

# Persistent hotspots of billfish (by)catch pervade the high seas of the tropical Pacific Ocean



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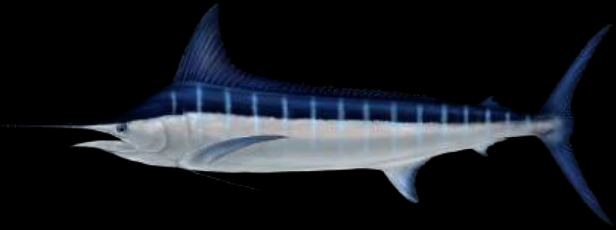
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# Oceanic Pacific Billfishes



## Striped Marlin

- Overfished
  - Western/Central North
  - Southwest
- Healthy
  - Eastern Tropical



## Blue Marlin

- Near full exploitation
  - Panmictic

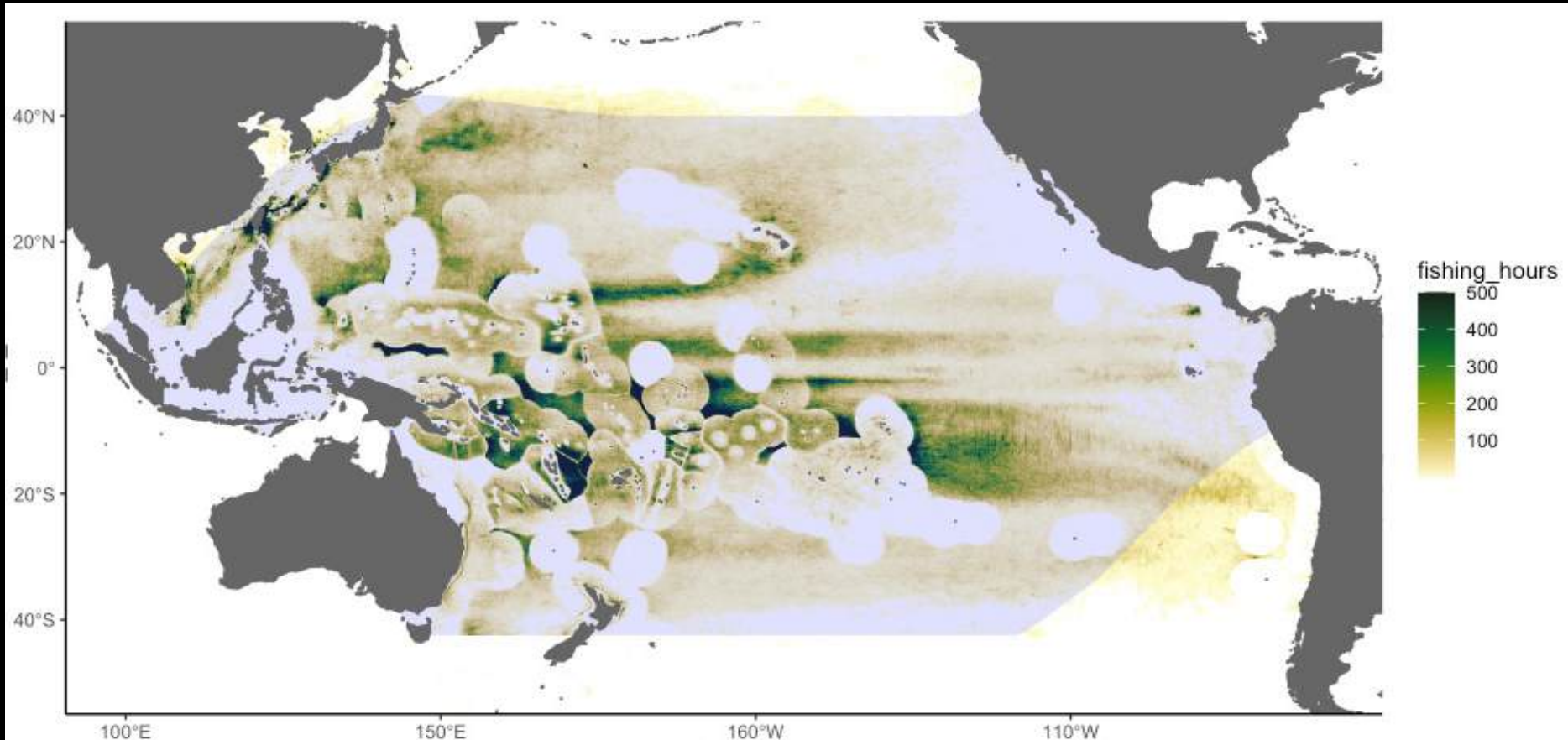


