

## Article

# Environmental Characteristics Associated with the Presence of the Pelagic Stingray (*Pteroplatytrygon violacea*) in the Pacific High Sea

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**Abstract:** The pelagic stingray (*Pteroplatytrygon violacea*), perhaps the only stingray to inhabit open ocean waters, is highly interactive with longline and purse seine fisheries. The threat to *P. violacea* posed by high bycatch mortality has received widespread attention. To date, the environmental preference of *P. violacea*, which is important in designing conservation and management measures, has not been well studied. Based on data collected during a 2016–2019 survey in the Pacific Ocean by national observers of tuna longline fisheries, the relationship between the presence of *P. violacea* and spatiotemporal and environmental variables was first analyzed using the Generalized Additive Model. The results showed that geographic location (latitude and longitude) was the most influential variable. Monthly, *P. violacea* is frequently present in the Pacific high sea from December to May. The El Niño–Southern Oscillation had a significant impact on the presence of *P. violacea* in the Pacific high sea, with both the cold (Ocean Niño Index <−0.5) and warm (Ocean Niño Index >1) phases leading to a decrease in its presence. Regarding the environmental factors, we found that high presence was associated with low salinity (33.0–34.5 psu), a relatively high concentration of chlorophyll (0.2–0.35 mg/m<sup>3</sup>), and warm water (>20 °C). *P. violacea* was most likely observed in the waters offshore, closer to seamounts, and with water depths between 4000 and 5000 m. Four areas, including those east of the Solomon Islands and east of Kiribati, areas west of the Galapagos Islands, and areas near the coastal upwelling of northern Peru, related to upwelling systems or seamounts, were identified as the potential key habitats of *P. violacea*. Predicted distribution maps showed a significant seasonal variation in the presence of *P. violacea*. Moreover, the yearly change in the presence of *P. violacea* in the Pacific high sea indicated a possible decreasing trend in recent years. The information first provided here is essential for developing conservation and management measures for *P. violacea* to prevent the unavoidable ecological consequences of bycatch or other anthropogenic factors.

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**Keywords:** *Pteroplatytrygon violacea*; environmental preference; bycatch; habitat hotspots; tuna longline

## 1. Introduction

Incidental catches in tuna longline fisheries have become the main threat to the diversity of the pelagic ocean ecosystem [1,2]. The Food and Agriculture Organization (FAO) developed technical guidelines for the conservation and management of bycatch, including sharks [3], seabirds [4], and marine turtles [5]. Great progress has been made in developing effective and commercially viable bycatch mitigation measures, which include finning controls, catch controls, operational and gear controls, as well as the

modification of fishing gear and fishing behaviors. However, recent studies show that the effectiveness of these approaches remains to be evaluated [6]. Given that, area-based management tools (ABMTs), such as time–area closures, selective area-based fishery/gear closures, marine protected areas (MPAs), and adaptive/real-time management in blue-water ecosystems, have become a leading topic in fishery management [7]. ABMTs, such as time–area closures, have been used for a long time in tuna regional fishery management organizations to reduce the interactions of fisheries with dolphins, juvenile tunas, and swordfish [7]. However, ABMTs have seldom been applied to reduce the bycatch of other bycatch species, such as sharks, rays, and marine turtles, especially in the high seas, as the knowledge of habitat preference for these species is limited [8].

Stingrays are a group of sea rays, which are elasmobranchs related to sharks. They are common in coastal tropical and subtropical marine waters throughout the world. Most stingrays are found in warm, temperate oceans and fresh waters, but some are pelagic. *P. violacea*, perhaps the only stingray to inhabit open ocean waters, is widespread throughout the tropical and subtropical areas of the Pacific, Atlantic, and Indian Oceans [9]. It is listed as “Least Concern” on the IUCN list [10]. This species is a regular bycatch in both purse seine and longline gears and is also caught by coastal fisheries for local subsistence. In the Pacific Ocean [9], *P. violacea* is one of the most frequently caught elasmobranchs in tuna longline fisheries [11,12]. Although a previous study found that this species increased in central Pacific longline fisheries between the 1950s and 1990s [13], a report based on Hawaii-based longline fishery catch and effort data showed a 5.4% annual decline in the catch rate of *P. violacea* [14]. More recently, ecological risk assessments for both tuna fisheries in the western and central Pacific Oceans and the eastern Pacific Ocean showed that *P. violacea* has been at medium risk due to its large removal from pelagic waters [15,16].

