#### Comisión Interamericana del Atún Tropical Inter-American Tropical Tuna Commission



Spatiotemporal dynamics of the dolphin-associated purse-seine fishery for yellowfin tuna in the Eastern Pacific Ocean (SAC-09-09) Haikun Xu, Cleridy Lennert-Cody, Mark Maunder, and Carolina Minte-Vera

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Spatiotemporal dynamics of the dolphin-associated purse-seine fishery for yellowfin tuna in the Eastern Pacific Ocean

- Background
- Data and Methods
- Results
- Discussion





• Goal: develop a standardized purse-seine index of abundance for yellowfin tuna in the Eastern Pacific Ocean

 Importance: index of abundance directly informs trend in population biomass and is a key input in the stock assessment

 Difficulties: index of abundance for yellowfin is derived solely from fishery-dependent CPUE data –> preferential sampling and sparse spatial coverage



Dolphin-associated purse-seine fishery for yellowfin

- Takes the largest catch of yellowfin in the EPO
- Contributes to two of the five indices of abundance used in the stock assessment
   10° 10° 10° 10° 9° 8° 7°





#### Data

- Per-vessel catch (in metric tons) and effort (in days fishing) data from the vessels with >75% dolphin-associated sets
- Spatial resolution of 1° x 1° and temporal resolution of 1 day during 1975-2016
- Nominal CPUE = catch / effort

Spatial coverage of the dolphin-associated fishery by quarter



delta-generalized linear mixed model: VAST (Thorson and Barnett, 2017)

separately models encounter probability (*p*) and positive catch rate ( $\lambda$ ):  $p_i = logit^{-1} (\beta_1(t_i) + L_{\omega_1}\omega_1(s_i) + L_{\varepsilon_1}\varepsilon_1(s_i, t_i) + L_{\delta_1}\delta_1(v_i))$  $\lambda_i = \exp(\beta_2(t_i) + L_{\omega_2}\omega_2(s_i) + L_{\varepsilon_2}\varepsilon_2(s_i, t_i) + L_{\delta_2}\delta_2(v_i))$ 

 $\beta_1(t_i) \& \beta_2(t_i)$ : intercept in year  $t_i$   $\omega_1(s_i) \& \omega_2(s_i)$ : spatial variation at location  $s_i$   $\varepsilon_1(s_i, t_i) \& \varepsilon_2(s_i, t_i)$ : spatiotemporal variation at location  $s_i$  in year  $t_i$  $\delta_1(v_i) \& \delta_2(v_i)$ : effect of vessel  $v_i$  on catchability



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Autocorrelated spatial and spatiotemporal residuals:

\omega_1 \sim MVN(0, R_1)

\omega_2 \sim MVN(0, R_2)

\varepsilon_1(t) \sim MVN(0, R_1)

\varepsilon_2(t) \sim MVN(0, R_2)
```

where





#### Results: quarter1 as an example



#### Results: quarter1 as an example



# Results: combine all four quarters



Due to a large seasonal variation in fishing locations, each quarter's data were fitted to the spatiotemporal model separately.

Quarter 1-Trimestre 1 Quarter 1-Trimestre 1 Quarter 2-Trimestre 2 Quarter 2-Trimestre 2 Quarter 2-Trimestre 2 Quarter 3-Trimestre 3 Quarter 4-Trimestre 4 Quarter 4-Trimestre 4

Spatial coverage of the dolphin-associated fishery by quarter



Due to a large seasonal variation in fishing locations, each quarter's data were fitted to the spatiotemporal model separately.

- 1. Catchability could be quarter-specific (environmental conditions vary from quarter to quarter)
- 2. Autocorrelations in the spatial and spatiotemporal residuals could be different (driven by different processes: static vs. dynamic)
- 3. Imputed catch rates for the unsampled region could be biased in different ways by quarter



# Discussion: standardized index vs. nominal index

The standardized index represents an **improvement** because the approach:

1. Estimates the coefficient of variation (CV) of the index



# Discussion: standardized index vs. nominal index

The standardized index represents an **improvement** because the approach:

- 2. Includes vessel effects on catchability
- 3. Accounts for preferential sampling



Each vessel was better at finding fishing hotspots over time -> overestimating the trend in biomass

New vessels were more efficient than old vessels -> overestimating the trend in biomass



# Discussion: next phase of this project







