

Catch substitution between Coastal Pelagic Species under climate change scenarios

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Abstract

Fisher do not only catch one species. They have a set of possible choices, called ‘fishing portfolio’. Some species are easier to shift as gear and method used are similar between them. Therefore, the cost of shifting between species is low and fishers could adapt quickly to a shift in fish species spatial distributions in response to climate change. Nevertheless, it is not clear whether actually this substitution happen as other constraint may be in play. Port constraints, as well as market characteristics could reduce substitution between species. In this study we analyze how changing in spatial distribution affects landing substitution between three coastal pelagic species: Pacific sardine, market squid and Northern anchovy. We primary focus on the substitution that ocur between these species, and we project change on catch composition due to future climate change.

1 Introduction

Fishing portfolios are an important mechanism to safeguard fishers livelihood. Diversification strategy have been principally associated to reducing income variability. For instance, when a species abundance is reduced due to environmental conditions, fisher can change the targeted species. However, there is no always room for diversification. Switching between species can be costly if gears are quite different between species, or a new permit is required for legal fishing. Moreover, even though fisher may have flexibility switching between species, port infrastructure and markets may impose some restrictions on this flexibility. Therefore, it is not clear how change in species distribution would be reflected in landings.

In this study we analyze how changing in spatial distribution affects landing substitutions between three coastal pelagic species (CPS): Pacific sardine (PSDN), market squid (MSQD) and Northern anchovy (NANC). Moreover, using climate projections we can predict how catch composition is affected under climate change. Our analysis is focused on the US west coast CPS fishery.

Our papers builds on the model developed by [Smith et al. \[2021\]](#) for sardine landings. We expand their research Fisheries historically have been analyzed as single species. We expand their model including a landing equation for

all of the three species analyzed. Moreover, we use the probability of presence obtained from Species Distribution Models (SDM) as explanatory variable instead of landing by species. This allows us to project landing over time using SDM predictions for different climate scenarios. Additionally, we analyze how species distribution interact between each others. We expect that this additions allows us to characterize better how fishers portfolio is composed, and also to understand better species interactions on catch rates.

The remainder of the paper is organized as follows: Section 2 provides background on the CPS fishery in the US west coast. In Section 3 we discusses our data set and empirical strategy. Section 4 presents the results of the estimations, and we conclude in Section 5.

2 Coastal Pelagic Species fishery

Before Pacific sardine closure in 2015, the CPS fishery have been mainly dominated by Pacific sardine and market squid in landings (see Figure 1). In terms of revenue, due to the low prices received for sardine, the revenues in the CPS fishery are the highest for market squid.

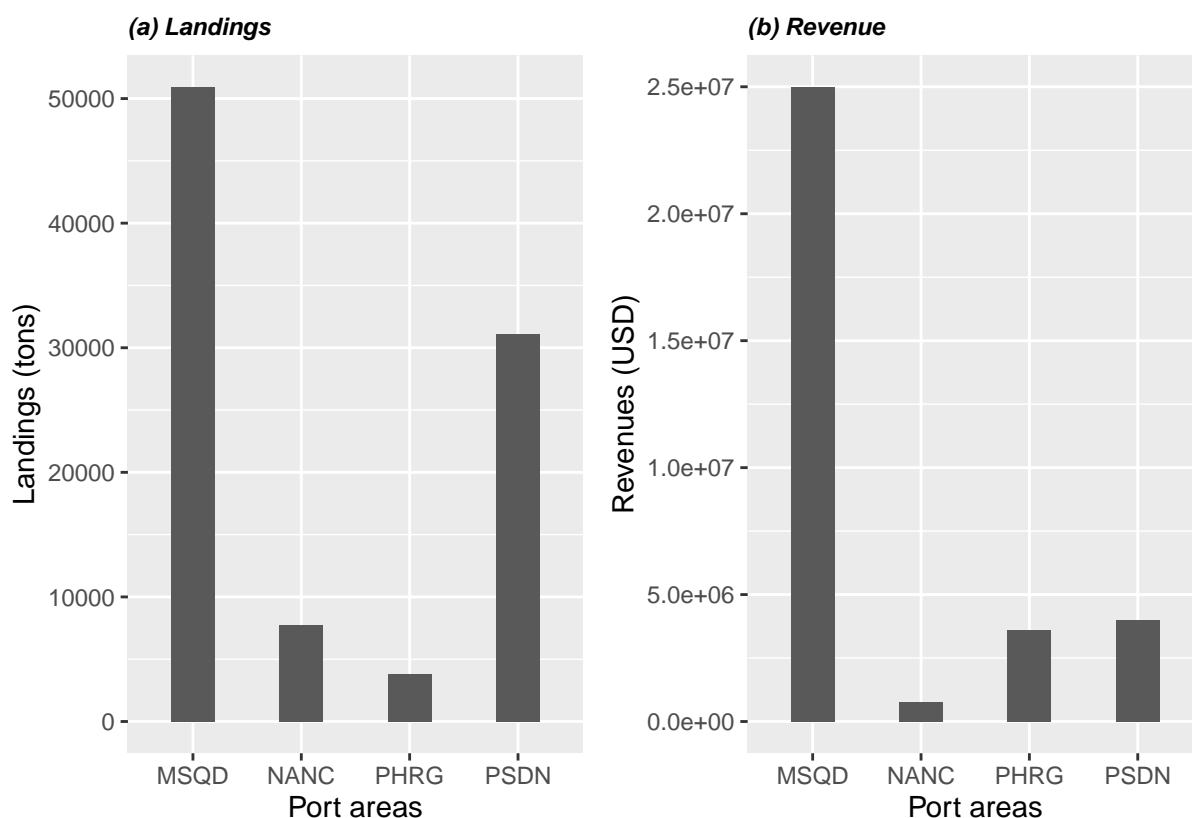


Figure 1: Average annual landing and revenues for the CPS fishery by species.

Landings composition varies geographically. We show in Figure 2 average annual landings by ports areas. We can observe that market squid is mostly landed in the southern ports located in Los Angeles, Santa Barbara and Monterrey areas, while Pacific sardine is mainly landed in Los Angeles and Monterrey areas in California, and also in the Columbia river area in Oregon. Substitution between species seems to be more likely in Los Angeles, Monterrey and Santa Barbara area ports (and in some lower scale at San Francisco area) as positive values for market squid, Pacific sardine and Northern anchovy landings are observed.

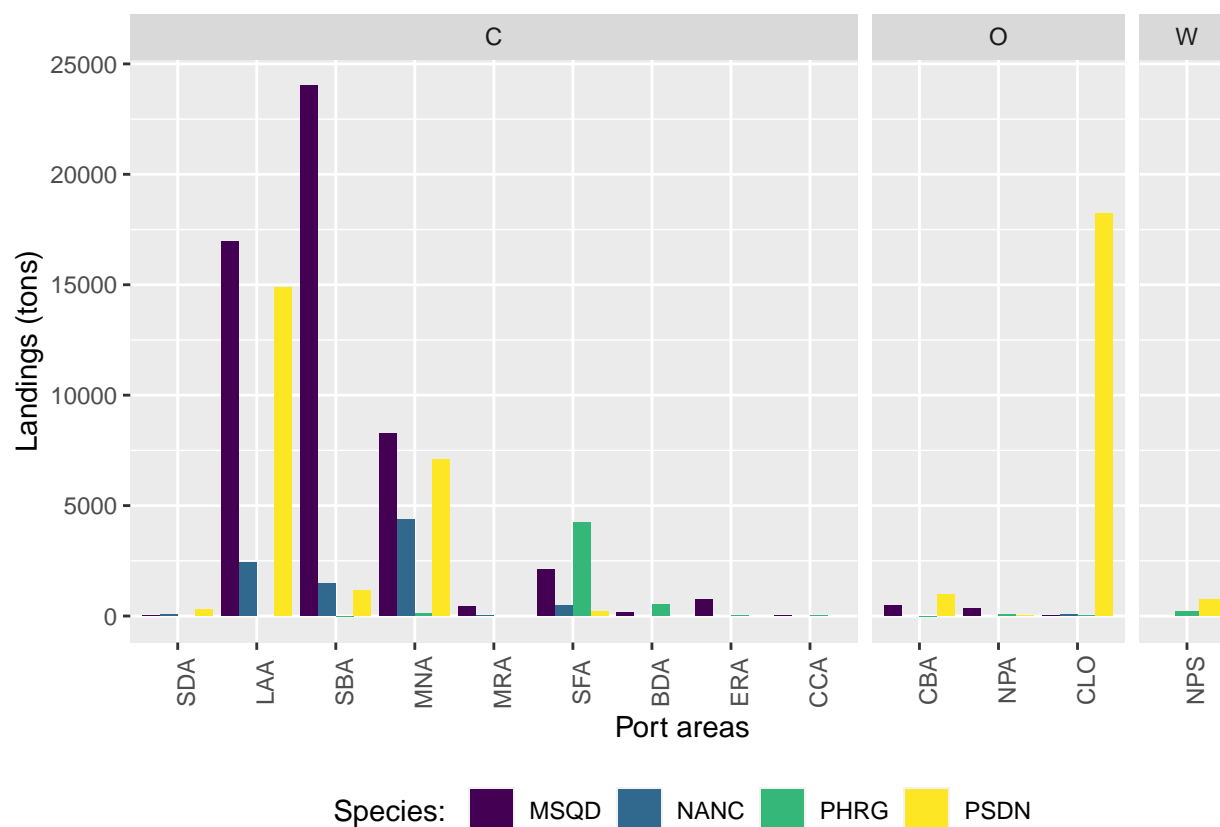


Figure 2: Annual average landings by port area.

To analyze substitution more in detail, we compute total annual landing by port during 1980-2020 period (Figure 3). There is a general conception that vessel that harvest Pacific sardine switch to market squid when the conditions are favorable. However, we can notice from Figure 3 that landings of market squid have been decreasing in the last five years. We would expect that the closure of the Pacific sardine fishery would have switch targeted species, but the graphs show some complementary instead of substitution. Under this scenario is useful to ask to ask whether after the closure, vessel that harvest Market squid left the fishery or stayed. We show the number of vessels by species in Figure 4.

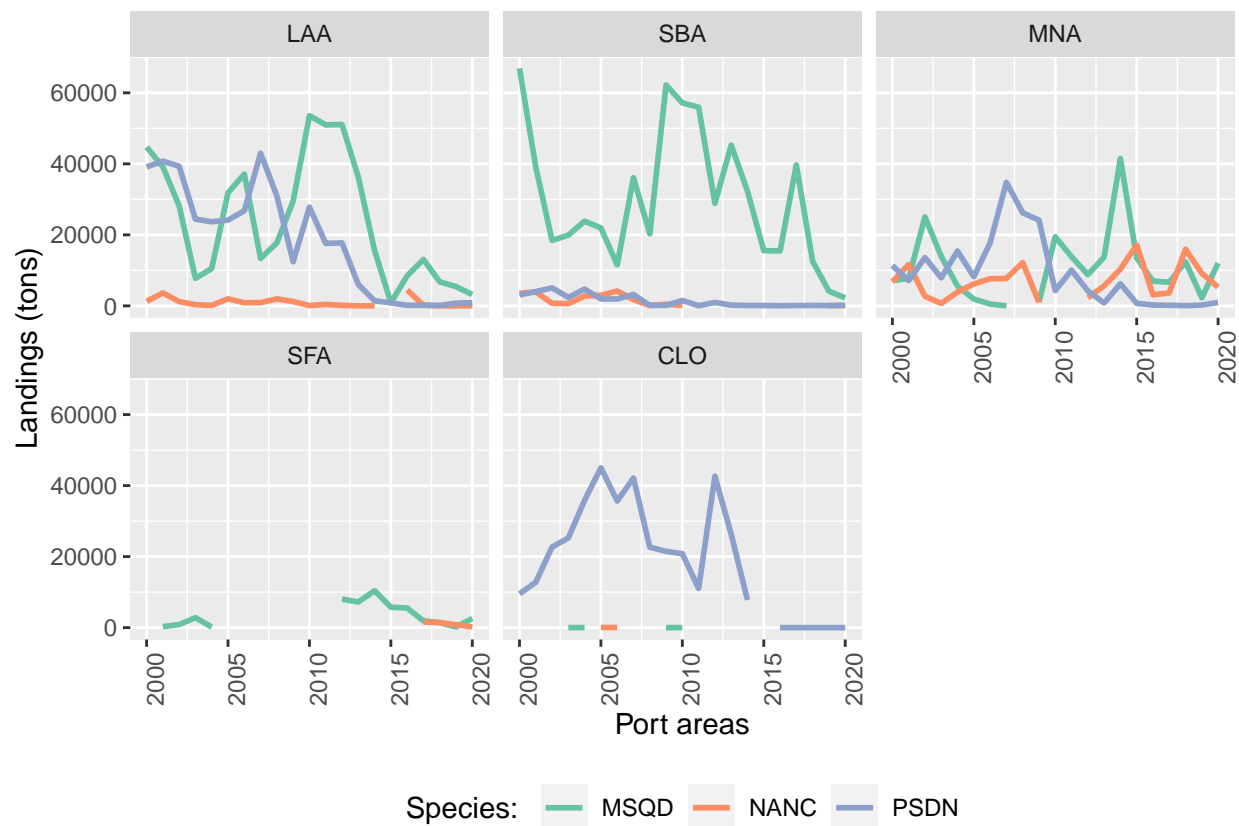


Figure 3: Total annual landing by port area. 2000 - 2020. *Notes:* CLO = Columbia River (OR); LAA = Los Angeles; MNA = Monterey; SBA = Santa Barbara; SFA = San Francisco.

