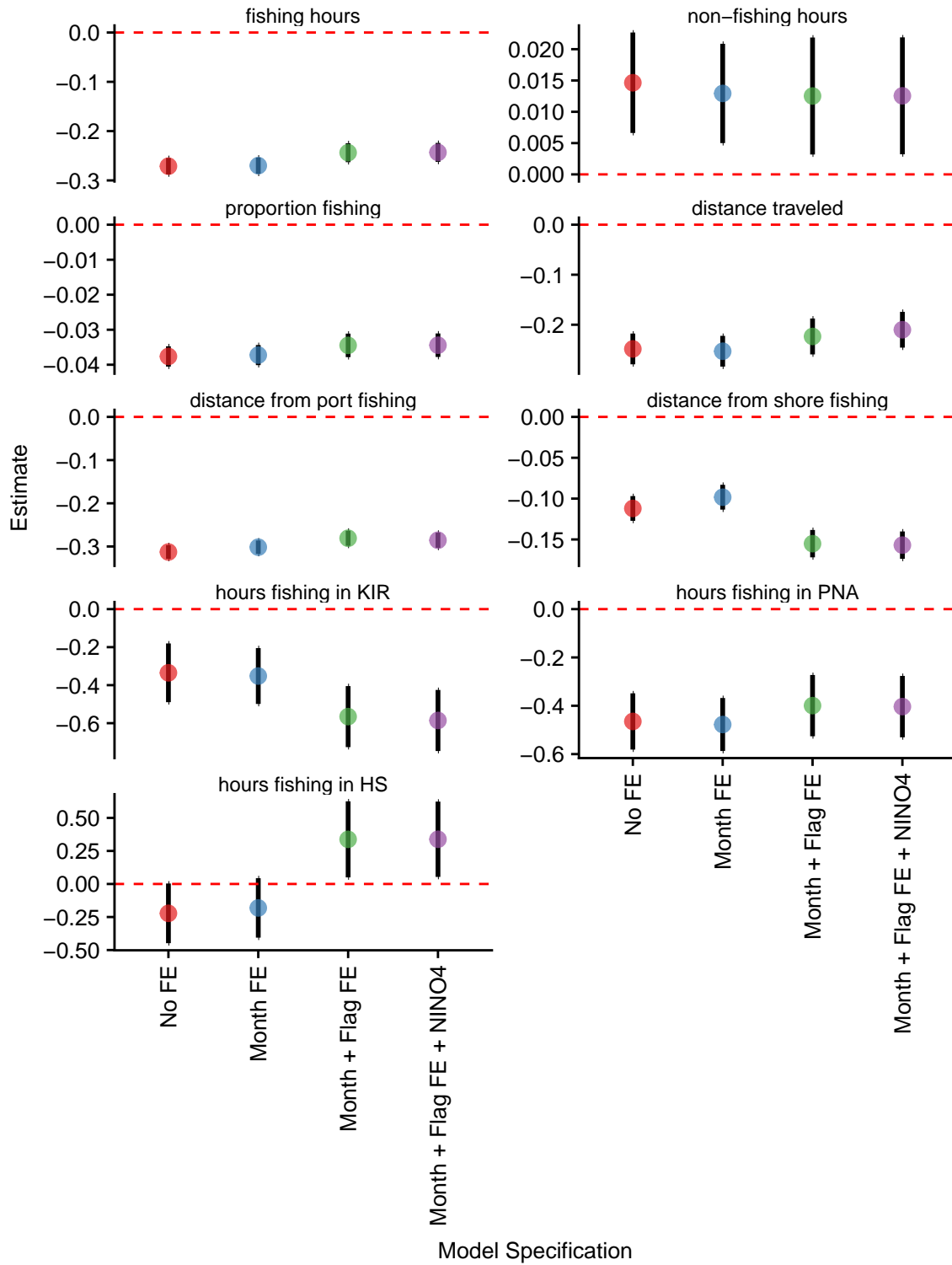


Figure S9: Alternative difference-in-differences estimates for our variables of interest using different model specifications. Table S3 reports estimates for models with month and flag fixed effects, and NINO4 index (*i.e.* green dots).