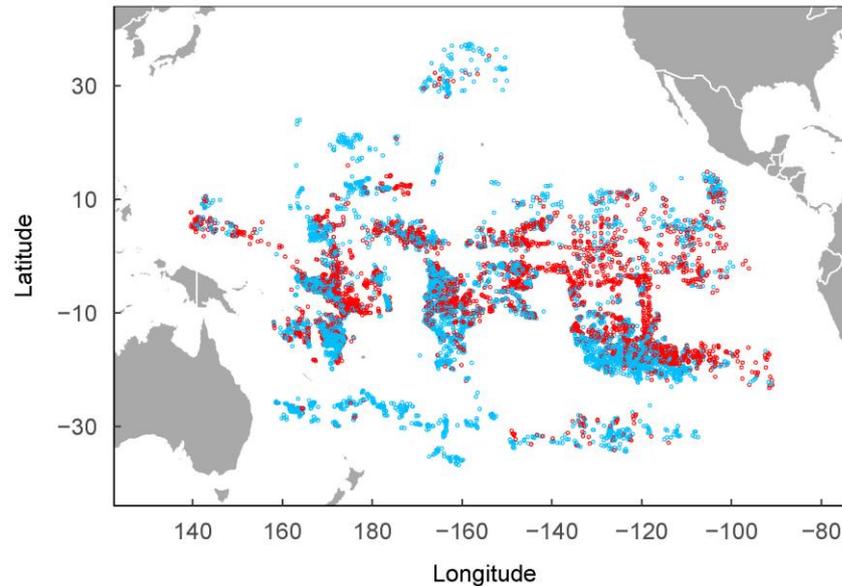
The key to minimizing the interactions of fisheries with bycatch species and identifying areas of high bycatch rates is to understand the relationship between species distributions and spatial–temporal and environmental factors [17]. Observer programs, designed to monitor catch and bycatch and collect fishery-relevant information [18], are an important source of ecological data that underpin the application of ecological-based fishery management [19]. Data from these programs have been widely used in building species distribution models (SDMs) to explore the relationship between species’ presence or abundance and environmental characteristics and predict the potential habitat of species with high ecological risks [17,20–23]. Several abiotic factors, such as temperature, salinity, and dissolved oxygen, and biotic factors, such as plankton aggregations, have been demonstrated to influence the movement of sharks and rays either indirectly or directly [8]. Climate changes are also important factors affecting the migration and distribution of sharks and rays [24]. In the case of *P. violacea*, a study has shown that the catch rate of *P. violacea* in the Spanish longline fleet in the Mediterranean during 2000–2013 was correlated with the climatic variability driven by the North Atlantic Oscillation. The study revealed that the high catch rate along the coast of the Iberian Peninsula might be induced by the negative North Atlantic Oscillation phase, which increases the contribution of land-based nutrients to the sea [10].

A few pieces of research have been performed to better understand its biological and fishery characteristics [12,25,26]. However, studies on the environmental characteristics associated with the presence of *P. violacea* are few in number, especially in the Pacific high sea. Thus, this paper aims to use observer data from Chinese tuna longline observer programs to (1) explore the environmental characteristics associated with the presence of *P. violacea* in the high seas of the Pacific Ocean and (2) find its potential key habitat and dynamics for future conservation issues.

## 2. Materials and Methods

### 2.1. Study Areas

The Pacific Ocean is the world's largest tuna fishing ground, providing more than 55% of the global tuna catch [27]. The study area covers most of the tuna longline fishing grounds (30°S–25°N and 150°E–100°W), as shown in Figure 1, including both tropical and temperate waters.



**Figure 1.** Observed sets of tuna longline fishery from 2016 to 2019. The red dots represent the observed presence of *P. violacea*, and the blue dots represent the observed absence of *P. violacea*.

### 2.2. *P. violacea* Data

According to the Conservation and Management Measures for the Regional Observer Programs in the Western and Central Pacific Fisheries Commission (WCPFC), a total of 10,291 fishing data sets were collected by observers onboard Chinese tuna longline vessels in the Pacific Ocean from 2016 to 2019 [28]. The scientific observers were first rigorously trained to collect fishery data on tunas and other pelagic fish stocks. Next, the observers were randomly allocated to longline vessels on the fishing grounds. During their trips, the observers were required to record basic information about the fishing operations (which included the fishing/hauling time, positions, hooks between floats, length of float and branch lines, setting speed of the mainline and vessels, and target species) and catch information (which included all species captured by the sets) at the species level. More details are available in the China annual report to the WCPFC [28].

### 2.3. Environmental Data

The details of the environmental variables are shown in Table 1. The values of the dynamic oceanographic variables for each fishing set were derived from the Copernicus Marine Service (CMEMS) (<https://marine.copernicus.eu/>) (accessed on 8 April 2022). The Ocean Nino Index (ONI), derived from the Climate Prediction Center, National Oceanic and Atmospheric Administration (<https://origin.cpc.ncep.noaa.gov>) (accessed on 8 April 2022), was used to explore the impact of climate change (the El Niño–Southern Oscillation, ENSO) on the presence of *P. violacea*. The values of bathymetry and distance to the nearest land were obtained from Global Marine Environment Datasets (<https://gmed.auckland.ac.nz/index.html>) (accessed on 8 April 2022). The distance of each longline set to the closest seamount was estimated using the simple spherical law of cosines and a Pacific seamount dataset [29,30]. Only seamounts with summits shallower

than 400 m, which had a significant aggregation effect on pelagic species, such as tuna, sharks, billfishes, marine mammals, and sea turtles [31], were used in the distance calculation.