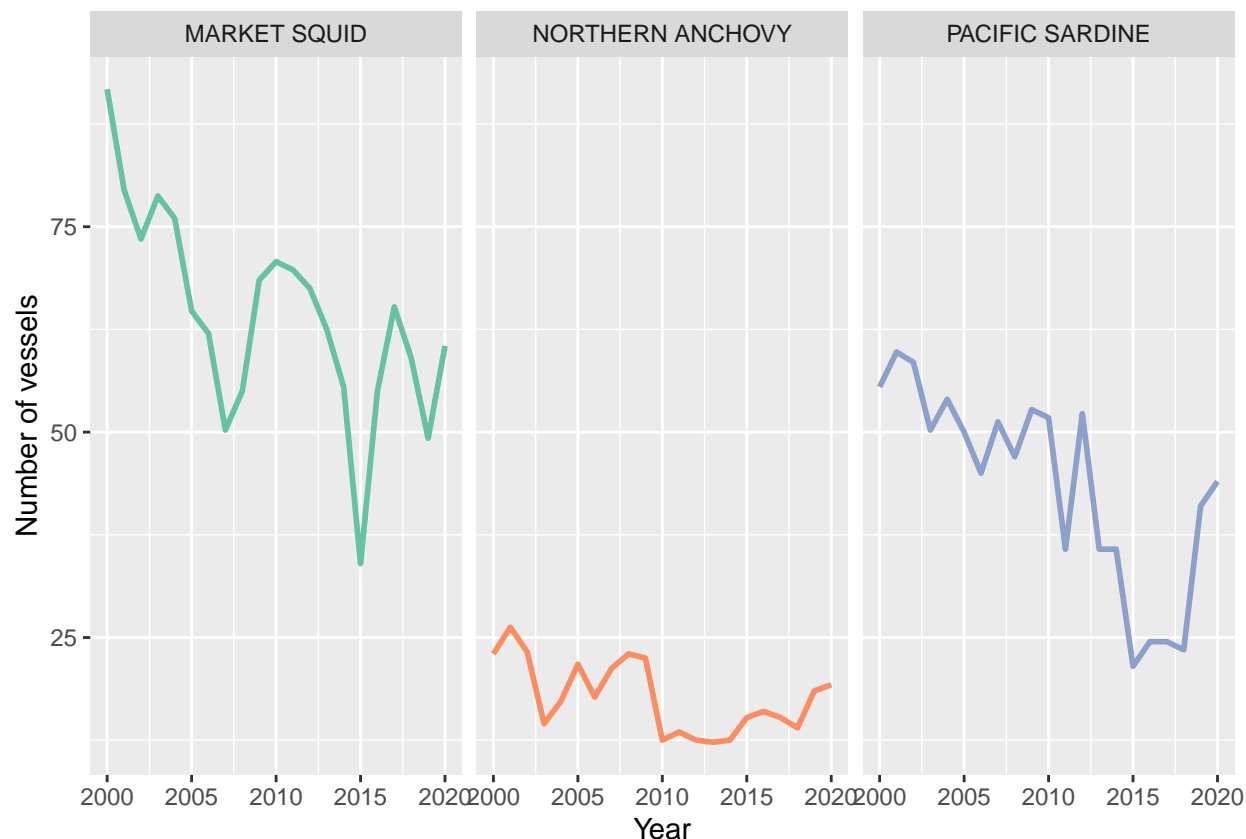


Figure 4: Total number of vessel by species. 2000 - 2020.

3 Methods

3.1 Data

Our data set contain a number of variables measured at the port and *vessel levels* for CPS the fishery located in the U.S. West Coast. Our outcome variables are landings by port areas in a year and landing by vessel in a month. These two different outcome variables would allows us to study the degree of flexibility that vessel have in comparison to ports in regard to species substitution and catch composition. If vessels have strong contracts with processor associated with a port, then we should observe that substitution of vessels and port is similar. Yearly panel data data on landing by port areas during the the period 1980 - 2020 is publicly available from [PacFIN](#). **Vessel level data was obtained upon request from...** We only include ports areas where substitution could happen. In practical terms, we drop port that have never landed either Pacific sardine, market squid or Northern anchovy during our period of analysis.¹ We assume that this criteria would allows to identify ports that have the infrastructure to land all of the species in consideration.

¹N/A were converted to zero.

Our main treatment variables are species probability of presence. These variables were obtained from Species Distribution Models (SDM), and future forecast of these variables allows us to simulate the CPS fishery in the future. Prior landings models have shown that the probability of presence has a large contribution on explaining landings. For instance, [Smith et al. \[2021\]](#) use mean monthly probability of presence of sardine within 60 km of the port as explanatory variable. They found a positive effect of probability of presence on Pacific sardine landings. Moreover, landings were mostly explained by this variable. We follow their same procedure to associate SDM's outputs with ports. For Pacific sardine, we compute the average probability of presence within the same radius of 60 kilometers around the port. This radius also coincides with the average distance with two standard deviation traveled by vessels based on logbook availables for this fisheries. For market squid and Northern anchovy, we set this radius to 90 km. We also use logbook in the case of market squid, while for the Northern anchovy, the optimal distance for Northern anchovy was chosen empirically.²

Figure 5 shows the relationship between the probability of presence and landings by species in three main port areas: Los Angeles, Monterrey and Santa Barbara areas. The graph suggests that Pacific sardine landings are positively correlated with the probability of presence of this species, similar to [Smith et al. \[2021\]](#). This is also true in Monterrey area for the Northern anchovy. In the case of market squid, we cannot distinguish correlation between landings and probability of presence. Note, however, that the evidence shown in this figure may not capture the actual effect of the probability of presence as other effects may be in play. Our empirical strategy is designed to isolate the effect of the probability on landing in a multivariate model framework.

²To see an animation of how far vessel travel please visit [this link](#).

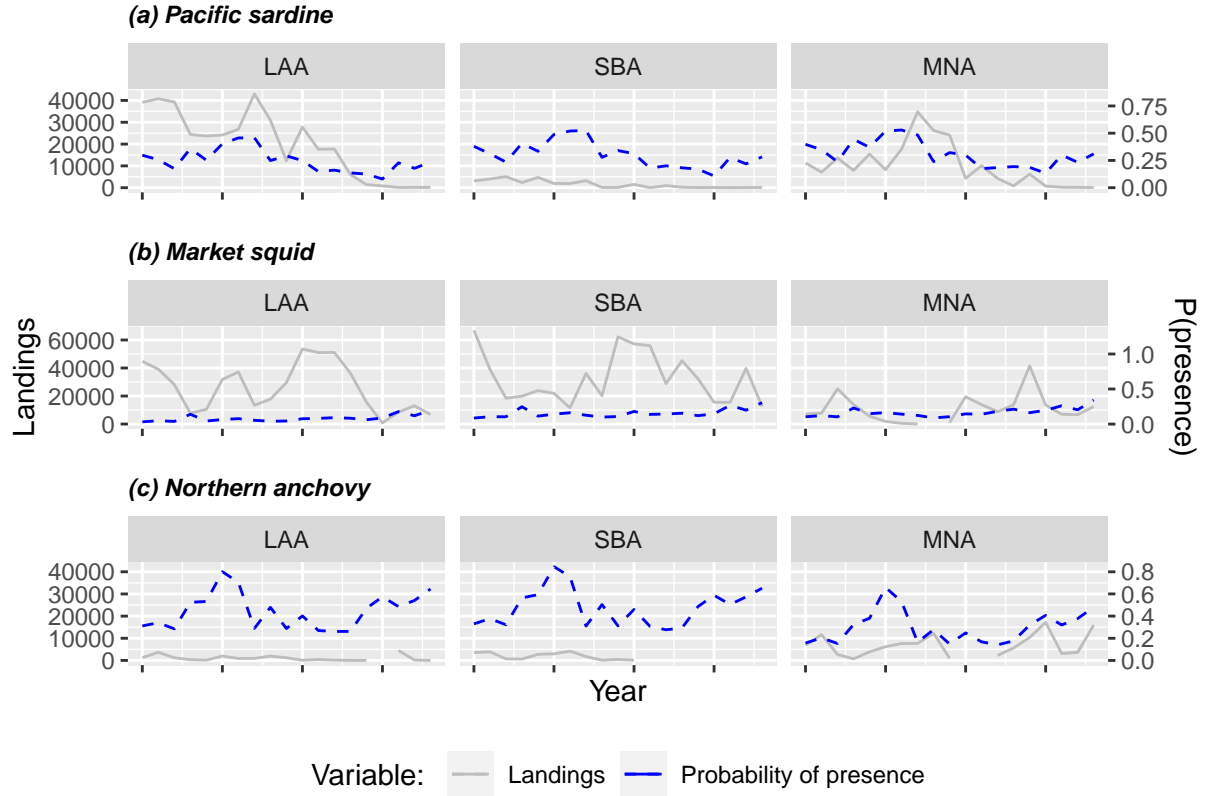


Figure 5: Landings v/s probability of presence by port area. *Notes:* LAA = Los Angeles; MNA = Monterey; SBA = Santa Barbara.

Beside the biological stock, landings are affected by socio-economics conditions. Some of them are harvest cost, prices received by species and their substitutes, and regulations imposed by the authorities. In our data we include as a proxy of harvest cost average distances traveled by vessel from the port of origin (*TO BE INCLUDED USING GLOBAL FISHING WATCH...*) and fuel cost (*TO BE INCLUDED...*). Own price was included and it was obtained from PacFIN landings dataset. When price was missing, we replace this value for the average price of the species in all ports for the corresponding year. As regulatory variables, we construct a binary variable called **dClose** that takes the value “1” when the Pacific sardine fishery is close, and the value “0” when its close. We also include the Annual Catch Limit (ACL) for the Pacific sardine model. The ACL for Pacific sardine were obtained from the [CPS Fisheries Management Plan](#)???. A summary statistics of our data set is shown in Table 1.

3.2 Empirical model

We use a Hierarchical Bayesian Hurdle model to estimate the effect of species distribution on fish landings. We estimate a separate model for the three species in consideration. We use a Bayesian framework for several reasons.

Table 1: Summary Statistics

Variable	N	Mean	Std. Dev.	Min	Pctl. 25	Pctl. 75	Max
Landings: PSDN	92	10828.886	13499.627	0.1	156.025	20991.3	45016.5
Landings: MSQD	103	13383.623	17166.411	0.1	222.2	19693.75	66890.3
Landings: NANC	59	2769.698	3928.055	0	149.55	3651.65	17180.2
Prob(presence): PSDN	322	0.261	0.104	0.08	0.177	0.341	0.569
Prob(presence): MSQD	280	0.17	0.078	0.025	0.12	0.217	0.467
Prob(presence): NANC	308	0.306	0.153	0.085	0.181	0.39	0.846
Price: PSDN	437	0.079	0.039	0	0.05	0.108	0.37
Price: MSQD	437	0.266	0.121	0	0.185	0.32	0.5
Price: NANC	437	0.091	0.054	0	0.06	0.114	0.35
Revenue: PSDN	92	1497414	1920819.177	0	31471	2646660.75	8979099
Revenue: MSQD	103	7734746.485	9938728.985	0	97348	11193641.5	43742173
Revenue: NANC	59	311373.814	427339.961	17	42203	437046.5	1930490
Annual Catch Limit: PSDN	436	82613.151	55614.651	0	30259	122747	186791

First, it allows us to consider uncertainty from modeling the process as well as from the imperfect observation of the process, assuming that all parameters are random variables. Second, it allows us to incorporate multilevel effects (i.e. hierarchical effects). Finally, it allow us to incorporate previous knowledge as a prior. For instance, we can include a prior the results obtained in [Smith et al. \[2021\]](#) for the Pacific sardine landing equation.

Specifically, we fit a hierarchical Bayesian hurdle model to model the zeros included in our landing data. We observe a zero when no landings occur in a specific time period (landings observations with “NA” were transformed to zero). This zero could mean that there was no incentives for the fleet/vessel to harvest the species in consideration. In general, our Bayesian models have the following structure:

$$[\theta_i | q_{i,t}] \propto f(q_{i,t} | \theta_i) \times [\theta_i]$$

where $q_{i,t}$ is the observed landings of the corresponding species in port $i \in (1, \dots, L)$ at year t , L is the total number of port, and θ_i are the parameters (i.e. random-coefficients) to be estimated at the port level. The latter give the name of hierarchical to our model.³ The distribution $f(q_{i,t} | \theta_j)$ can be rewritten as:

$$f(q_{i,t} | \theta_j) = \begin{cases} p_i & \text{if } q_{it} = 0 \\ [1 - p_i] \text{gamma}\left(q_{i,t} | \frac{\mu^2}{\sigma^2}, \frac{\mu}{\sigma^2}\right) & \text{if } q_{it} > 0. \end{cases} \quad (1)$$

where $\text{logit}(p_{i,j}) = \mathbf{X}\gamma$ and $\mu = \mathbf{X}\beta$. Specifically μ is defined as the following:

$$\mu_{i,j} = \beta_{0,i} + \beta_{1,j}P(\text{Precense})_{PSDN} + \beta_{2,j} + P(\text{Precense})_{MSQD} + \beta_{3,j}P(\text{Precense})_{NANC}. \quad (2)$$

where j indicate the species in consideration. $\text{logit}(p_{i,j})$ follows the same structure, but coefficient are replaced by γ .

³For more details about Hierarchy models, see [Hobbs and Hooten \[2015\]](#).