Table S4: Difference-in-differences estimates for our 9 variables of interest after removing Chinese vessels. 1) Daily fishing hours, 2) Daily non-fishing at-sea hours, 3) Daily proportion of fishing hours to total at-sea hours, 4) Daily distance traveled, 5) Daily mean distance from port for fishing events, 6) Daily mean distance from shore for fishing events, 7) Monthly fishing hours spent in Kiribati waters, 8) Monthly fishing hours spent in PNA waters, and 9) Monthly fishing hours in the high seas. Numbers in parentheses are heteroskedastic-robust standard errors.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Constant	0.058*** (0.019)	3.863*** (0.008)	−0.003 (0.003)	4.829*** (0.044)	13.817*** (0.044)	13.161*** (0.058)	4.007*** (0.347)	4.515*** (0.304)	2.909*** (0.274)
Post	0.824*** (0.019)	−0.254*** (0.010)	0.136*** (0.003)	0.145*** (0.034)	0.306*** (0.016)	0.322*** (0.016)	0.920*** (0.156)	1.157*** (0.121)	0.691** (0.291)
Displaced	0.108*** (0.013)	0.009 (0.007)	0.012*** (0.002)	0.270*** (0.031)	0.271*** (0.017)	0.158*** (0.017)	0.491*** (0.162)	0.150 (0.131)	−0.274 (0.235)
NINO4	−0.012 (0.011)	−0.007 (0.005)	−0.001 (0.002)	−0.380*** (0.018)	0.164*** (0.009)	0.061*** (0.008)	0.365*** (0.074)	0.122** (0.060)	0.464*** (0.126)
Post × Displaced	−0.212*** (0.021)	0.040*** (0.010)	−0.031*** (0.004)	−0.282*** (0.038)	−0.338*** (0.019)	−0.205*** (0.018)	−0.558*** (0.174)	−0.412*** (0.137)	0.374 (0.289)
Month FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Flag FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	75,327	75,327	75,326	75,390	28,449	28,449	1,570	2,279	633
R ²	0.102	0.073	0.108	0.017	0.075	0.091	0.128	0.209	0.266

Note:

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

Table S5: Difference-in-differences estimates for our 9 variables of interest after removing PNA vessels: 1) Daily fishing hours, 2) Daily non-fishing at-sea hours, 3) Daily proportion of fishing hours to total at-sea hours, 4) Daily distance traveled, 5) Daily mean distance from port for fishing events, 6) Daily mean distance from shore for fishing events, 7) Monthly fishing hours spent in Kiribati waters, 8) Monthly fishing hours spent in PNA waters, and 9) Monthly fishing hours in the high seas. Numbers in parentheses are heteroskedastic-robust standard errors.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Constant	0.512*** (0.024)	3.559*** (0.013)	0.083*** (0.004)	4.499*** (0.047)	13.165*** (0.028)	12.667*** (0.025)	3.271*** (0.263)	4.066*** (0.201)	2.876*** (0.434)
Post	0.781*** (0.022)	-0.159*** (0.012)	0.122*** (0.004)	0.047 (0.043)	0.114*** (0.023)	0.070*** (0.022)	1.168*** (0.226)	1.481*** (0.180)	0.295 (0.333)
Displaced	0.202*** (0.015)	0.040*** (0.009)	0.019*** (0.003)	0.437*** (0.037)	0.165*** (0.024)	-0.015 (0.022)	0.817*** (0.229)	0.533*** (0.182)	-0.408* (0.235)
NINO4	-0.019 (0.012)	-0.0003 (0.006)	-0.002 (0.002)	-0.387*** (0.020)	0.138*** (0.010)	0.025*** (0.009)	0.413*** (0.080)	0.232*** (0.066)	0.482*** (0.156)
Post × Displaced	-0.219*** (0.024)	-0.055*** (0.012)	-0.023*** (0.004)	-0.246*** (0.046)	-0.175*** (0.026)	0.010 (0.024)	-0.843*** (0.243)	-0.821*** (0.195)	0.729** (0.341)
Month FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Flag FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	64,560	64,560	64,559	64,625	22,654	22,654	1,366	1,928	511
R ²	0.093	0.069	0.099	0.022	0.063	0.066	0.127	0.203	0.214

Note:

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

Table S6: Difference-in-differences estimates for our 9 variables of interest after removing US and Tawianese vessels. 1) Daily fishing hours, 2) Daily non-fishing at-sea hours, 3) Daily proportion of fishing hours to total at-sea hours, 4) Daily distance traveled, 5) Daily mean distance from port for fishing events, 6) Daily mean distance from shore for fishing events, 7) Monthly fishing hours spent in Kiribati waters, 8) Monthly fishing hours spent in PNA waters, and 9) Monthly fishing hours in the high seas. Numbers in parentheses are heteroskedastic-robust standard errors.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Constant	0.536*** (0.023)	3.600*** (0.012)	0.082*** (0.004)	4.506*** (0.043)	13.002*** (0.022)	12.438*** (0.020)	3.850*** (0.209)	4.719*** (0.158)	2.420*** (0.419)
Post	0.796*** (0.019)	-0.217*** (0.010)	0.130*** (0.003)	0.021 (0.035)	0.290*** (0.016)	0.291*** (0.016)	0.870*** (0.156)	0.894*** (0.121)	0.732** (0.291)
Displaced	0.142*** (0.013)	0.016** (0.007)	0.015*** (0.002)	0.341*** (0.031)	0.227*** (0.018)	0.127*** (0.017)	0.490*** (0.163)	-0.017 (0.126)	-0.296 (0.239)
NINO4	-0.001 (0.011)	-0.001 (0.006)	0.001 (0.002)	-0.383*** (0.019)	0.189*** (0.009)	0.082*** (0.008)	0.325*** (0.075)	0.171*** (0.063)	0.441*** (0.122)
Post × Displaced	-0.212*** (0.021)	-0.002 (0.010)	-0.029*** (0.004)	-0.158*** (0.039)	-0.328*** (0.019)	-0.184*** (0.018)	-0.533*** (0.175)	-0.225 (0.138)	0.339 (0.291)
Month FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Flag FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	73,717	73,717	73,716	73,778	26,920	26,920	1,546	2,236	660
R ²	0.095	0.072	0.102	0.021	0.077	0.094	0.111	0.169	0.256

Note:

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

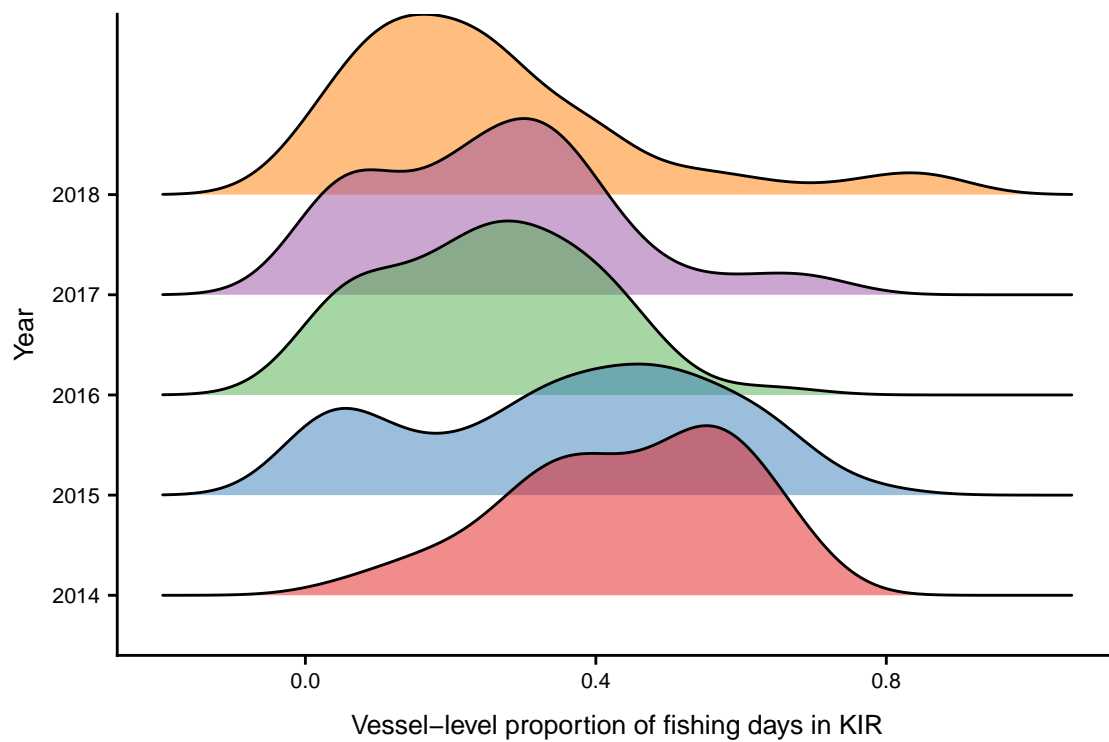


Figure S10: Ridgeplot for the density of the % of total fishing hours that take place within Kiribati EEZ waters by year for displaced vessels where the unit of observation is an individual vessel.

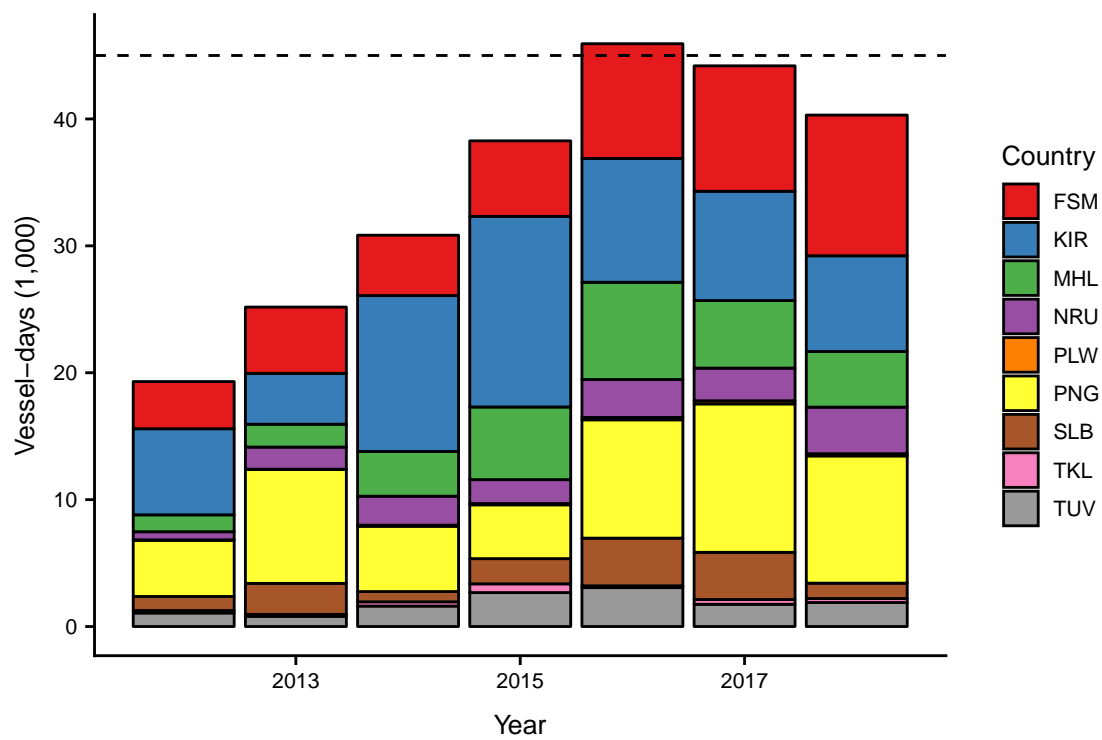


Figure S11: Annual vessel-days for all PNA countries, by country.