**Table 1.** Summary of the environmental variables.

Variables Acronym	Variable Name	Units	Average	Min	Max	Spatial Resolution	Temporal Resolution
SST	Sea surface temperature	°C	27.142	15.491	31.219	0.25°	Monthly
SAL	Sea water salinity	psu	35.306	33.053	36.707	0.25°	Monthly
SSH	Sea surface height	m	1.038	0.568	1.392	0.25°	Monthly
MLT	Mixed layer thickness	m	51.290	11.900	188.800	0.25°	Monthly
Chl	Chlorophyll concentration	mg/m <sup>3</sup>	0.128	0.029	0.475	0.25°	Monthly
O2	Oxygen concentration	mmol/m <sup>3</sup>	207.019	194.282	254.143	0.25°	Monthly
Phy	Phytoplankton concentration	mmol/m <sup>3</sup>	1.242	0.441	2.432	0.25°	Monthly
Ni	Nitrate concentration	mmol/m <sup>3</sup>	1.255	0.000	9.004	0.25°	Monthly
ONI	Ocean Nino Index	-	-	-	-	-	Monthly
Depth	Depth	m	4240.165	903.187	7781.958	5 arc-minute	-
Land_dis	Distance to the nearest land	Km × 1000	6.767	0.099	22.407	5 arc-minute	-
Seamounts_dis	Distance to the nearest seamount	Km × 1000	4.863	0.010	20.486	30 arc-seconds	-

#### 2.4. Statistical Analysis

The Generalized Additive Model (GAM) [32] was used to quantify the statistical relationship between the presence/absence and the environmental, spatial (latitude and longitude), and temporal (year and month) variables. The general structure of the GAM is as follows:

$$g(Y) = \alpha + \sum_{i=1}^n f_i(x_i) + \varepsilon$$

where  $g$  is the link function,  $Y$  is the expected response variable,  $\alpha$  is the intercept,  $f_i$  are smooth functions,  $x_i$  are the covariates, and  $\varepsilon$  is the residual error.

The degree of freedom of the smooth functions was restricted for each explanatory variable to avoid additional overfitting. The number of the basis function ( $k$ ) was set to  $k = 20$  for the interaction effects,  $k = 4$  for the year effects, and  $k = 6$  for the rest effects [22,23].

To avoid correlations, one variable from the variable pairs with a high correlation ( $r > 0.7/r < -0.7$ ) was removed based on the Pearson's rank correlation [33]. To deal with multicollinearity problems, a variance inflation factor analysis (VIF) was conducted with a cut-off value of 5 [17]. Based on these tests, the variables SSH, O<sub>2</sub>, Phy, and Ni were removed owing to correlation/collinearity with more ecologically important variables [8,22]. All other covariates available for the model selection had low cross-correlation and cross-collinearity scores (Figure S1 and Table S1).

The best-fit model was selected using the forward stepwise variable selection procedure (Table S2). The model with the lowest Akaike's Information Criterion (AIC) was chosen as the final model [34]. The percentage of the deviance explained was used to measure the model's goodness-of-fit.

A 5-fold cross-validation was applied to assess the model's performance [20,22]. This approach divides the data into two groups, where 80% of the data are used to establish relationships between the presence-absence data and the variables and 20% of the data are used to evaluate the quality of the model's predictions. The area under the receiver operating curve (AUC) was used to evaluate the model's performance [20,22]. It measures the ability of a model to correctly predict the presence or absence of a species. The AUC

value ranges from 0 to 1. The closer the AUC value is to 1, the better the model's prediction [35].

The spatial ( $1^\circ \times 1^\circ$ ) and seasonal distributions of *P. violacea* were predicted using the best-fit models. The seasons were defined as follows: winter, Jan–Mar; spring, Apr–Jun; summer, Jul–Sep; autumn, Oct–Dec.