4 Results

4.1 Pacific Sardine model

Pacific Sardine model

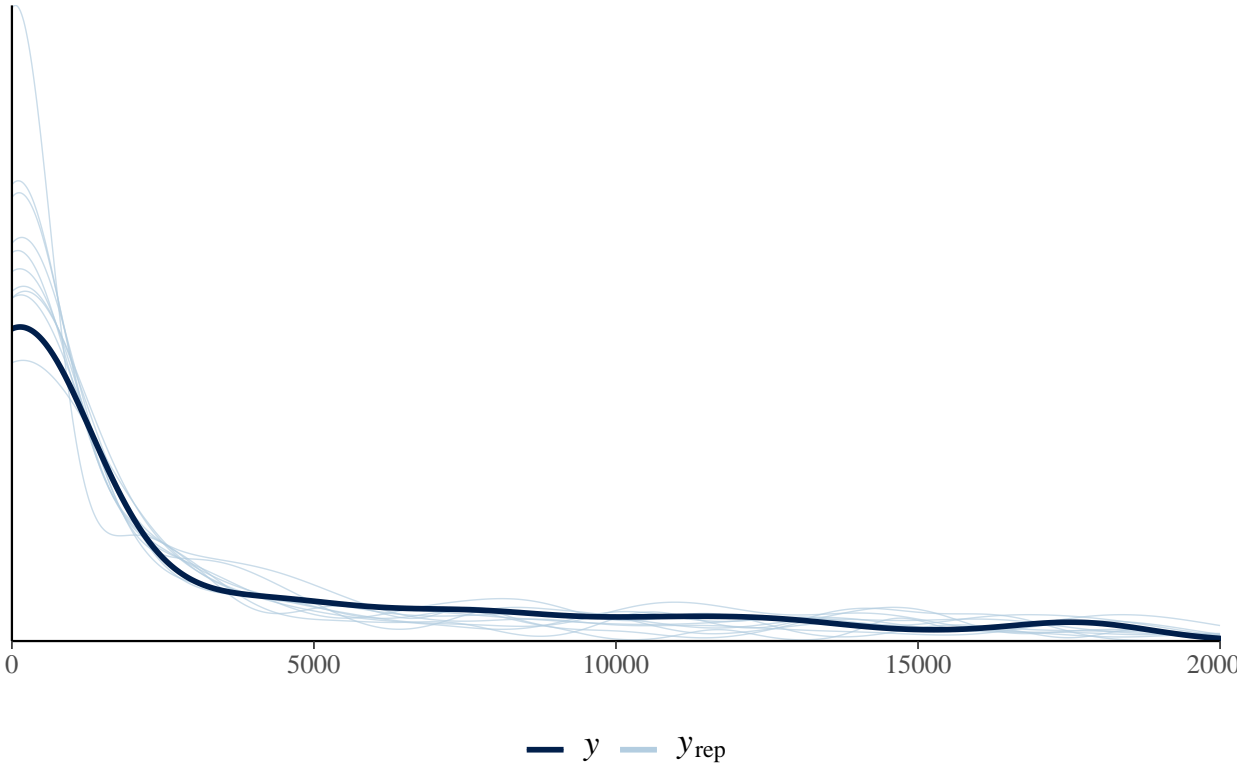
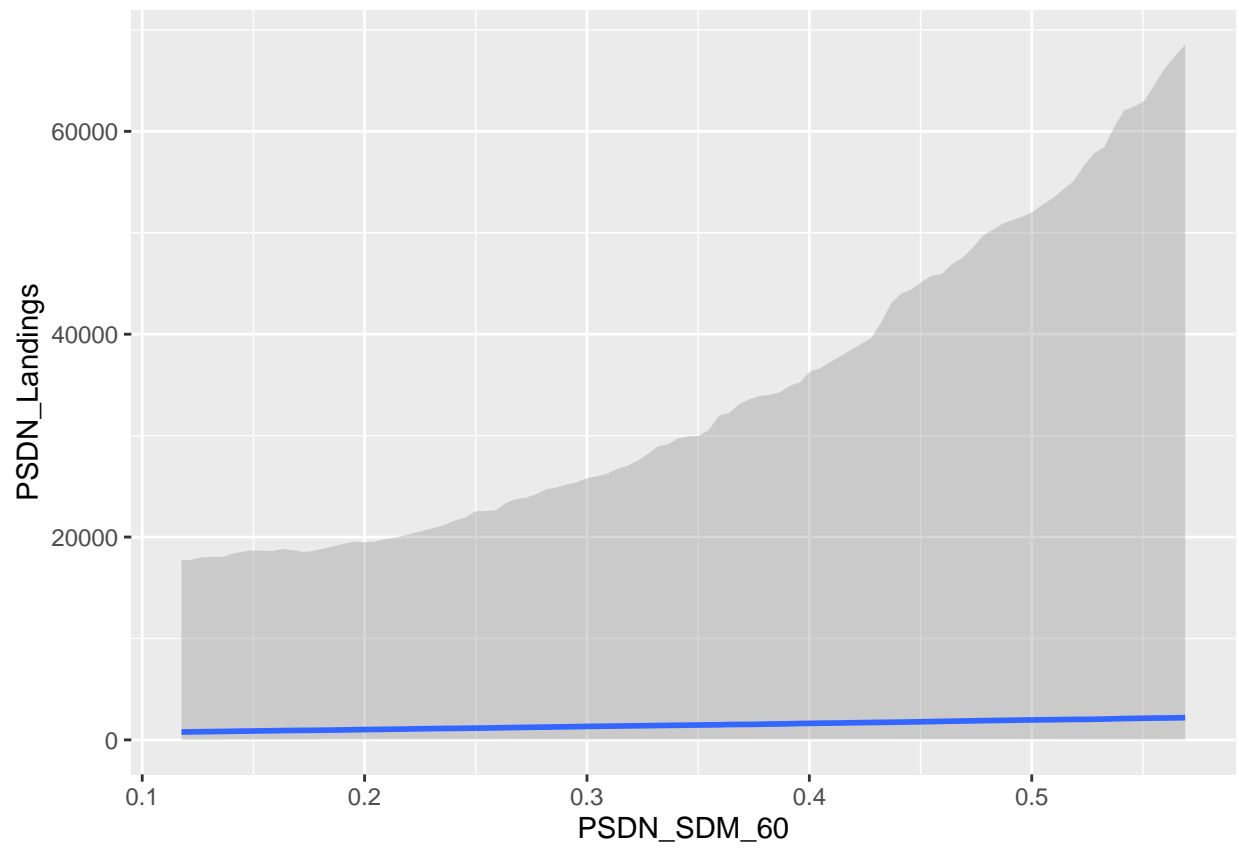
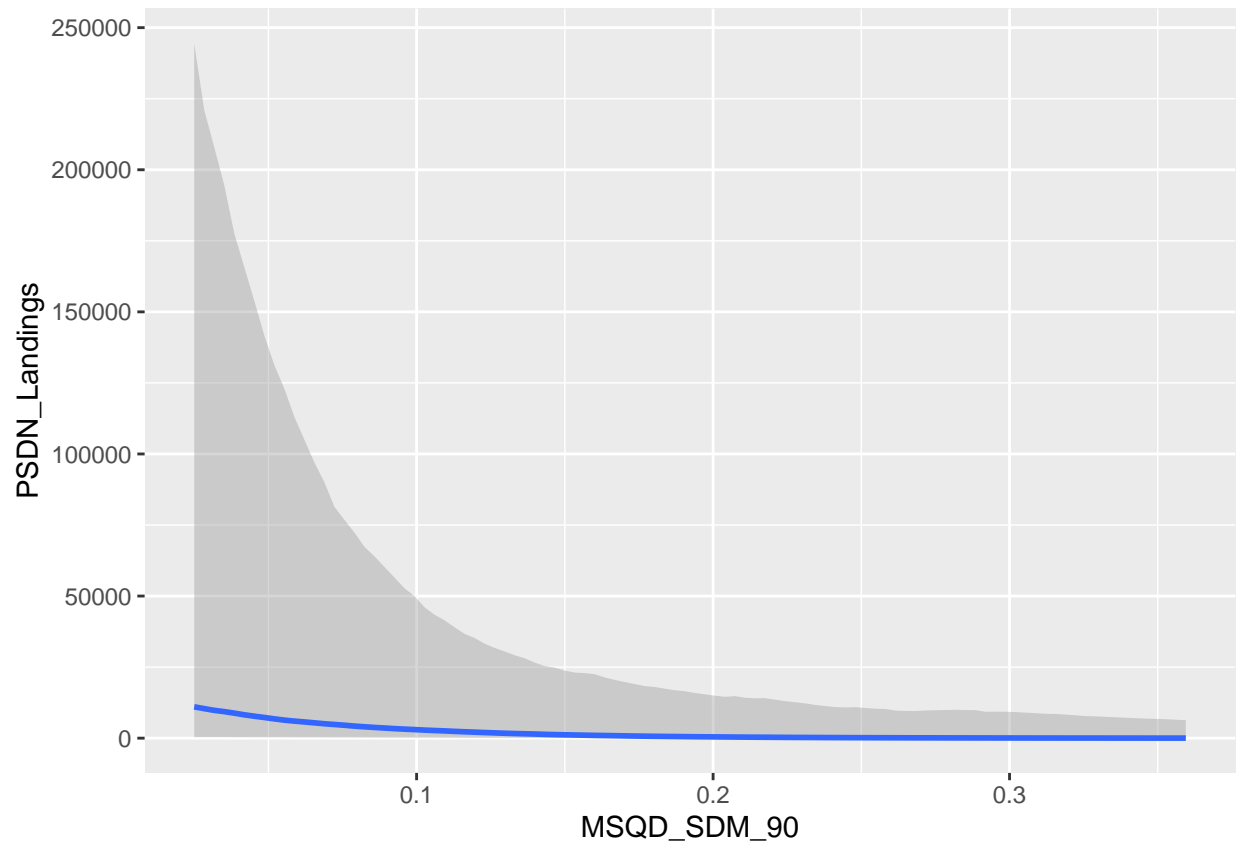
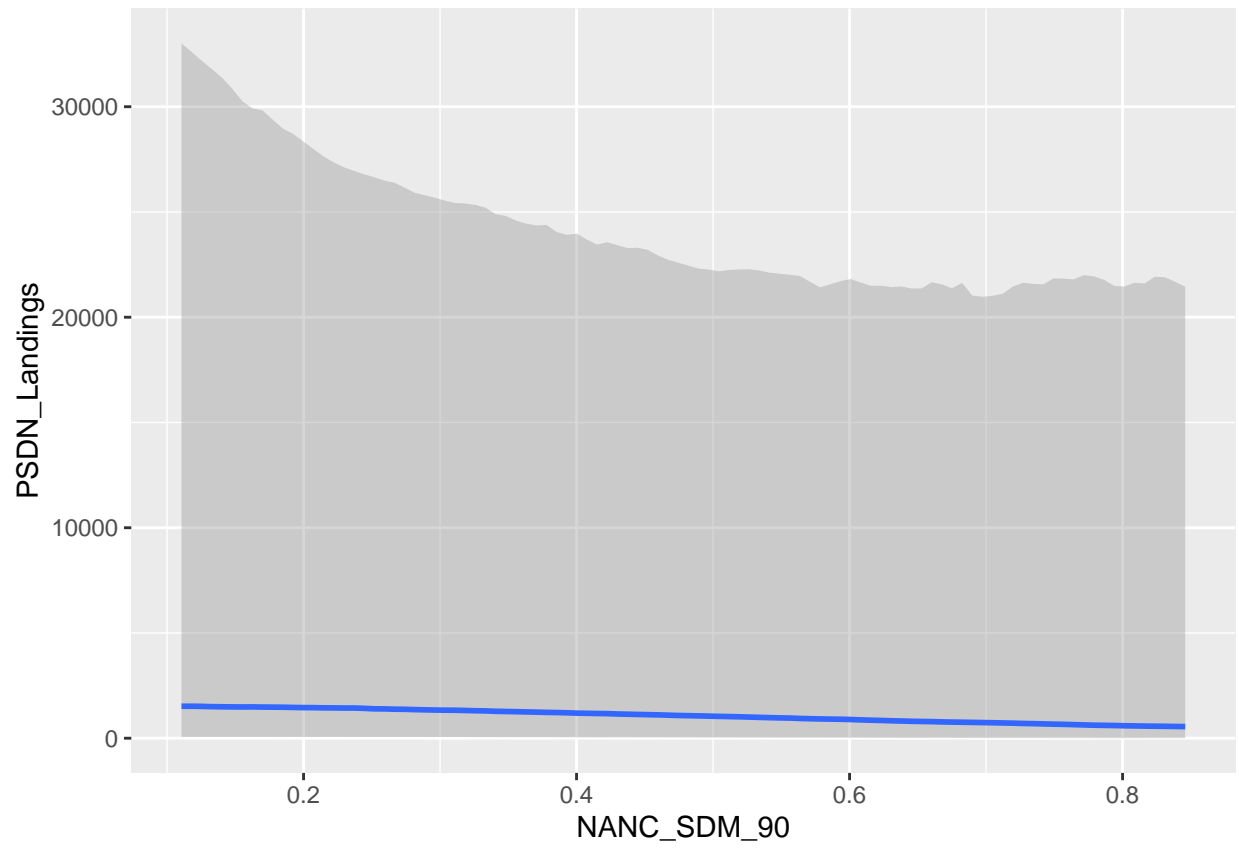


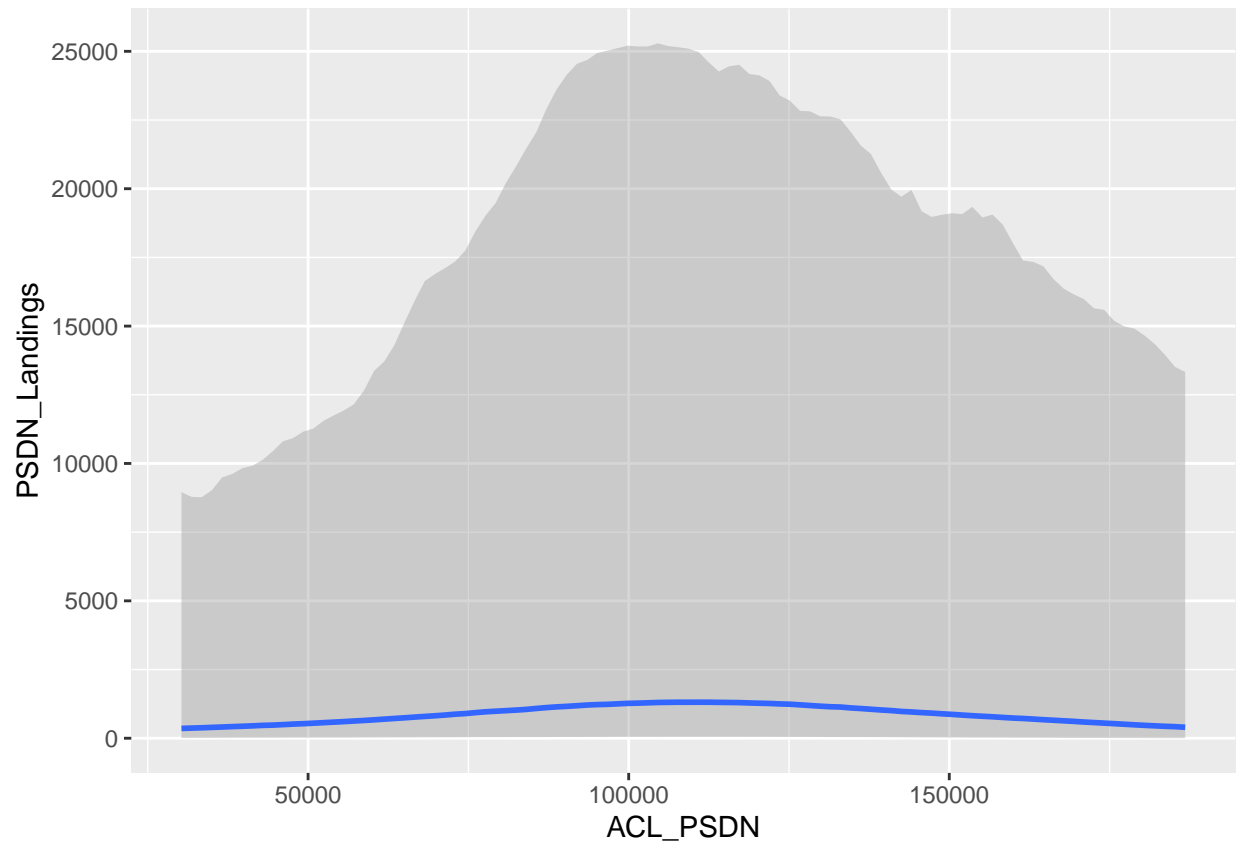
Figure 6: Graphical posterior predictive checks. (Gamma distribution)