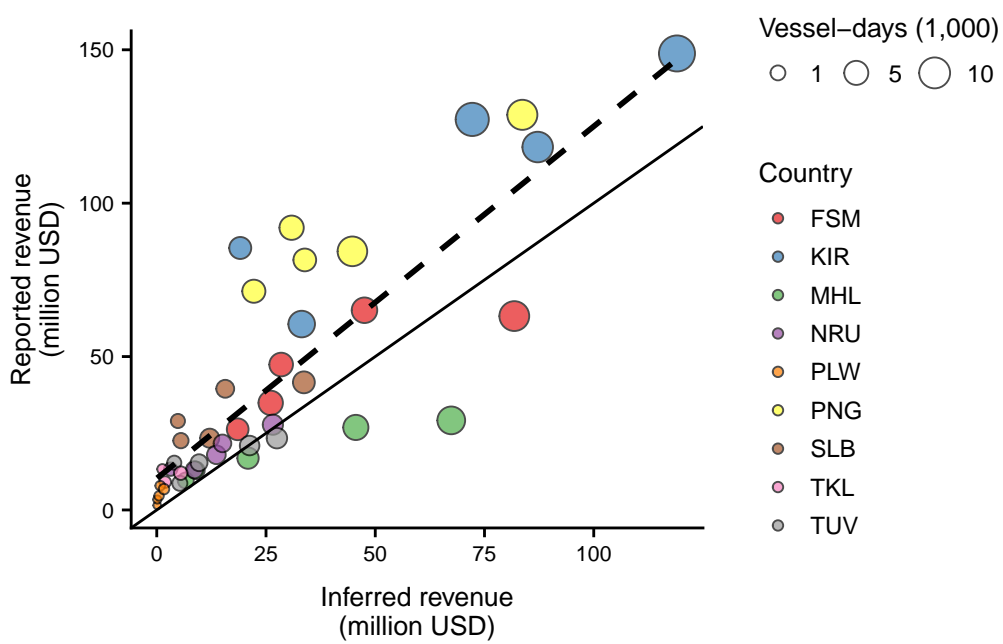


Figure S12: Inferred revenues vs. reported revenues. The dashed line represents line of best fit, and the solid line represents a 1:1 line.

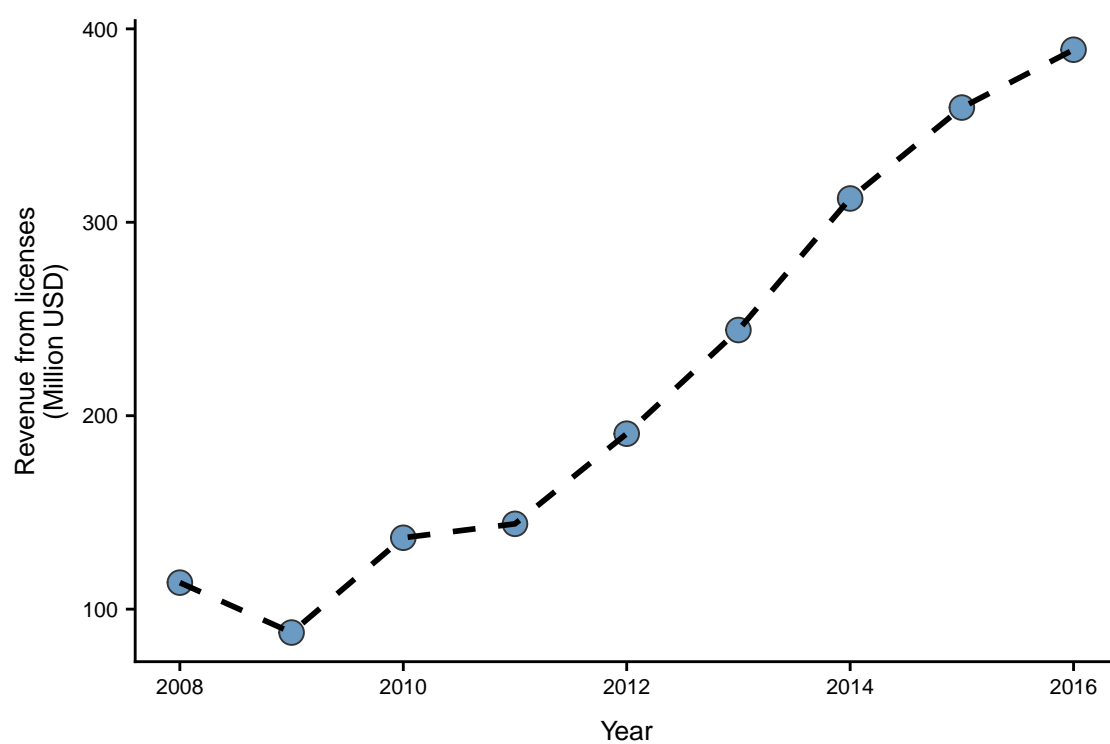


Figure S13: Total revenues for all PNA countries combined.