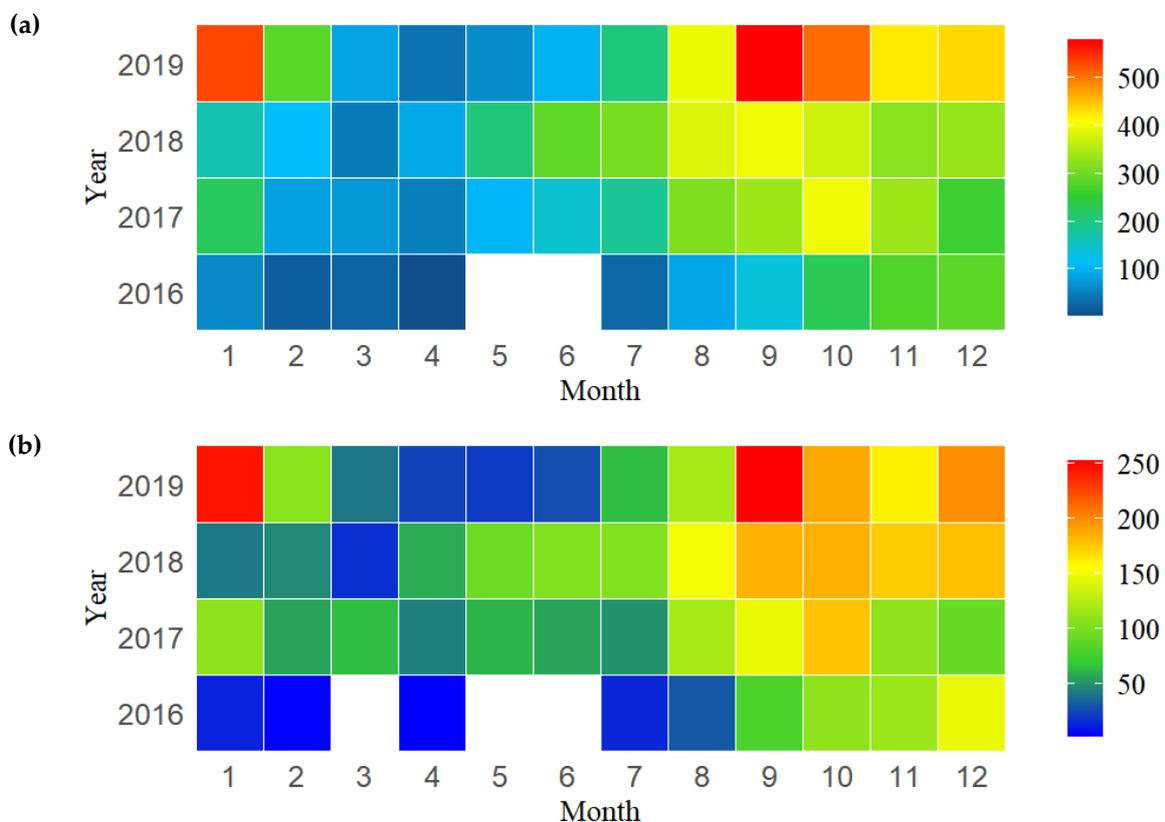
The functions and packages used in the analyses are shown in Table 2. All analyses were carried out in the R-4.1.1 software [36].

**Table 2.** Summary of the functions and packages used in the analyses.

Analysis Method	Function	Package	Reference
GAM	gam and prediction.gam	mgcv	[37]
Akaike's An Information Criterion	AIC	stats	[38]
Correlation analysis	cor and corrplot	stats and corrplot	[38,39]
Multicollinearity test	vif	car	[40]
Cross-validation	createMultiFolds, auc	caret and PresenceAbsence	[41,42]

### 3. Results

Out of the 10,291 sets, 4349 were found to have the presence of *P. violacea* during the study period between 2016 and 2019. Most of the fishing sets (75.6%) and the sets with the presence of *P. violacea* (76.1%) were observed from August to January. The highest number of fishing sets was observed during October (1508), followed by September (1451) and November (1356) (Figure 2a). The number of sets with the presence of *P. violacea* was the highest in September (662), followed by October (658) and December (615) (Figure 2b).



**Figure 2.** (a) Monthly distribution of the observed number of fishing sets in Chinese tuna longline fisheries. (b) Monthly distribution of the observed number of fishing sets with the presence of *P. violacea*.

Based on the lowest AIC, the variables included in the best GAM model were as follows: the interaction of longitude and latitude, month, SAL, ONI, Chl, SST, year, MLT, Land\_dis, Seamounts\_dis, and depth (Table 3).

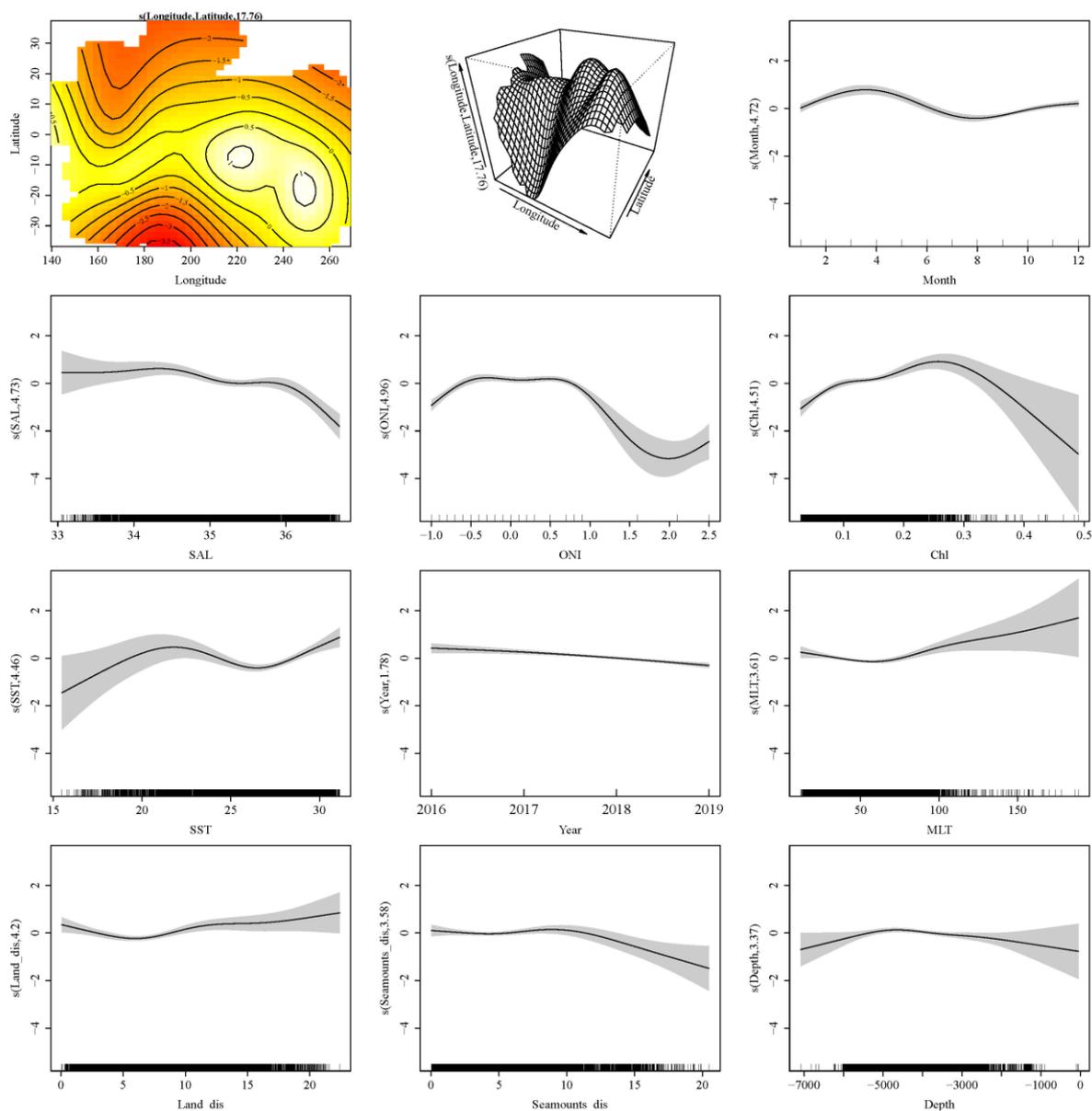
The model achieved a 15.65% deviance explained with an adjusted  $R^2$  of 0.187. The individual contributions of the variables are shown in Table 3. The most important covariates to explain the presence of *P. violacea* are the interaction latitude–longitude (10.84%), followed by month (1.06%), ONI (0.91%), and SAL (0.89%) (Table 3).