4.1.1 Effect of SDM's on Sardine landings

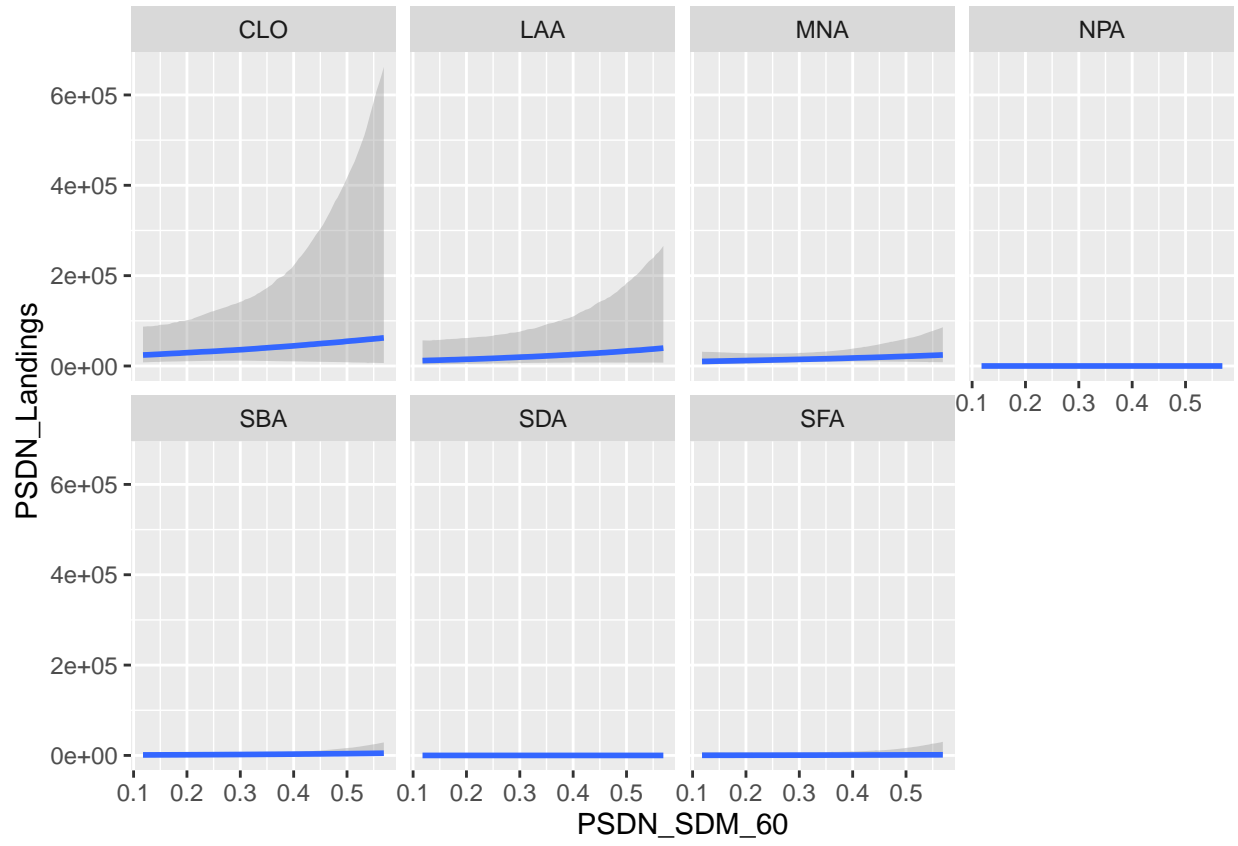


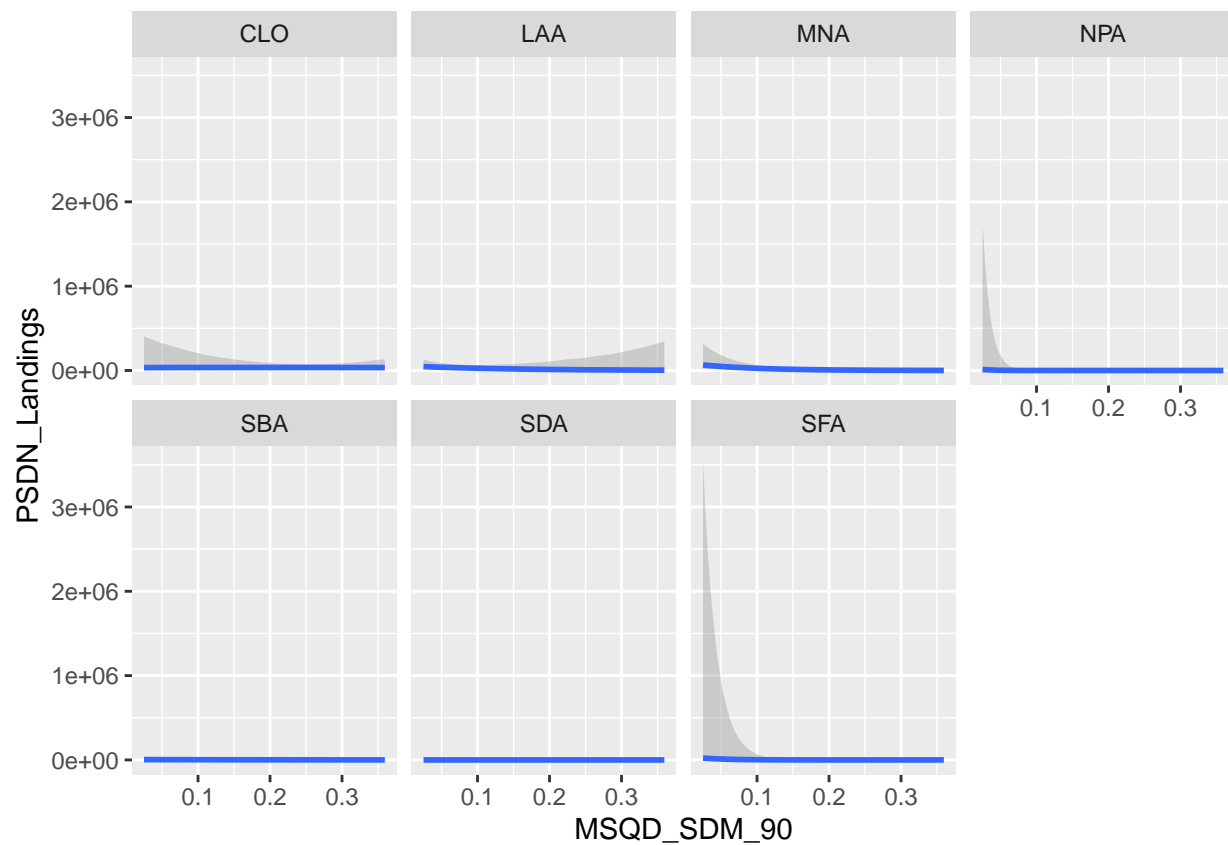


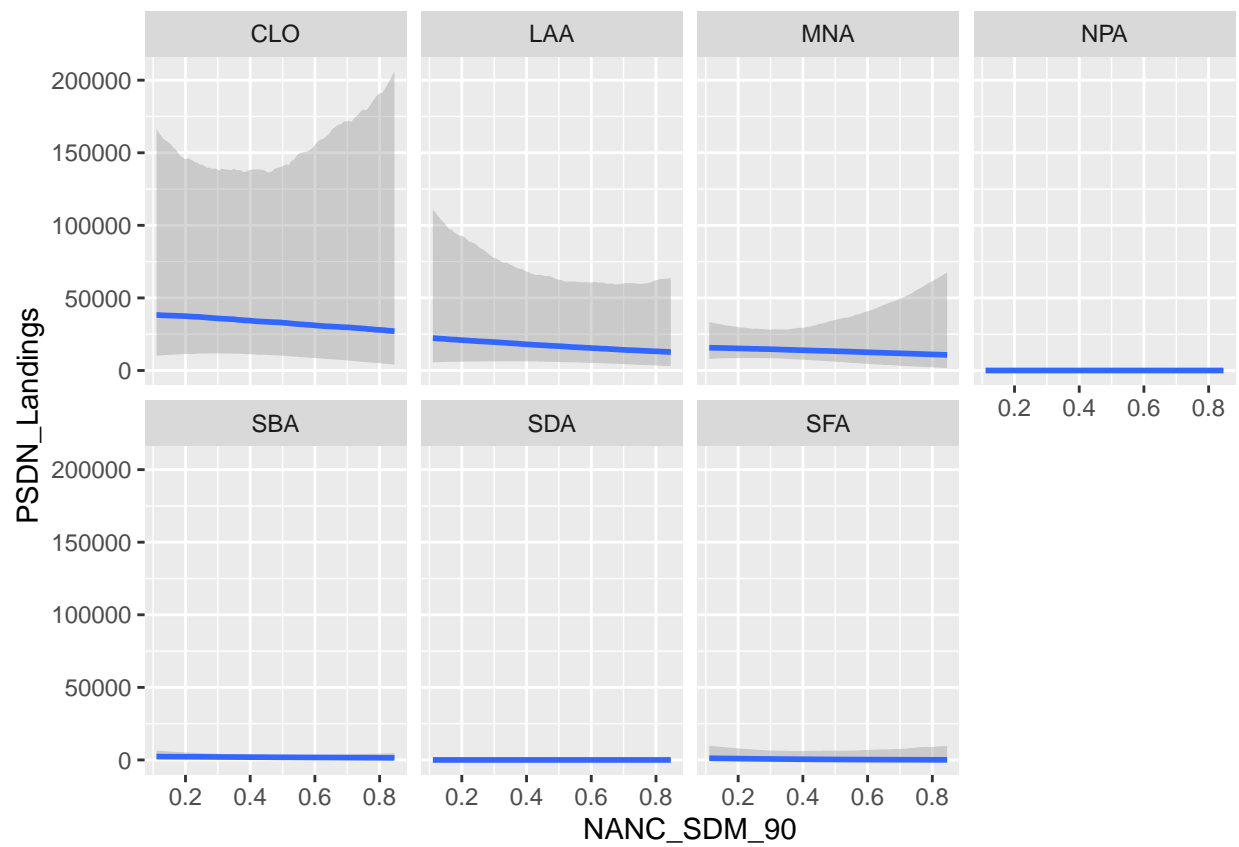


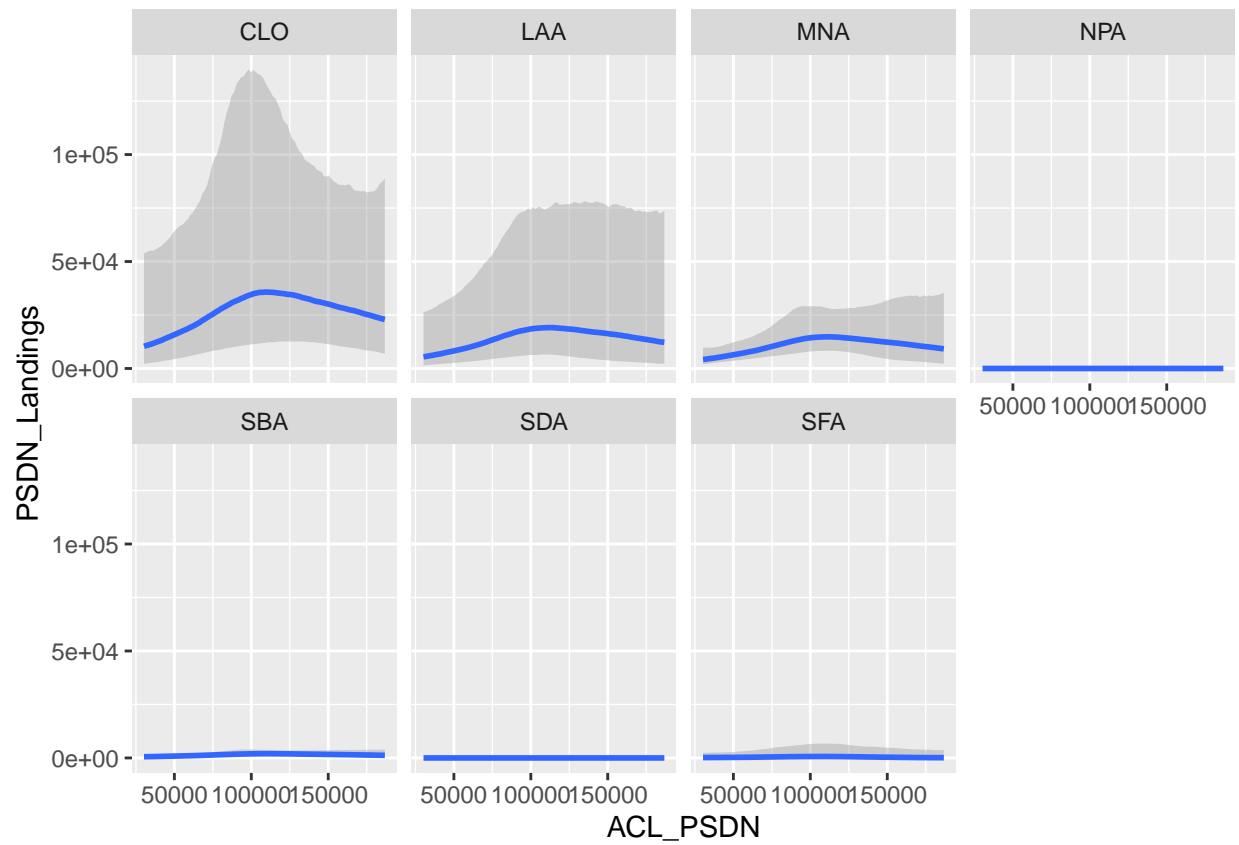


4.1.2 Results by port

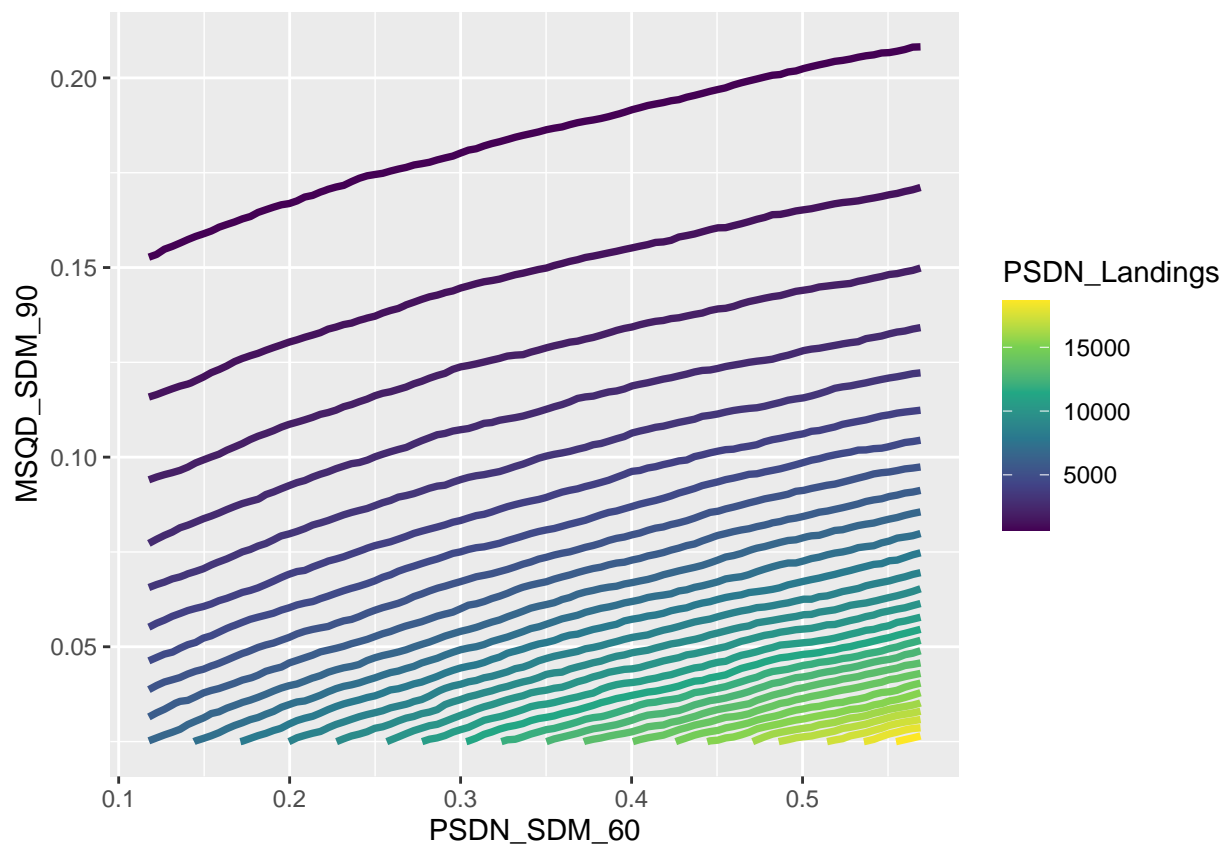


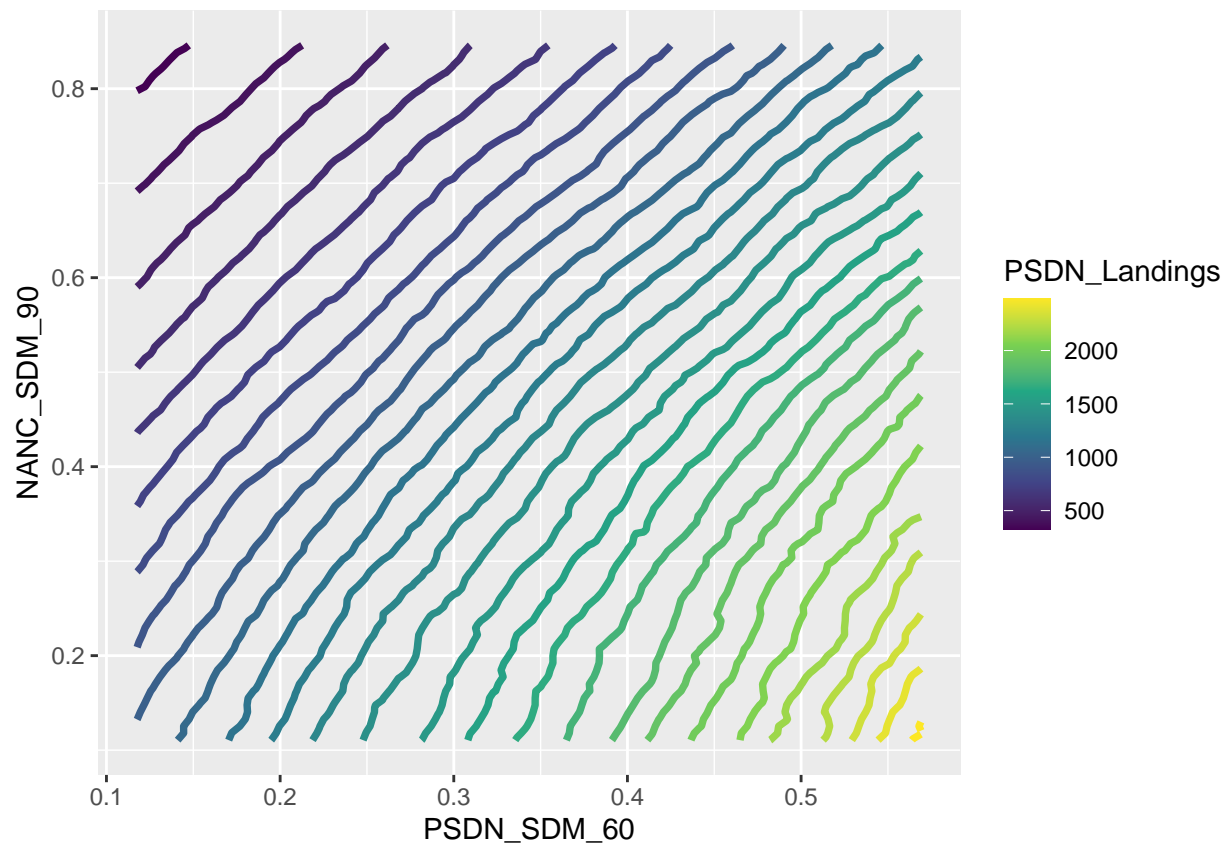