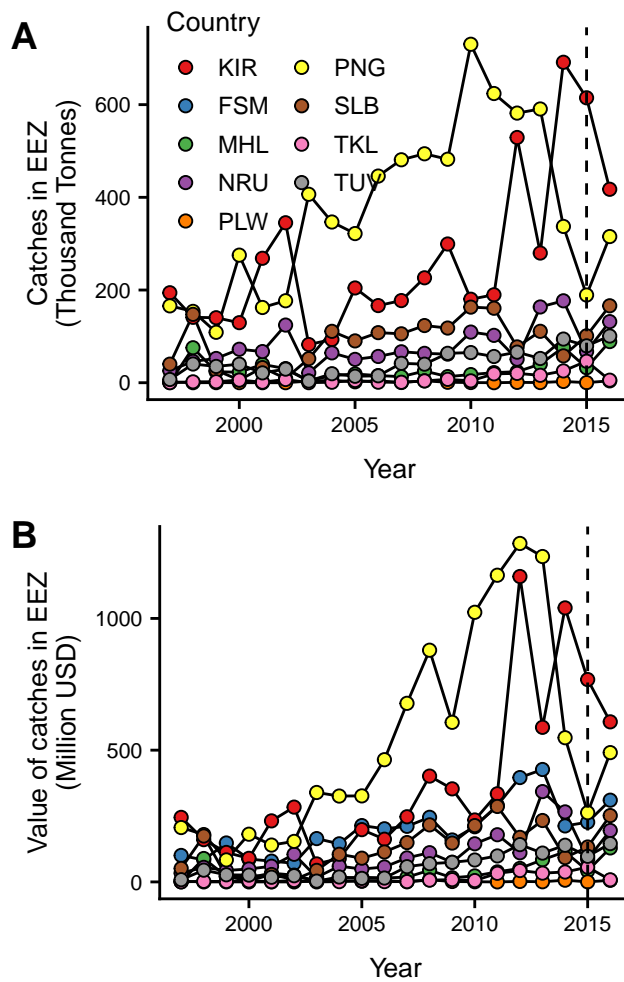


Figure S14: Financial indicators for PNA countries. A) Total annual purse seine catch by EEZ and, B) Total annual value of purse seine catch by EEZ. Vertical dashed line in both plots denotes implementation of PIPA.