**Table 3.** Summary analysis of the deviance of the generalized additive model.

Family	Binominal		
Link Function	Log		
Adjusted $R^2$	0.187		
Deviance Explained	15.65%		
	edf	P-Value	Deviance %
Latitude*longitude	17.762	<0.001	10.84
Month	4.719	<0.001	1.06
SAL	4.728	<0.001	0.89
ONI	4.96	<0.001	0.91
Chl	4.507	<0.001	0.63
SST	4.457	<0.001	0.36
Year	1.78	<0.001	0.20
MLT	3.612	<0.001	0.21
Land_dis	4.203	<0.001	0.27
Seamounts_dis	3.578	<0.001	0.18
Depth	3.369	0.023	0.10

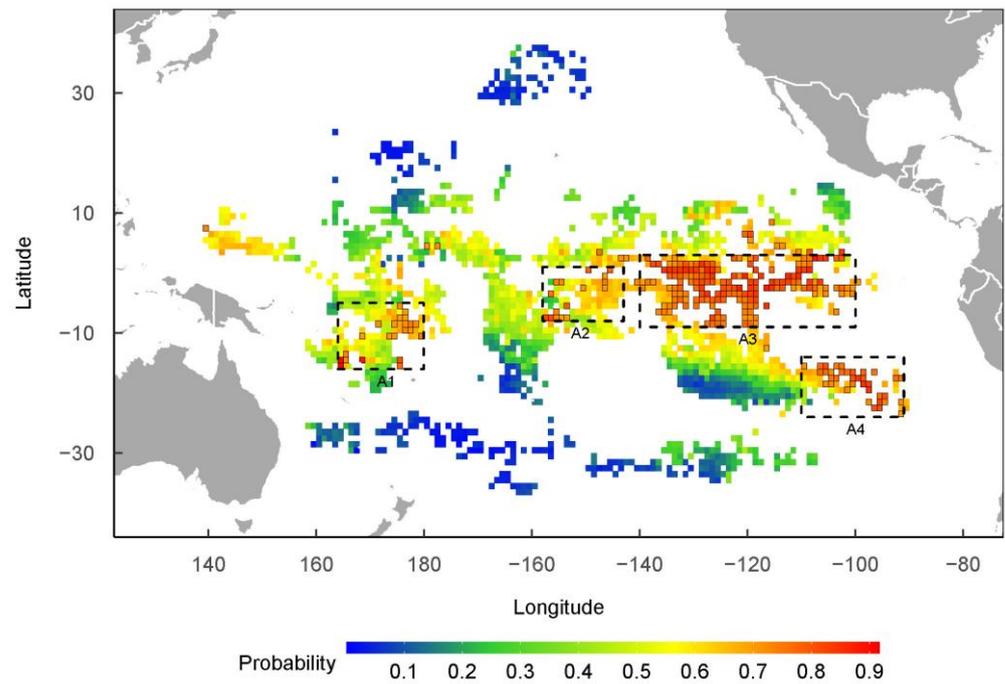
The relationships between the covariates and the response variables are shown in Figure 3. The model suggests that the presence probability of *P. violacea* was higher in the eastern tropical Pacific ( $0^{\circ}$ – $20^{\circ}$ S,  $100^{\circ}$ W– $140^{\circ}$ W) from December to May. The significant annual variation indicates that the presence of *P. violacea* has decreased in recent years in the Pacific high sea. The presence of *P. violacea* during the climate periods with  $-0.5 < \text{ONI} < 1$  was higher than that during the cold ( $\text{ONI} < -0.5$ ) and warm ( $\text{ONI} > 1$ ) phases. Regarding the environmental variables, *P. violacea* was more likely to be present in waters with low SAL (<34.5 psu), relatively high concentrations of chlorophyll (0.20–0.35 mg/m<sup>3</sup>), warm waters (>20 °C), and deep MLT (>80 m). *P. violacea* was more concentrated in open waters far offshore, closer to seamounts, and at seafloor depths of 4000–5000 m.