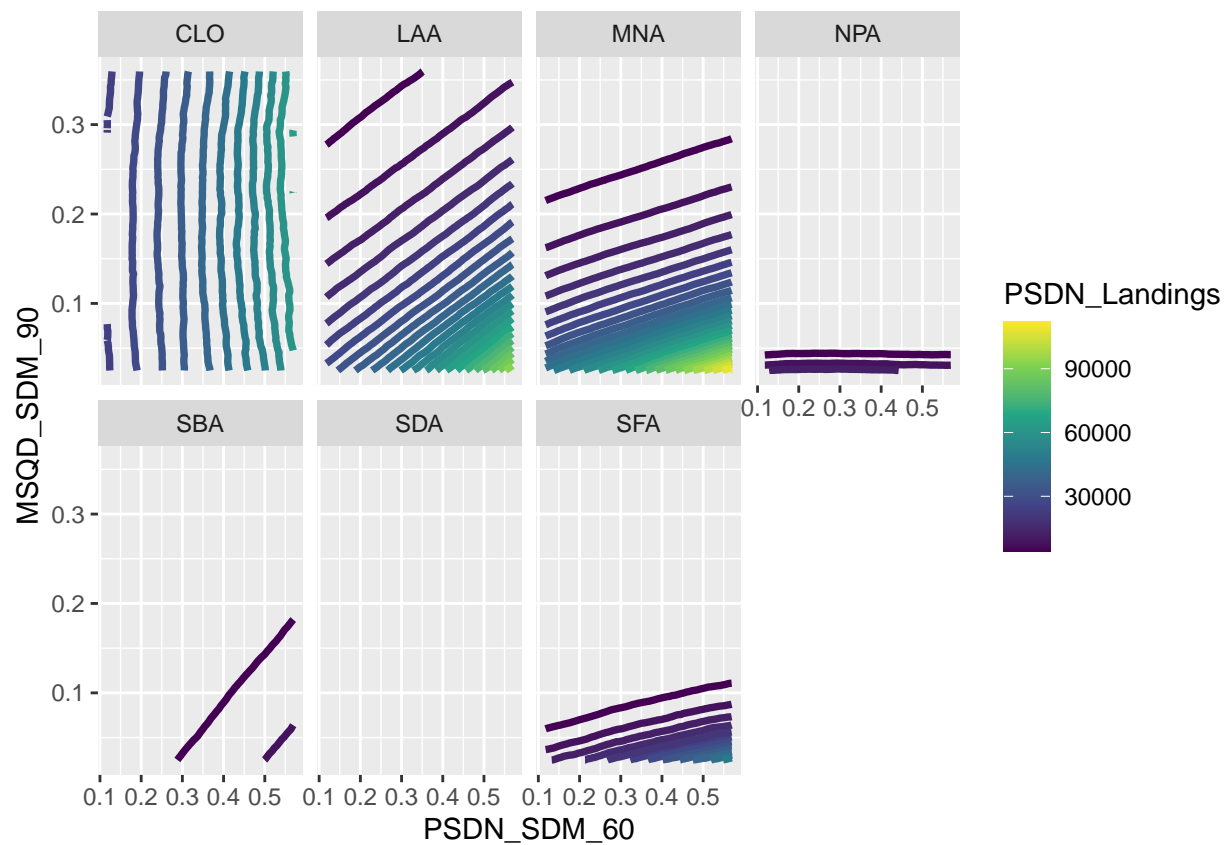


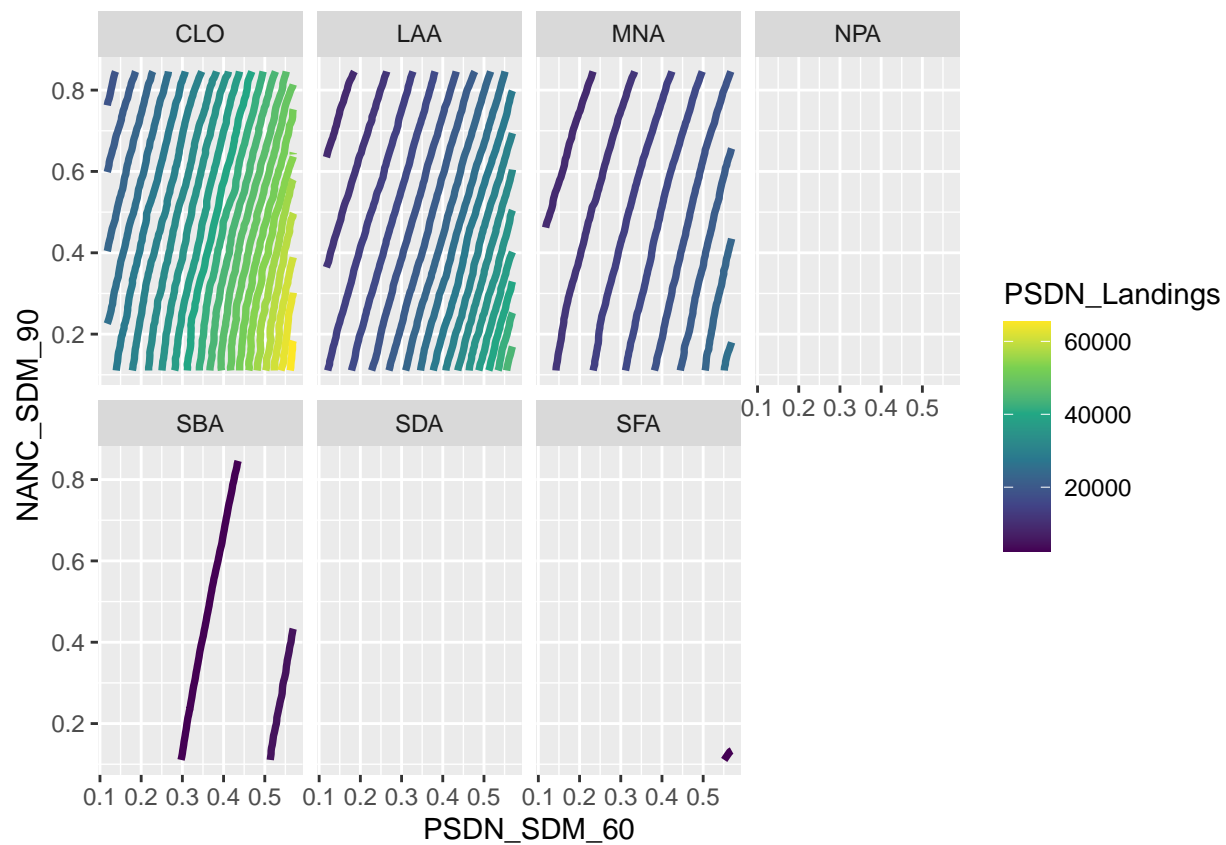


4.1.3 Interactive effects between Pacific Sardine and Market Squid SDMs









4.2 Market Squid model

4.2.1 Compare models

Market Squid landing model

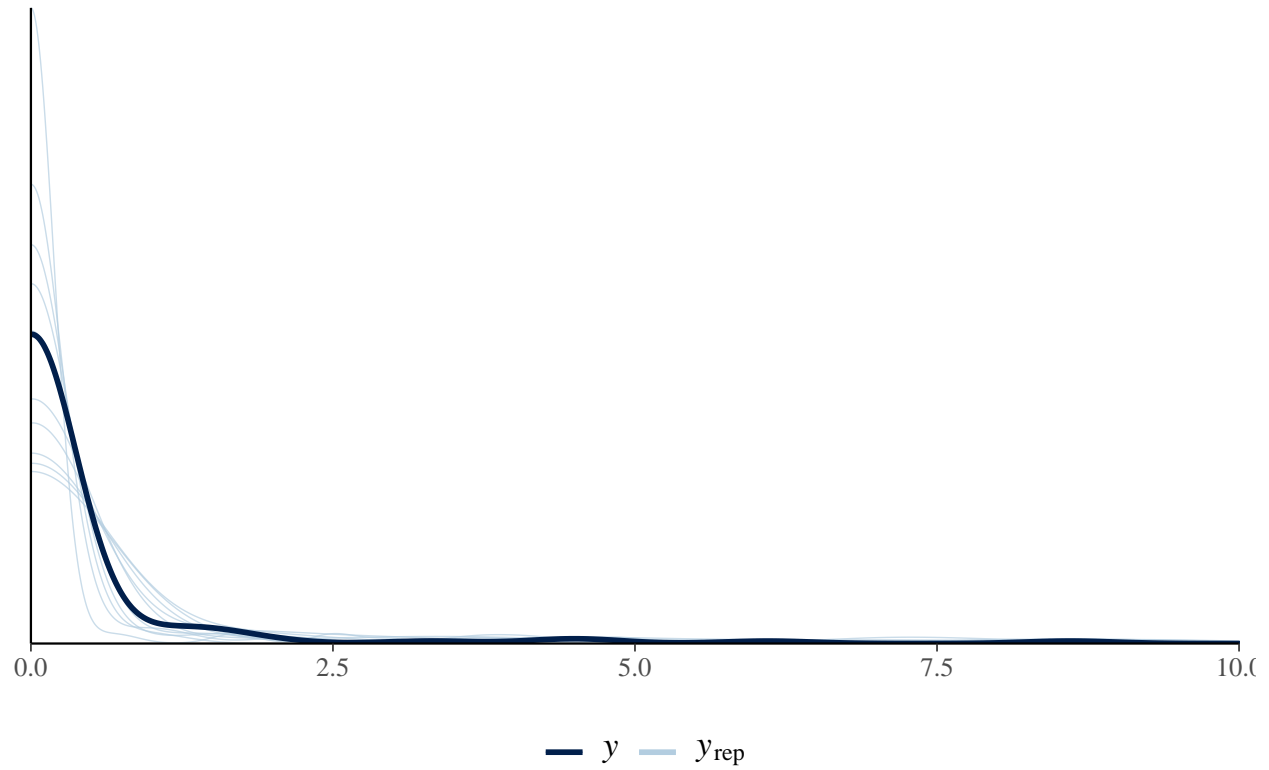


Figure 7: Graphical posterior predictive checks for Market Squid landings model.