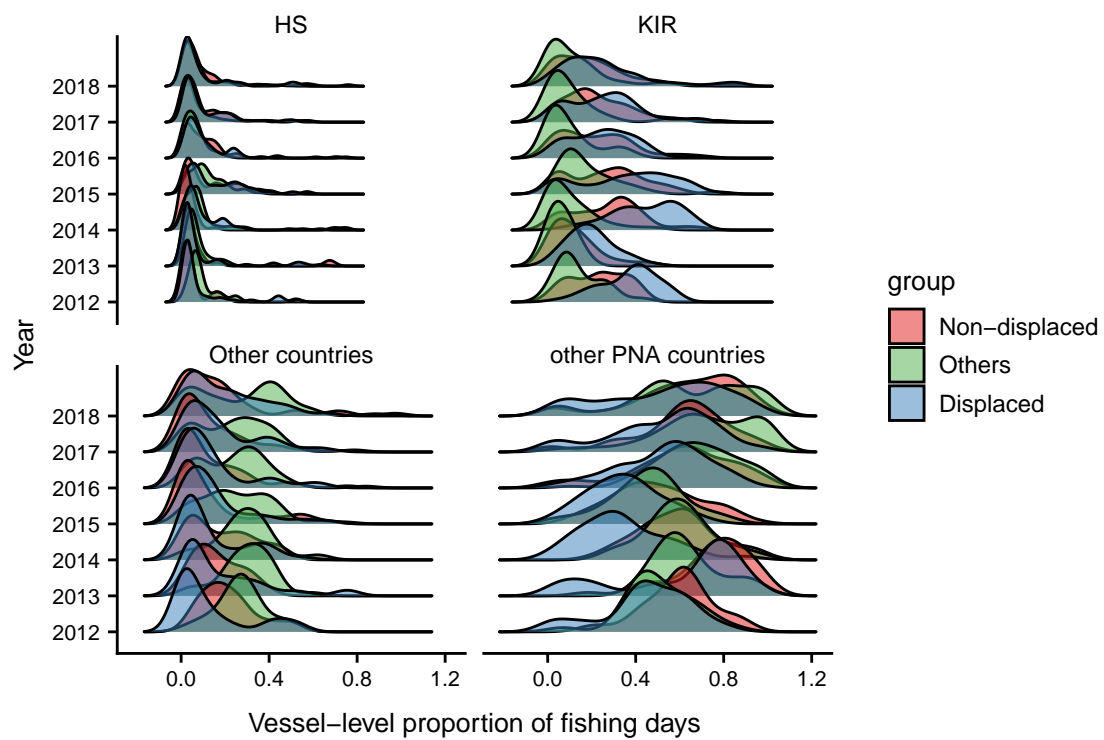


Figure S15: Ridgeplot for the density of the % of total fishing hours that take place in each region for all vessels

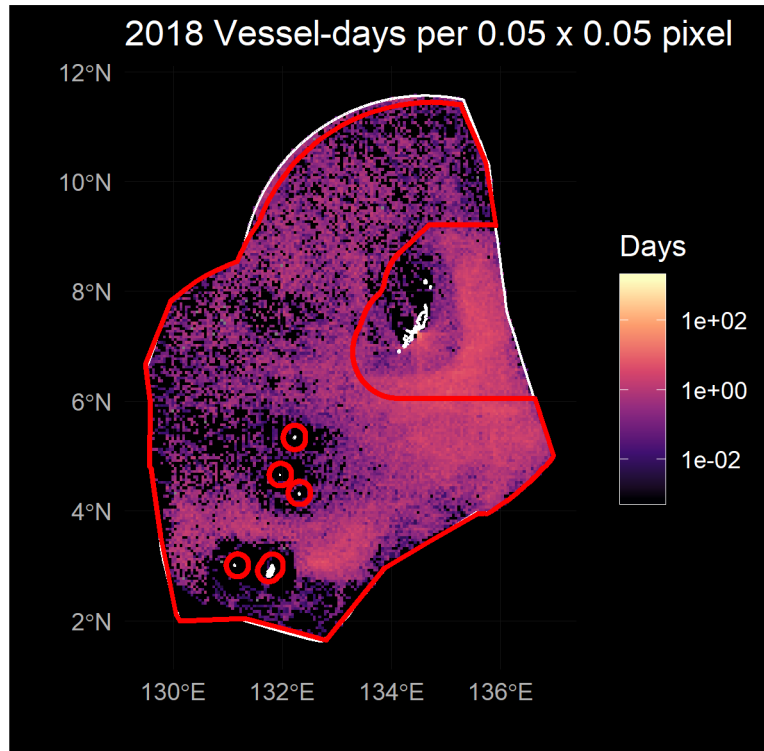


Figure S16: Longline and purse seine fishing effort in Palaw during 2018 at a 0.5 degree resolution. The red polygon shows the Palau National Marine Sanctuary.

I used this figure at the workshop last week, I do not intend to use it with the current design. I want to have a normal white background and split it into 2 (longliners and purse seiners) to better capture the effort that would be displaced by PNMS. I might also break it by year (2015 - 2018) to show that patterns are consistent through time.