The average AUC (0.76, 0.74, 0.74, 0.75, and 0.75) for the model was 0.75, which revealed that the ability of the model to predict the presence of *P. violacea* was moderate.

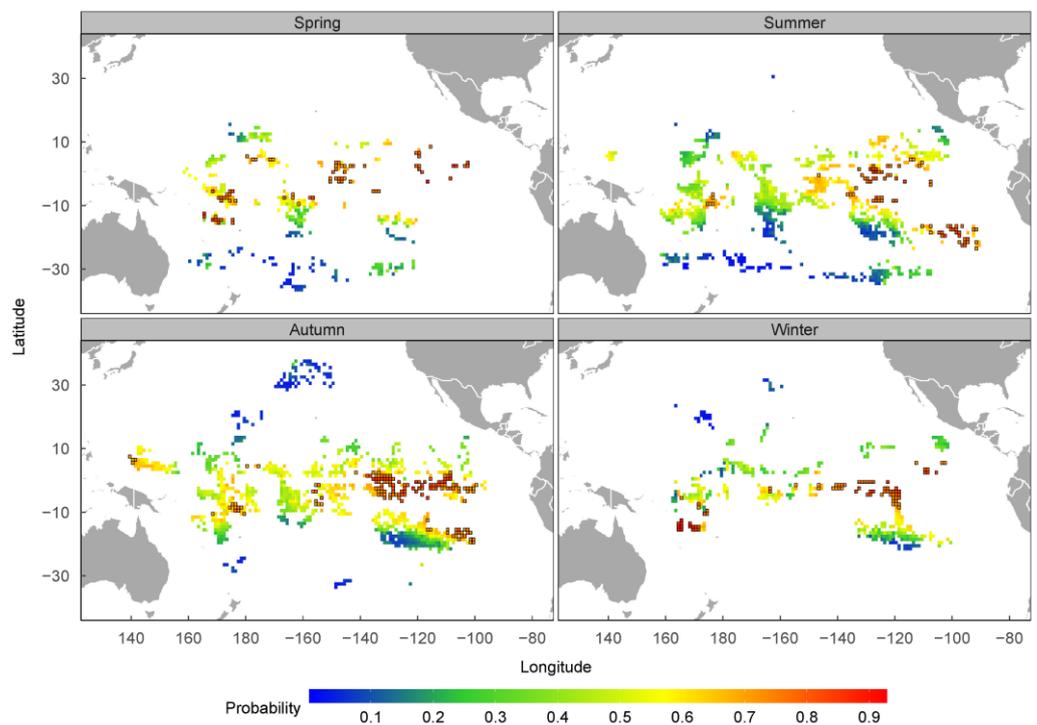


**Figure 3.** Effects of the spatial–temporal and environmental variables on the presence of *P. violacea* in the Pacific Ocean.

Based on the predicted spatial distribution of the presence probability of *P. violacea* from the best model, the following four areas with a concentrated distribution of grids ( $1^\circ \times 1^\circ$ ) with a high probability of presence ( $\geq 0.7$ ) were identified as potential habitats: A1 ( $5^\circ\text{S}$ – $16^\circ\text{S}$ ,  $164^\circ\text{E}$ – $180^\circ\text{E}$ ), A2 ( $1^\circ\text{N}$ – $8^\circ\text{S}$ ,  $143^\circ\text{W}$ – $158^\circ\text{W}$ ), A3 ( $3^\circ\text{N}$ – $9^\circ\text{S}$ ,  $120^\circ\text{W}$ – $140^\circ\text{W}$ ), and A4 ( $14^\circ\text{S}$ – $24^\circ\text{S}$ ,  $91^\circ\text{W}$ – $110^\circ\text{W}$ ), as shown in Figure 4. Most of the grids with a probability of  $\geq 0.7$  were concentrated in the A3 area. The predicted seasonal spatial distribution suggests that *P. violacea*'s presence might change seasonally. For example, the high presence probability occurred in each season in A3, but only in the summer and autumn in A4 and in the spring in A2. The area with a high probability of *P. violacea*'s presence was more extensive in the spring and winter than in the summer and autumn in A1 (Figure 5).