4.2.2 Effect of SDM's on Market SquidSardine landings

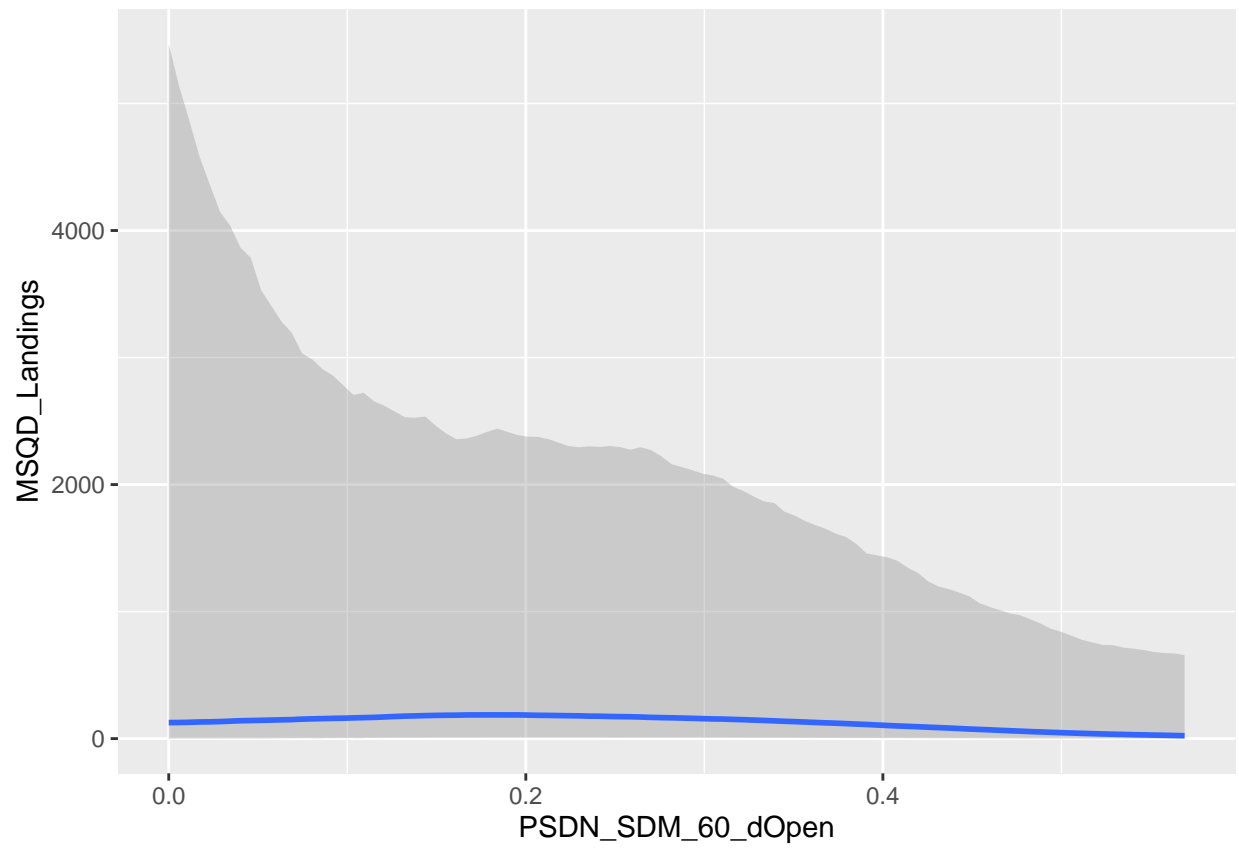


Figure 8: Conditional effect of probability of occurrence of Pacific Sardine on Squid landings.

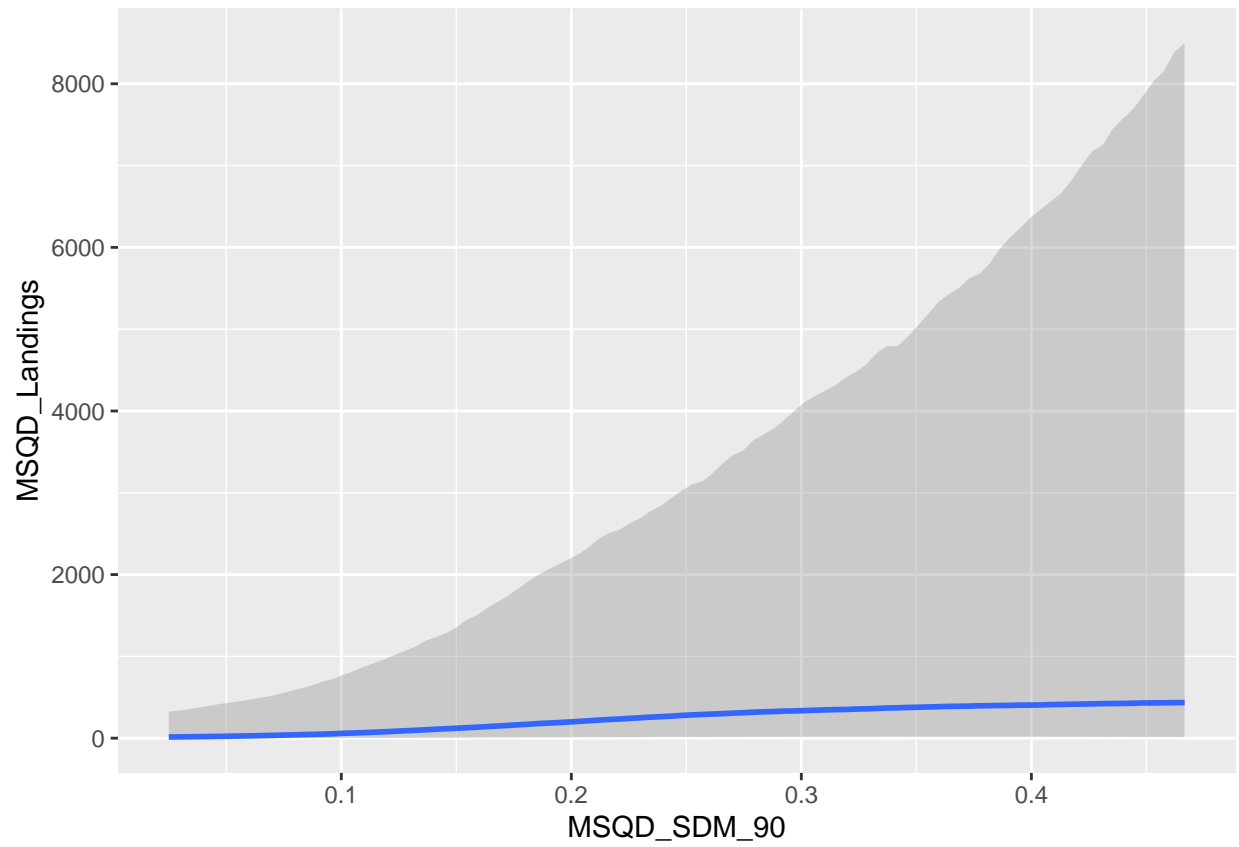


Figure 9: Conditional effect of probability of occurrence of Market Squid on Squid landings.

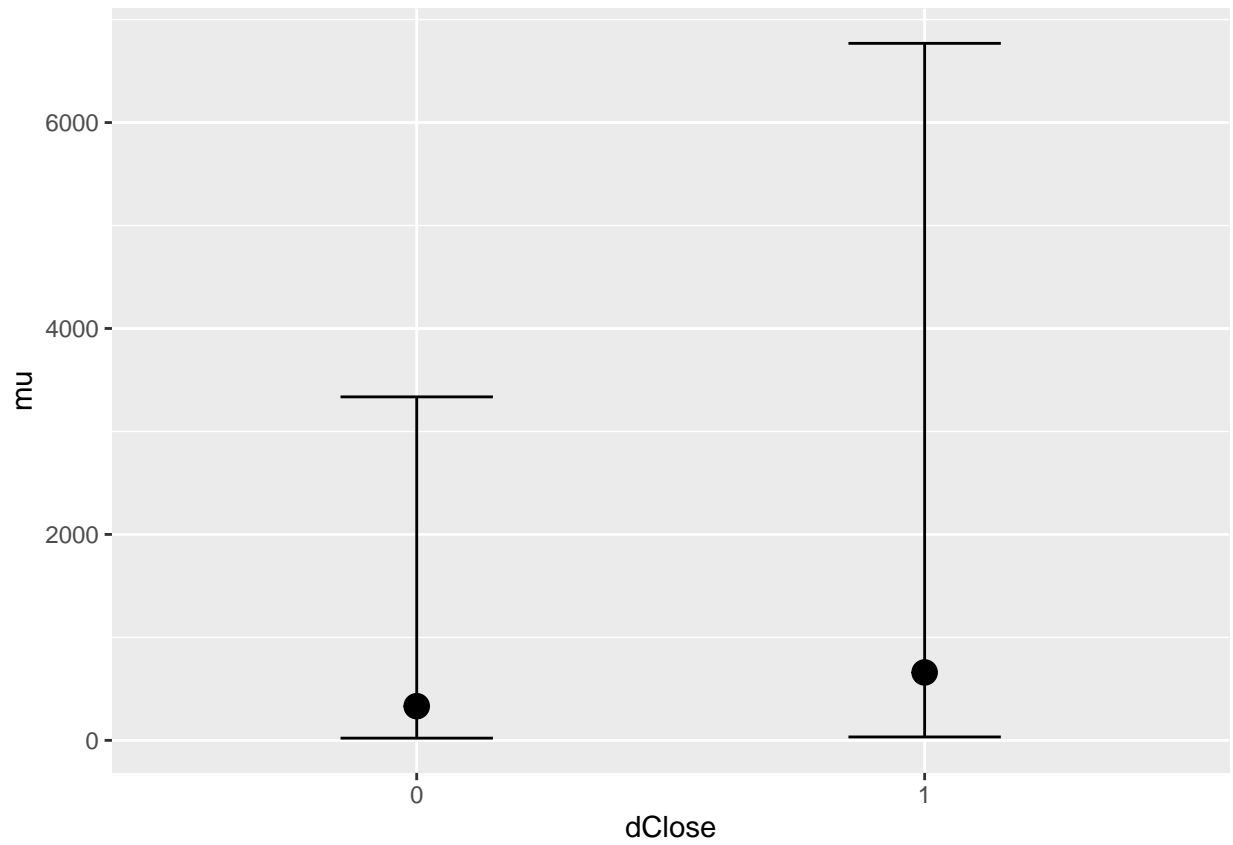
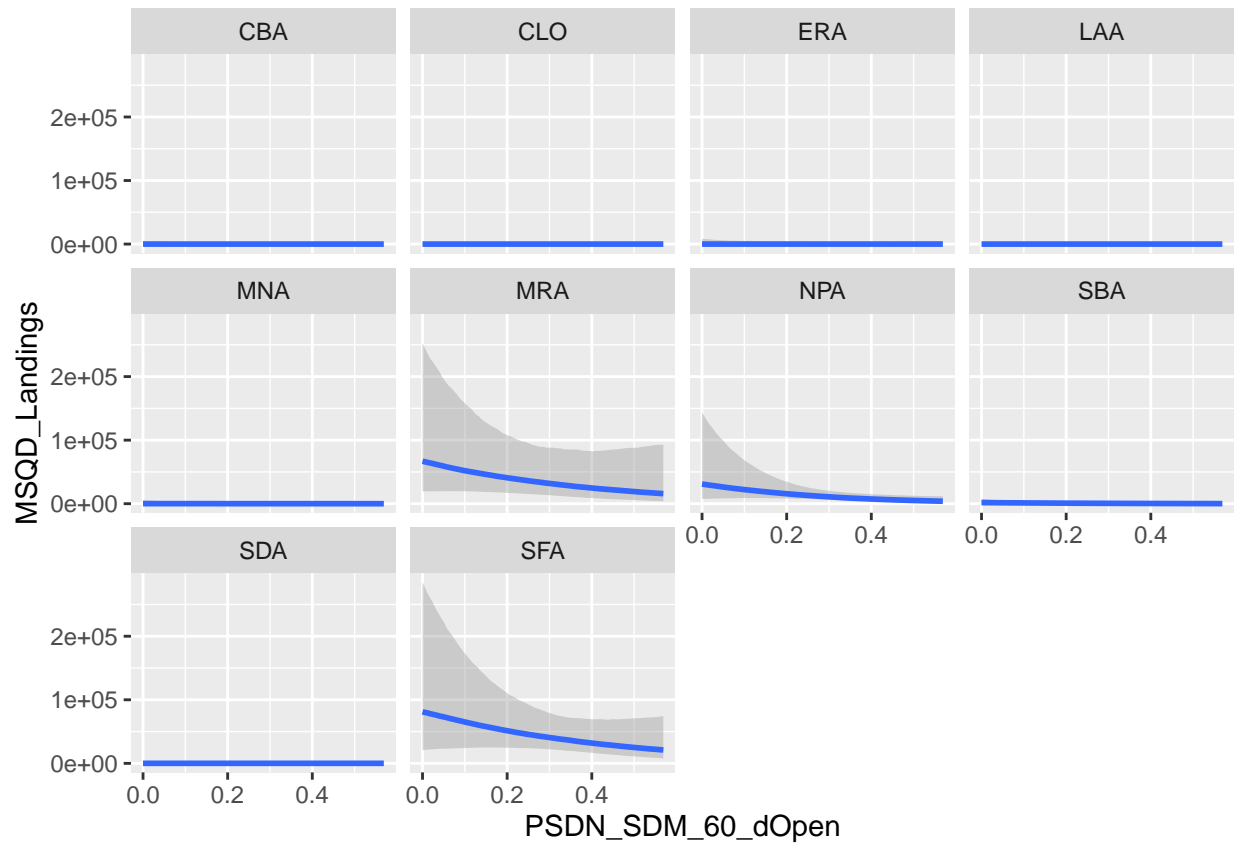
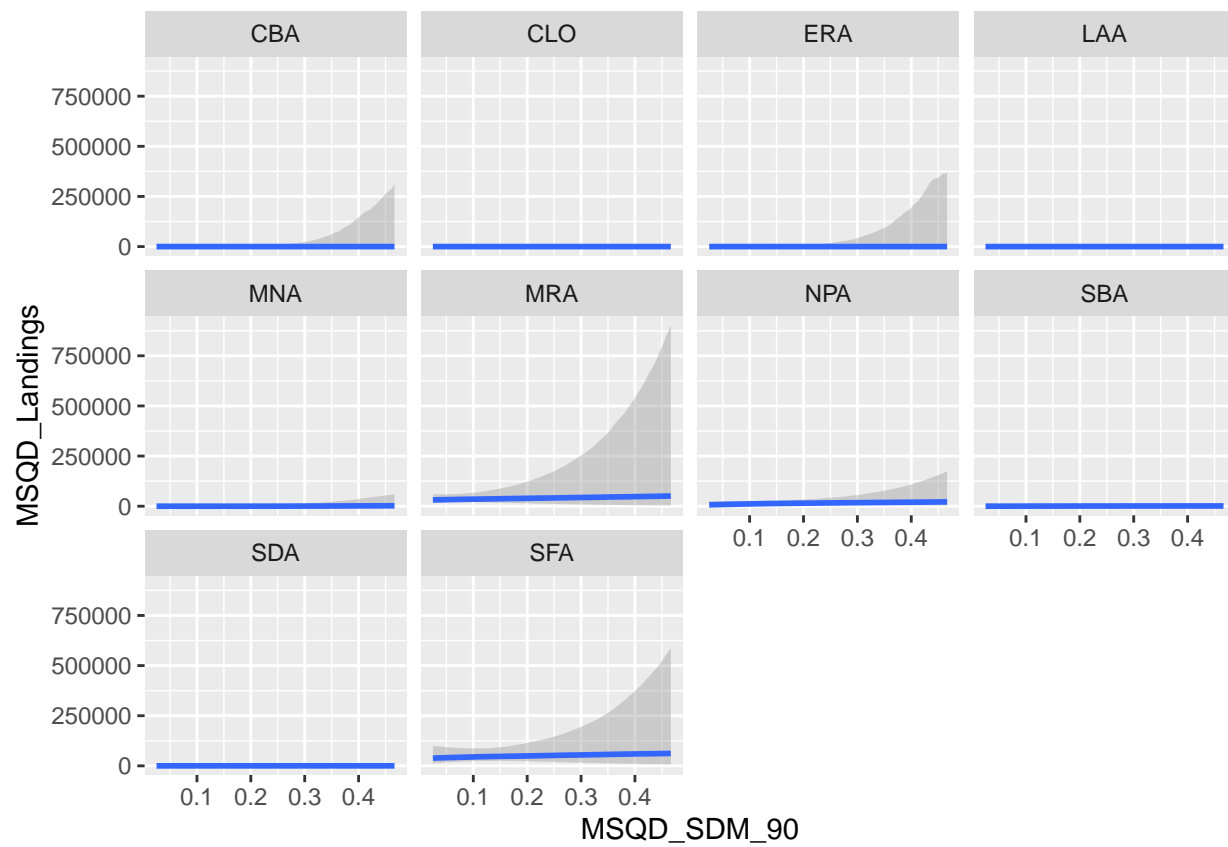
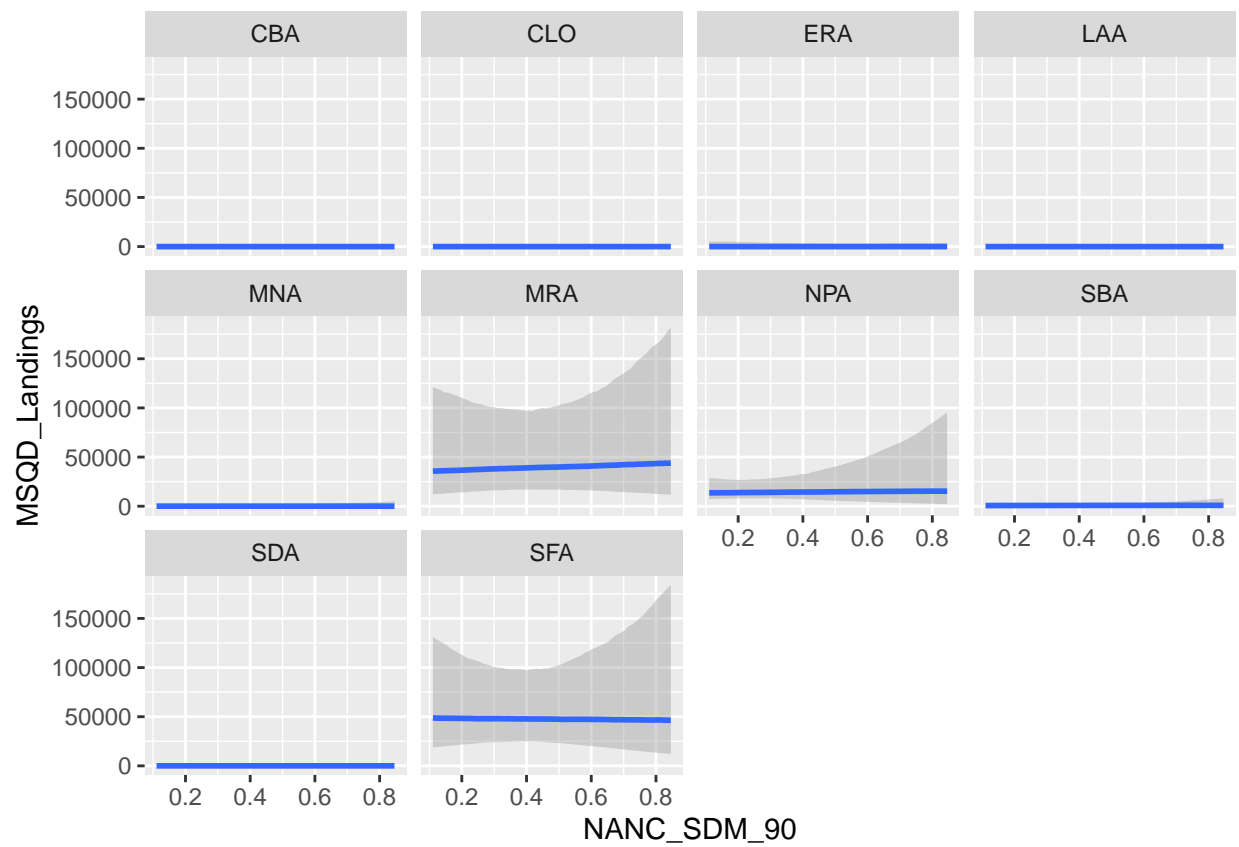