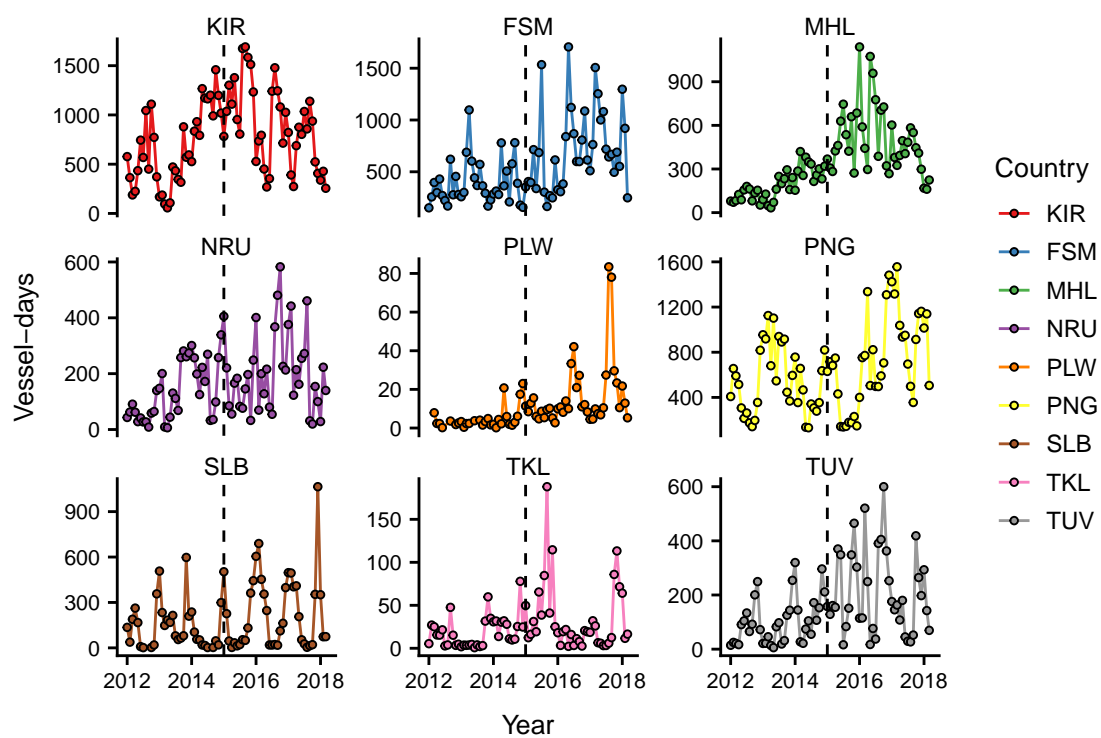


Figure S17: Time series of monthly observed vessel-days for each PNA country.

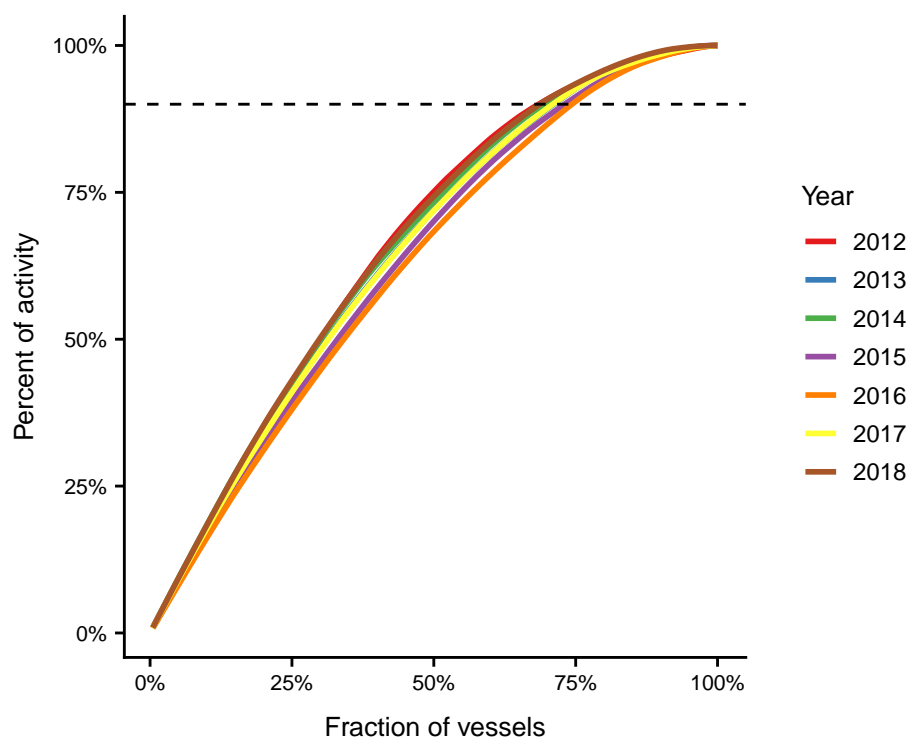


Figure S18: Cumulative activity (% of total hours) as a function of increasing fraction of vessels (% of total) by year.)