**Figure 4.** The spatial distribution of *P. violacea*'s presence probability, predicted from the GAM model. The grids ( $1^\circ \times 1^\circ$ ) with a presence probability of  $\geq 0.70$  are bordered in black. The areas within the dashed border are considered the potential key habitat of *P. violacea*, as most of the high-probability grids are distributed in these areas.



**Figure 5.** The seasonal spatial distribution of *P. violacea*'s presence probability, predicted from the GAM model. The grids ( $1^\circ \times 1^\circ$ ) with a presence probability of  $\geq 0.70$  are bordered in black.

#### 4. Discussion

Researchers have paid the most attention to the biological characteristics of *P. violacea*, but research on the species' ecological habits and environmental preferences is lacking. This study examined the presence probability of *P. violacea* in the Pacific high sea using bycatch data from Chinese tuna longline fisheries. The key environmental variables that may affect *P. violacea*'s presence in this region were identified by using GAMs. The results revealed that the presence of *P. violacea* in the Pacific high sea was strongly associated with environmental variables (SAL, Chl, and SST) and was significantly impacted by the ENSO. The habitat hotspots for *P. violacea* generally occurred in waters associated with upwelling systems and showed significant annual and seasonal variations. This study will be fundamental to developing effective spatial fishery management regulations.

##### 4.1. Environmental Preference of *P. violacea*

By understanding the environmental preference of the species, we could know if there is a higher probability of finding the species in other studied areas. Salinity and temperature are well-known to have a strong influence on the physiology of fish and are key drivers of fish migration [8,43]. Whereas many animals can tolerate a wide range of salinities, most sharks and rays strictly occupy a narrow range of salinities [43]. The normal range of ocean salinity ranges between 33 and 37 psu. Near the equator, the tropics receive the most consistent rain. As a result, the freshwater falling into the ocean helps to decrease the salinity of the surface water in that region [44]. Although *P. violacea* is nearly universal in subtropical and tropical seas and also occurs at temperate latitudes [9], our findings show that it is more likely present in waters with relatively low salinities, such as those near the equator, and it is rarely observed in temperate high seas, especially in areas where the salinity is higher than 35 psu. Regarding the SST, the previous study showed that *P. violacea* prefers water temperatures above 19 °C and will die if the temperature drops to 15 °C [25]. Our study further demonstrated that *P. violacea* prefers warm water habitats with temperatures ranging between 20 and 25 °C, which is also a suitable temperature for many pelagic elasmobranchs, such as the scalloped hammerhead (*Sphyrna lewini*), the smooth hammerhead (*Sphyrna zygaena*), and the silky shark (*Carcharhinus falciformis*) [45].

In the open ocean, areas with high chlorophyll concentrations are generally associated with upwelling systems, indicating an abundance of primary productivity. Unlike the spinetail devil ray (*Mobula mobular*), which prefers habitats in coastal waters with relatively high concentrations of chlorophyll (0.5–1 mg/m<sup>3</sup>) [22], we found that *P. violacea* was more likely to be found in areas with chlorophyll concentrations between 0.20 and 0.35 mg/m<sup>3</sup>. Such environmental conditions, in combination with other oceanographic characteristics, have also attracted multiple pelagic species to form a species diversity hotspot in the eastern and central tropical Pacific [2,46].

##### 4.2. Impact of ENSO on the Presence of *P. violacea* in the Pacific High Sea

For marine fisheries, the volume and dominant species of fish catches can change dramatically depending on the type of ENSO [47]. ENSOs are often simplified to reflect the following two main phases: El Niño, an anomalous warming phase, characterized by cooler ocean temperatures in the central and eastern tropical Pacific Ocean due to increased upwelling, and an opposite cooling phase called La Niña, characterized by an anomalous increase in SST in the eastern Pacific Ocean as upwelling of cold water occurs less or not at all in offshore northwestern South America. Our study first showed that the presence probability of *P. violacea* in the Pacific high sea was lower both during the cold (ONI < -0.5) and warm (ONI > 1) phases than in the neutral phase. Sharks and rays are known to exploit the thermal heterogeneity in their environment by selecting different temperatures throughout the day, a behavior known as thermotaxis [48,49]. For example,

studies have shown that sharks and rays prefer to forage in warm waters, which maximizes their flexibility, and rest in cold waters to reduce their daily energy costs [24,50]. Thus, elasmobranchs may not change their latitudinal range as the climate warms but rather select deeper, colder water areas to enhance their physiological processes [24]. Based on this, one possible explanation for the relationship between *P. violacea*'s presence in the Pacific high sea and ENSO is that the El Niño events might drive *P. violacea* into deeper waters or away from the equator, and *P. violacea* might migrate towards warmer waters in the central and/or western Pacific under La Niña events. However, given the limited amount of data available, these effects are difficult to characterize with certainty.