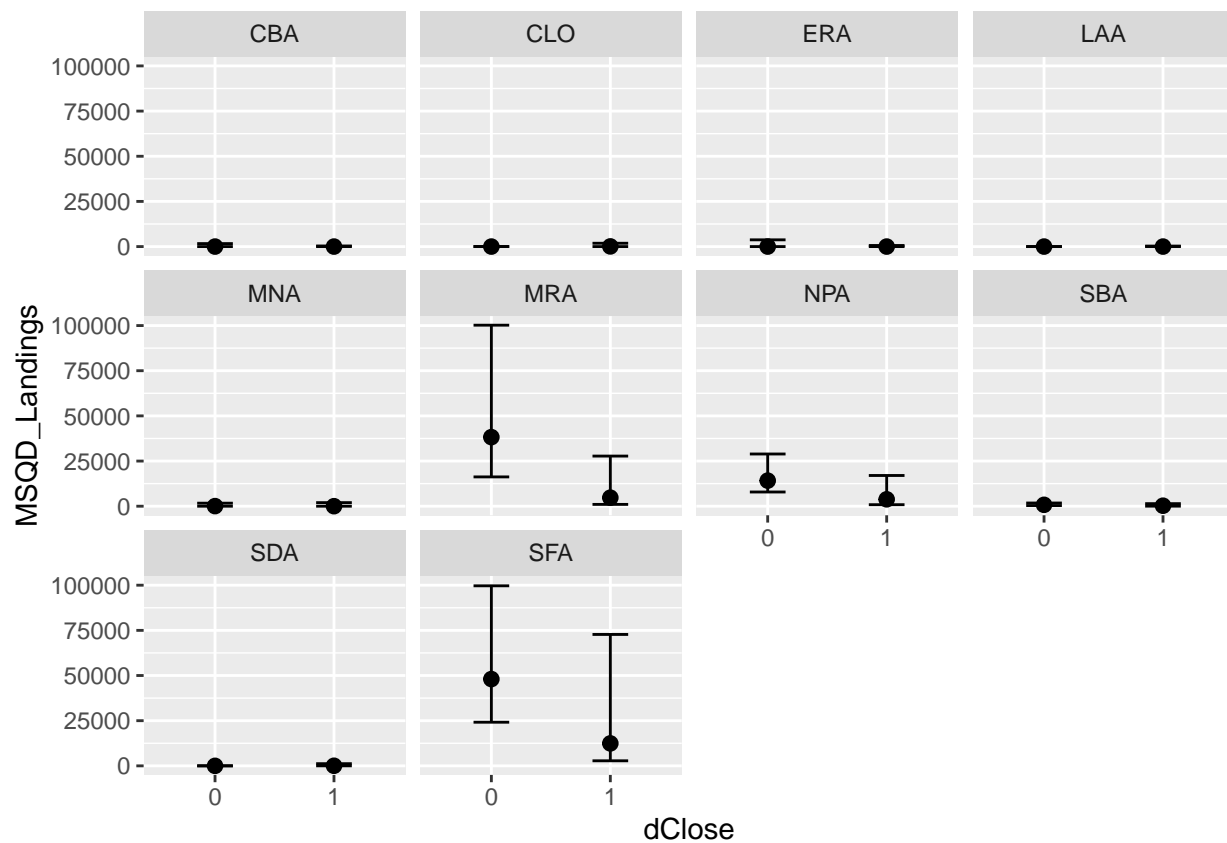
Figure 10: Conditional effect of Pacific Sardin Closure on Market Squid Landings

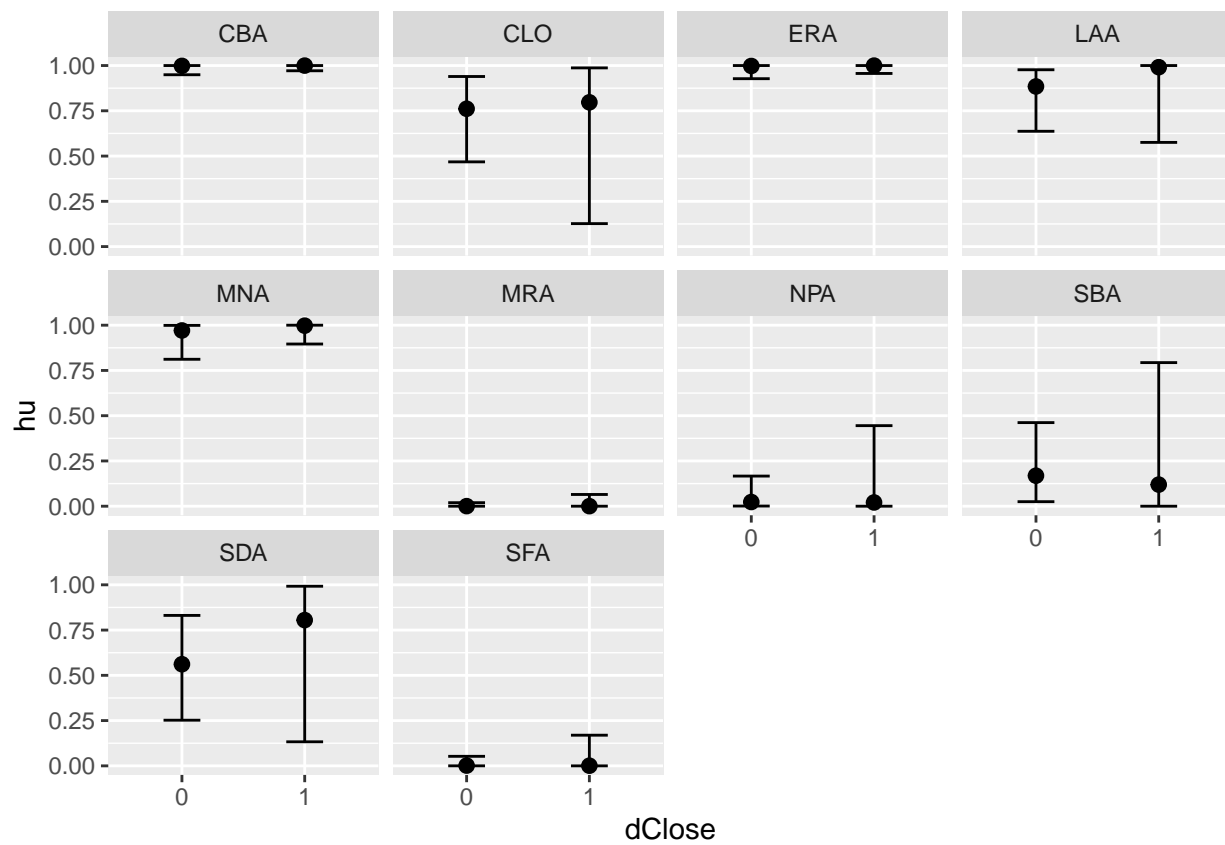
4.2.3 Results by port

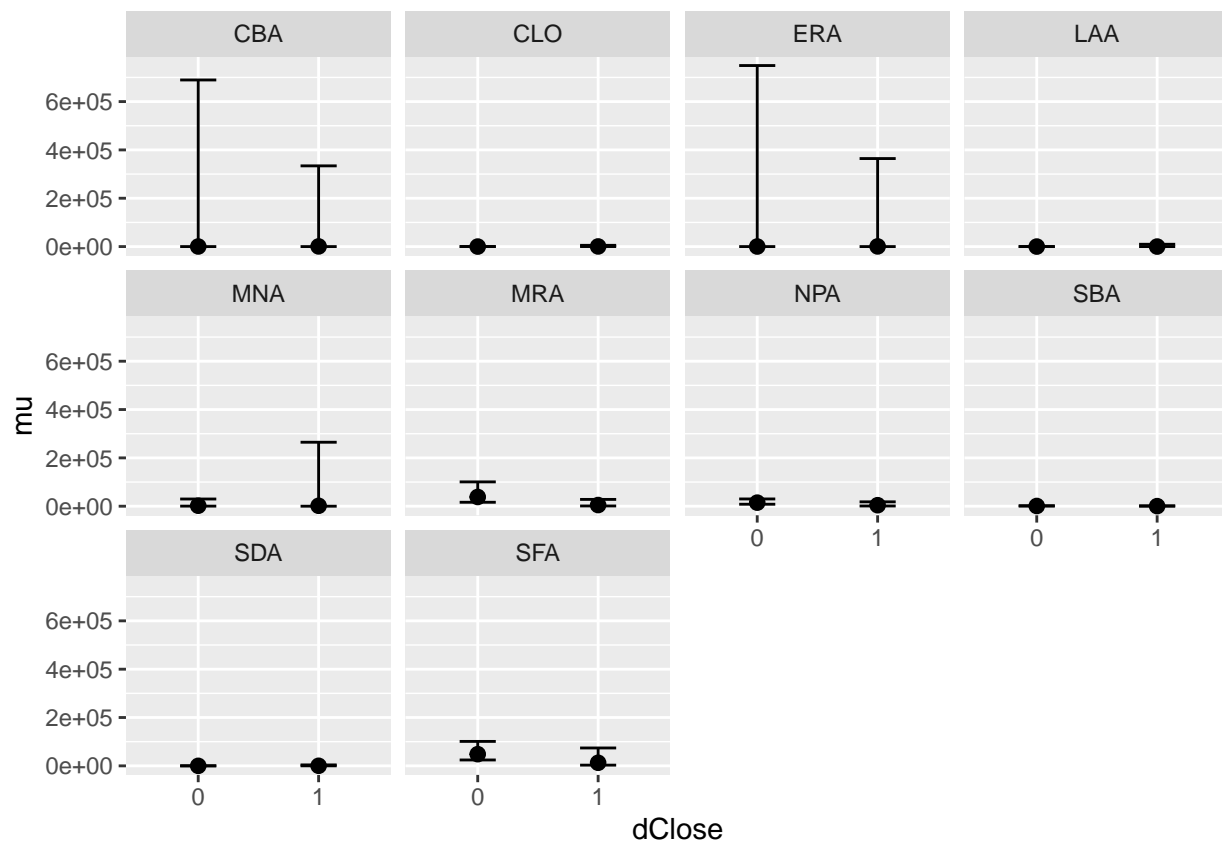




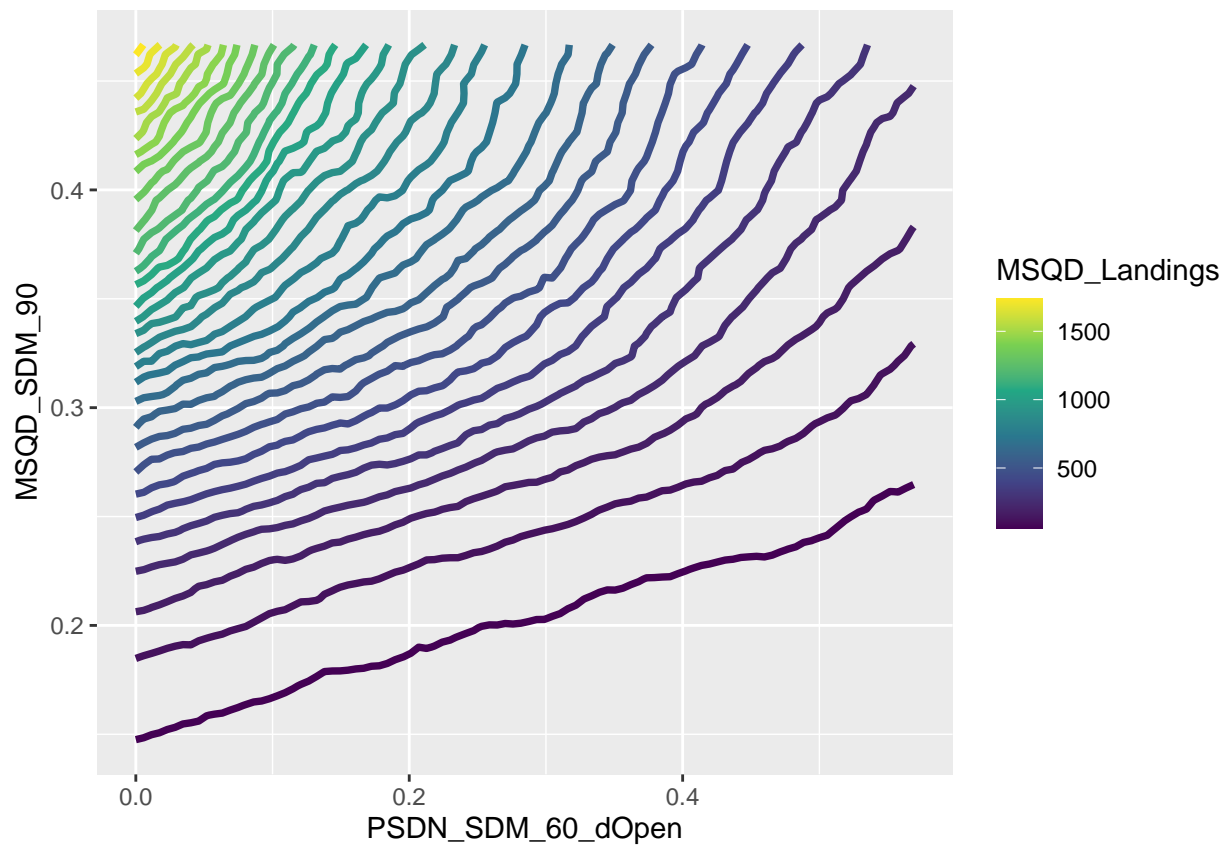


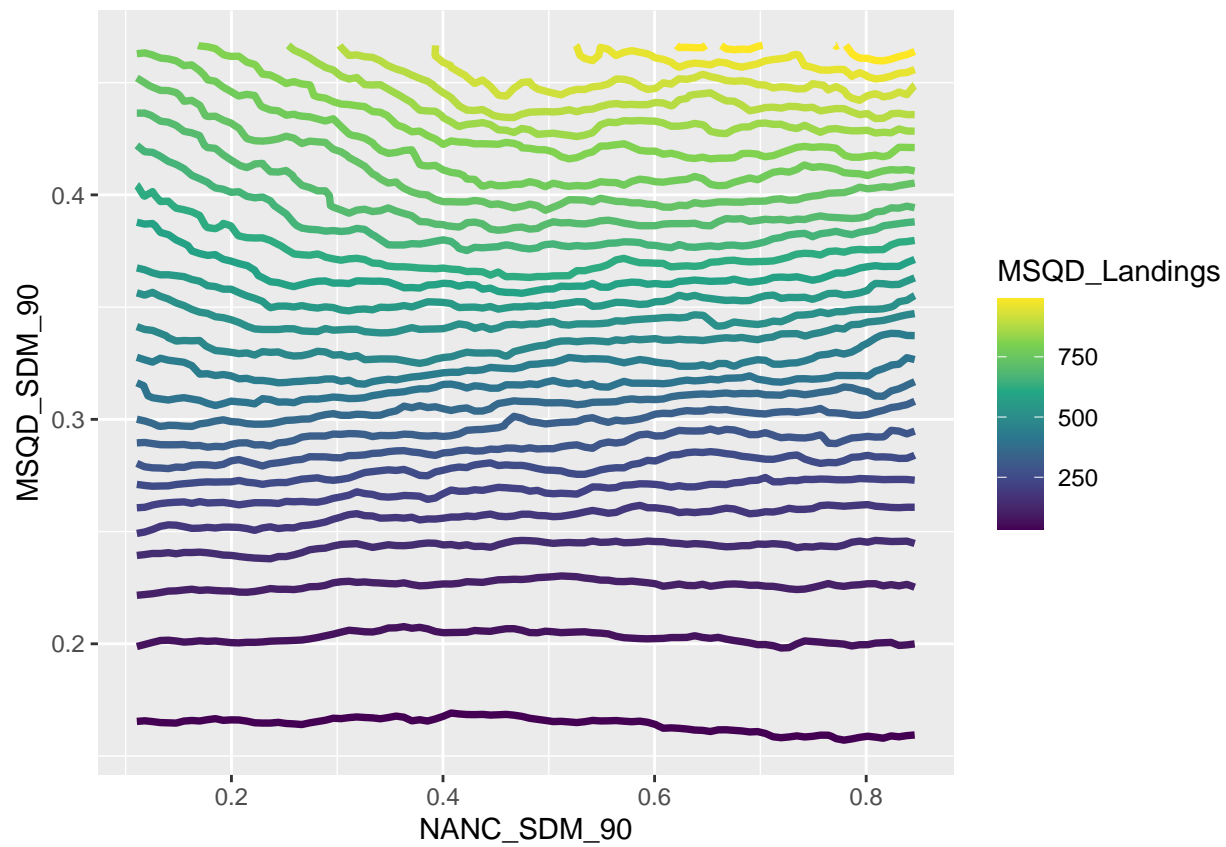


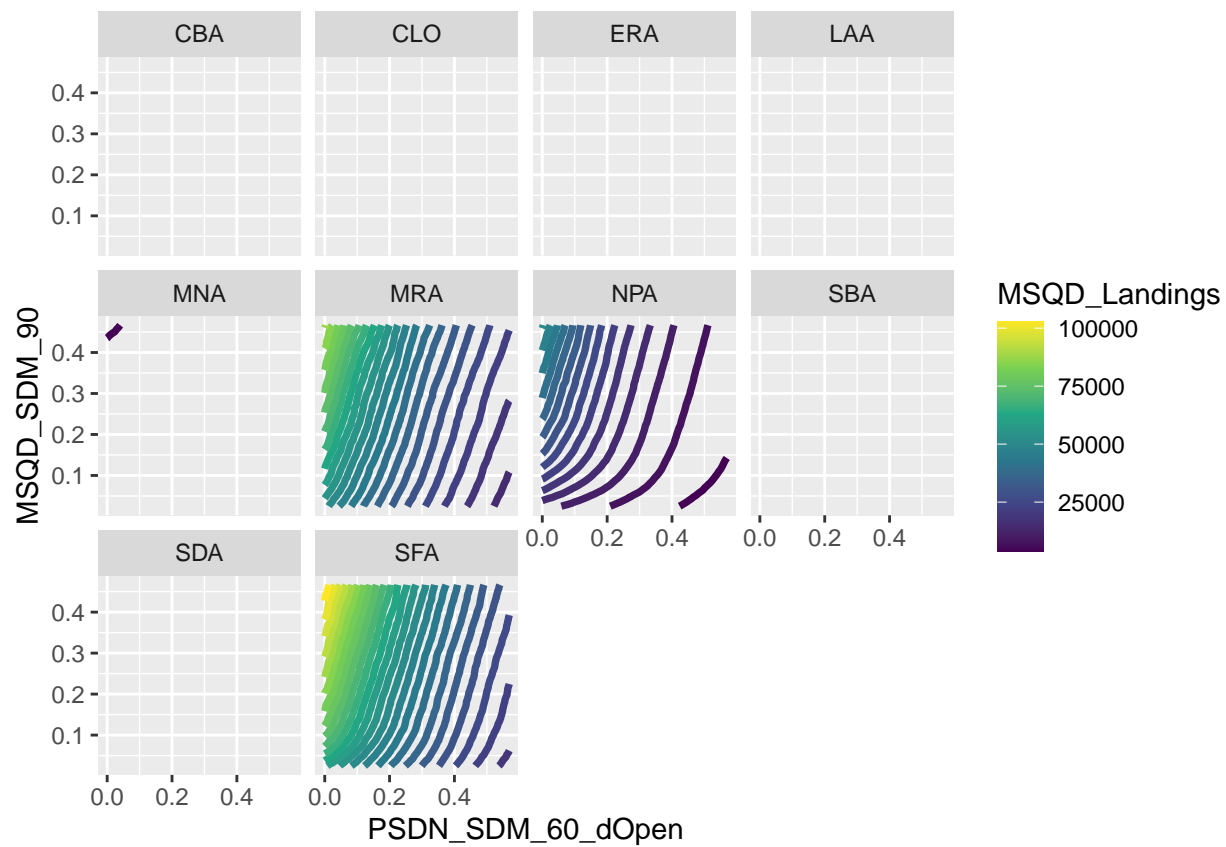


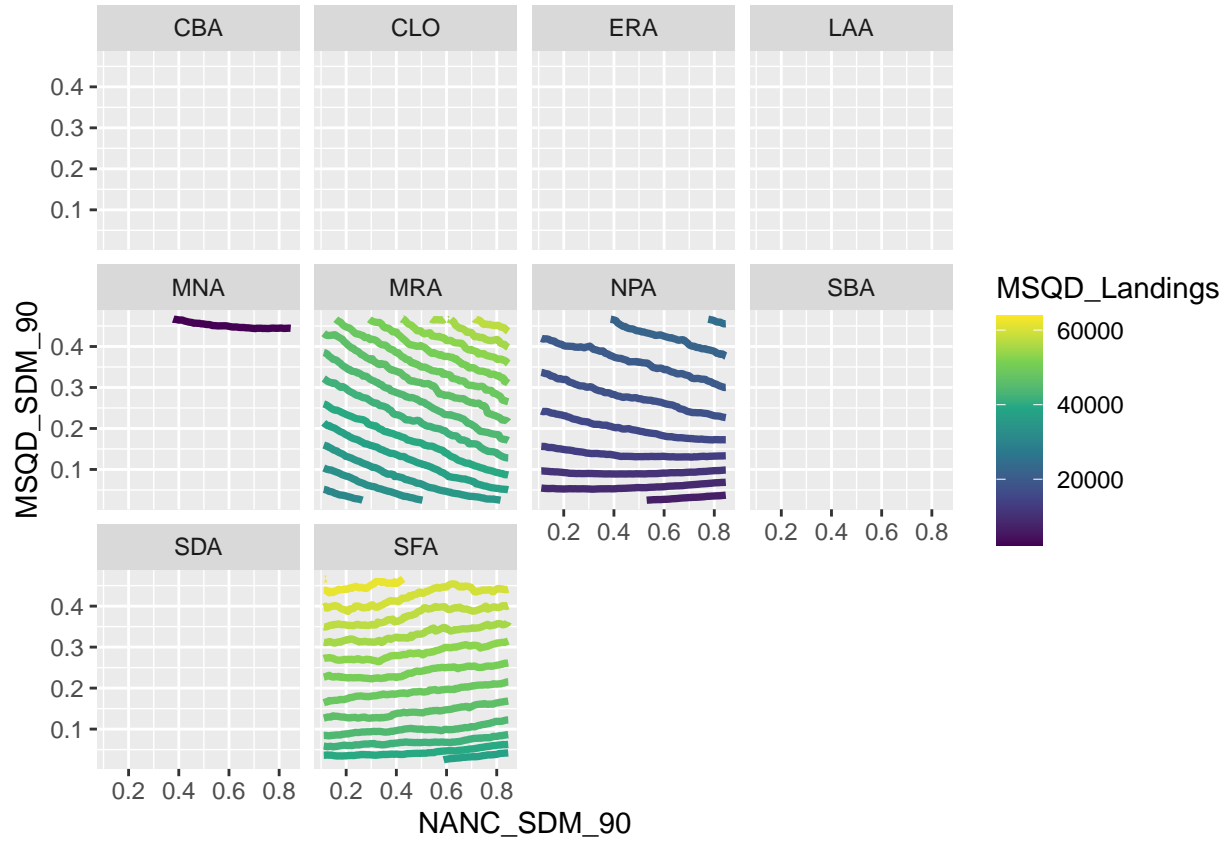


4.2.4 Interactive effects between Pacific Sardine and Market Squid SDMs









5 Conclusions

A possible extension of our research is to consider spatial autocorrelations between ports.⁴ Ports landing maybe correlated as vessel have the incentives to choose the port of landing, conditional on whether the port have the infrastructure for this. It is likely that they just land wherever is closer to the area they are fishing. Nevertheless, differential in prices could encourage them to travel a little further for higher prices.

References

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⁴See [Morris et al. \[2019\]](#) for an application of a spatial model in a Bayesian framework.

James A Smith, Barbara Muhling, Jonathan Sweeney, Desiree Tommasi, Mercedes Pozo Buil, Jerome Fiechter, and Michael G Jacox. The potential impact of a shifting pacific sardine distribution on us west coast landings. *Fisheries Oceanography*, 2021.