#### 4.3. Spatial–Temporal Distribution of *P. violacea* in the Pacific High Sea

The typical habitats or biodiversity hotspots of pelagic fishes are mostly located in areas characterized by upwelling systems, seamounts [29,31], or ocean fronts [51]. Our results showed that the hotspots of *P. violacea*'s presence in the Pacific high sea were mainly in four regions. The area east of the Solomon Islands, near the zone of dense seamount distribution [30], is characterized by increased turbulence, mixing, and mesoscale eddies that can enhance local production by transporting nutrients to the poleward zone, where the high primary productivity and abundant bait attract predators to congregate in the area [29,52,53]. The area east of Kiribati and west of the Galapagos Islands, which is rich in nutrients due to equatorial upwelling, is an important habitat for elasmobranchs [20,22]. The waters off the Peruvian coast, which are also influenced by upwelling systems, have a large accumulation of nutrients that affect the abundance and distribution of marine organisms [54]. These areas are either rich in fishery resources, such as the coastal waters of Peru, or have high biodiversity as critical habitats for many pelagic species, such as the area east of Kiribati and the west of the Galapagos Islands, and would be the priority areas for attention and management.

*P. violacea* spends the winter in oceanic waters near the equator and moves into higher latitudes and toward the coast in the spring [25]. The seasonal changes in *P. violacea*'s presence in the Pacific high sea might reflect its migrations between the coast and the open ocean. Furthermore, the significance of the year in the model indicated that there had been a decreasing trend concerning the presence of *P. violacea* in the Pacific high sea over the years. In the long run, the observed catch rate for *P. violacea* in longline and purse seine sets within the western and central Pacific Oceans tends to exhibit large annual fluctuations. The catch rate declined in both gears and all set types from the mid-1990s to the mid-2000s and then seemed to increase again after 2010 [55]. Researchers speculated that this increase may be related to the decline in mesopredator release as a result of declines in sharks and billfishes [10]. The short-term interannual variability in our study might also indicate a change in its abundance, or it may be related to the interannual fluctuation of climate change affecting its presence in the Pacific high sea. Whatever the underlying causes, the results suggested, to some extent, that anthropogenic factors have directly or indirectly influenced the biomass or distribution of *P. violacea*. Further consideration should be given to whether this impact is negative or neutral to prevent irreversible effects on the population.

#### 4.4. Conservation Consideration

Stingray species are progressively becoming threatened or vulnerable to extinction, particularly as a consequence of unregulated fishing. The Food and Agriculture Organization of the United Nations published the International Plan of Action for Conservation and Management of Sharks, which includes non-shark species of chondrichthyan fishes, such as skates, rays, and chimaeras, in 1999 [3]. WCPFC was the first tuna regional fishery management organization to establish a formal shark research plan covering stock assessment, research coordination, and fishery statistic improvements [56]. Currently, more than ten species have been designated as key sharks, which have priorities for conservation and management [57]. *P. violacea* made up a high proportion of

the non-key shark bycatch in longline fisheries [12]. To be listed as a WCPFC key shark species, the management of *P. violacea* could be improved by enhanced reporting by vessels. However, the scientific commission did not recommend *P. violacea* as a WCPFC key shark species because of their medium vulnerability to most fishing gear used in the WCPFC, as well as their medium productivity [15,55]. In this sense, there are no targeted bycatch mitigation measures or data collection and research plans for *P. violacea*. However, several bycatch mitigation measures, such as adjusting the gear depth, reducing the use of J-hooks, and encouraging the use of line cutters or de-hookers to release, have also been effective in reducing the bycatch rates and post-release mortality of *P. violacea* [6]. The continued removal of large numbers of these organisms from the pelagic realm will have unavoidable ecological consequences.

Data based on fisheries, especially those with long time series and large spatial coverage, provide us with information on the spatial and temporal distributions of bycatch species [12,19]. SDMs, based on fishery observer programs, are widely used to explore the relationship between environmental variables and bycatch species distribution, as well as to identify and predict habitat hotspots [17,20,22]. The predictive maps of *P. violacea* allowed us to identify the seasonal–spatial dynamics of *P. violacea*'s distribution [58]. As data become more abundant for *P. violacea* and other vulnerable species, the dynamic changes in key habitats could be further predicted monthly, weekly, and even daily [59]. Such studies serve as a basis for dynamic adjustments of fishing locations to prevent vulnerable species from encountering fishing gear [7,58,60,61].

## 5. Conclusions

This study improves our understanding of *P. violacea*'s environmental preferences and spatial and temporal distributions in the Pacific high sea. Seawater salinity, the concentration of chlorophyll, and sea surface temperatures were the key environmental variables that shaped the distribution patterns of *P. violacea* in the study area. Both the warm and cold phases of ENSO could reduce the presence of *P. violacea* in the Pacific high sea. The key habitats of *P. violacea* were generally found in areas close to the seamounts in the western Pacific or the upwelling systems in the eastern Pacific. Moreover, the prediction output of the model revealed the seasonal variation and the interannual fluctuation of *P. violacea*'s presence in the Pacific high sea. This study provides important information for managers on how to shape their bycatch mitigation strategies to reduce mortality by understanding the ecological dynamics of bycatch species.

**Supplementary Materials:** The following supporting information can be downloaded at: <https://www.mdpi.com/article/10.3390/fishes8010046/s1>, Figure S1: Pearson's correlation coefficient between the variables; Table S1: Collinearity between the variables by calculating the Variance Inflation Factor (VIF); Table S2: Model selected by stepwise analysis

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