## Shortbill Spearfish

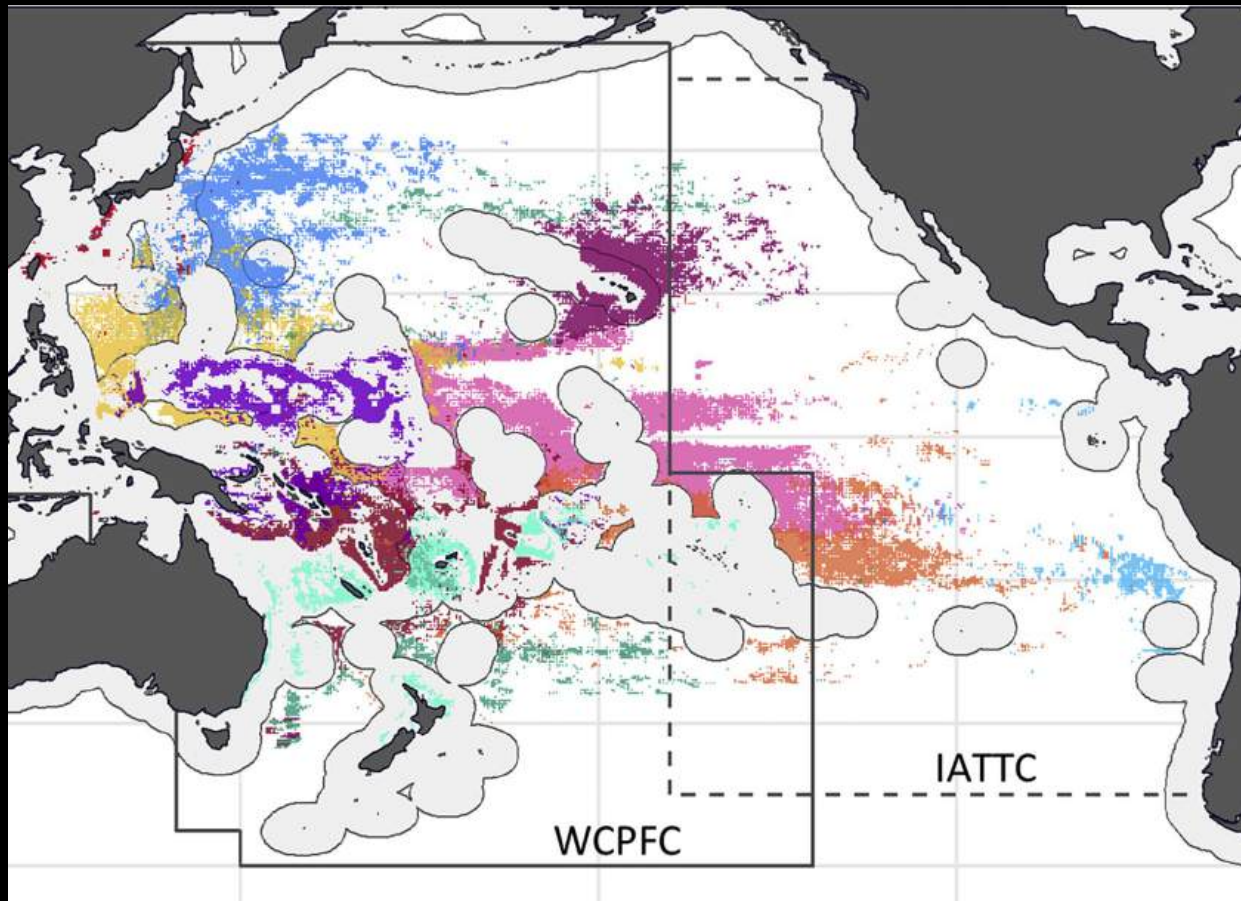
- No stock assessment
  - Likely North-South stock delineation

Longlines responsible for  
> 90% of landings

# Pacific Longline Effort



# Functional Longline Fleets of the Pacific



- West Tropical Foreign
- Tropical Distant Water
- USA Offshore
- Southwest Foreign
- Southwest Domestic
- Southeast Distant Water
- Taiwan Offshore
- Japan Offshore
- Northwest Domestic
- East Distant Water
- Dual-hemisphere Distant Water

Fleets are:



Multi-national



Multi-jurisdictional

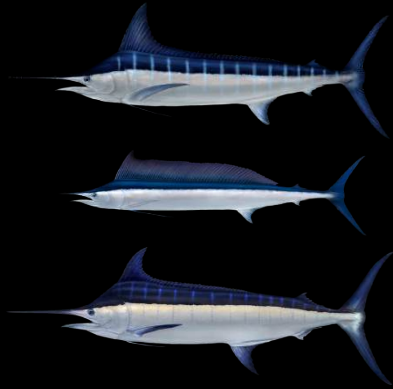


Multi-specific

# Identifying Billfish (By)catch Hotspots

## Phase 1 – Billfishes

- Build predictive species distribution models (SDMs)
- Predict daily occurrence



## Phase 2 – Longline Fleets

- Build predictive vessel distribution models (VDMs)
- Predict daily occurrence



Overlap



Dynamic (By)catch Susceptibility

*How changing ocean conditions create hotspots of risk*

# Data – Billfish SDMs

## Billfish Data Sources

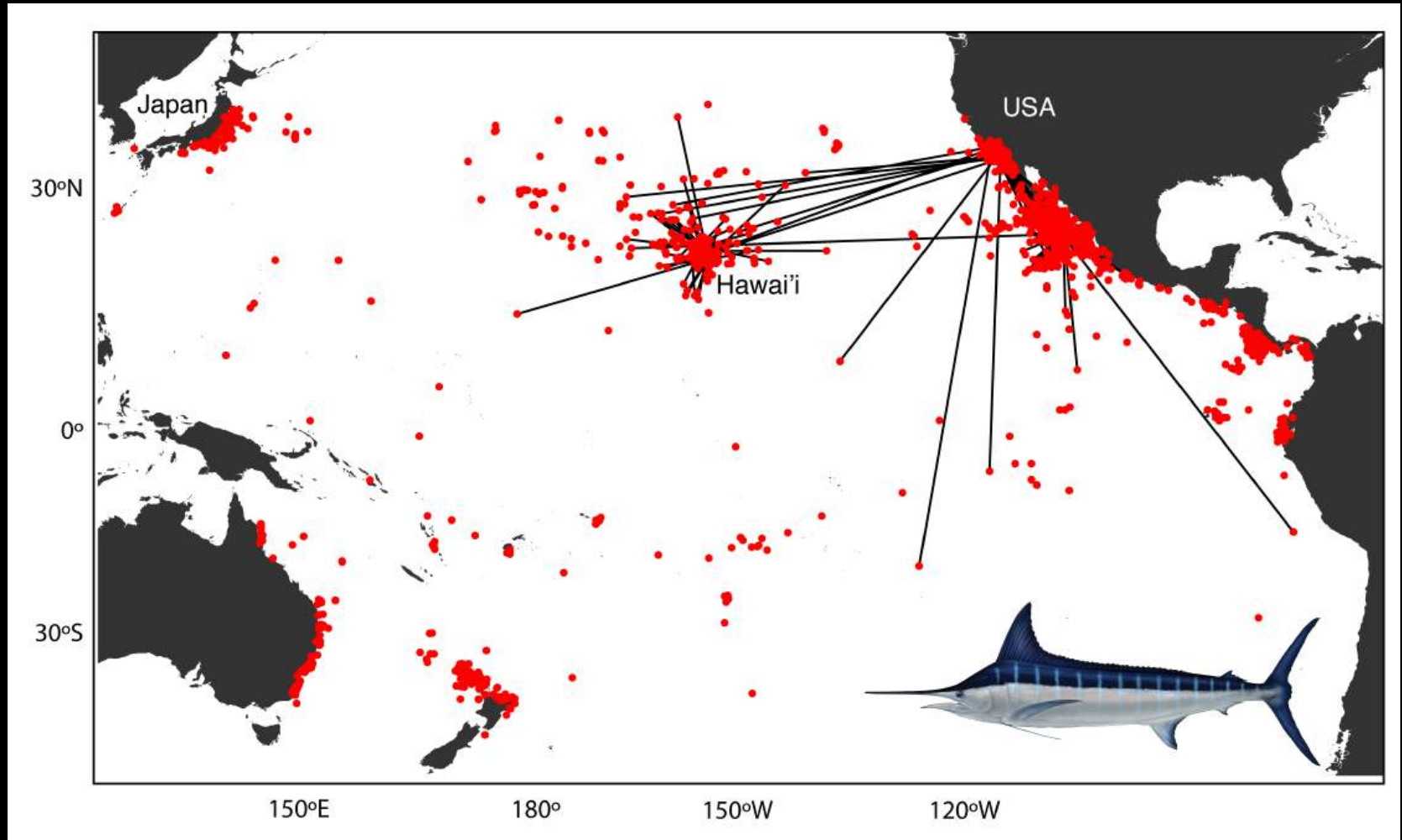
- Conventional Tagging
- Hawai`i Longline Observer
- Taiwan Longline Logbook
- IATTC Longline & Purse-Seine Observer

## Modeling Approach

- Boosted Regression Tree (BRT)
- 12 variables: daily, 0.25° resolution
- 27 years of training data (1997-2023)
- N = 121,000 – 534,000 records per species

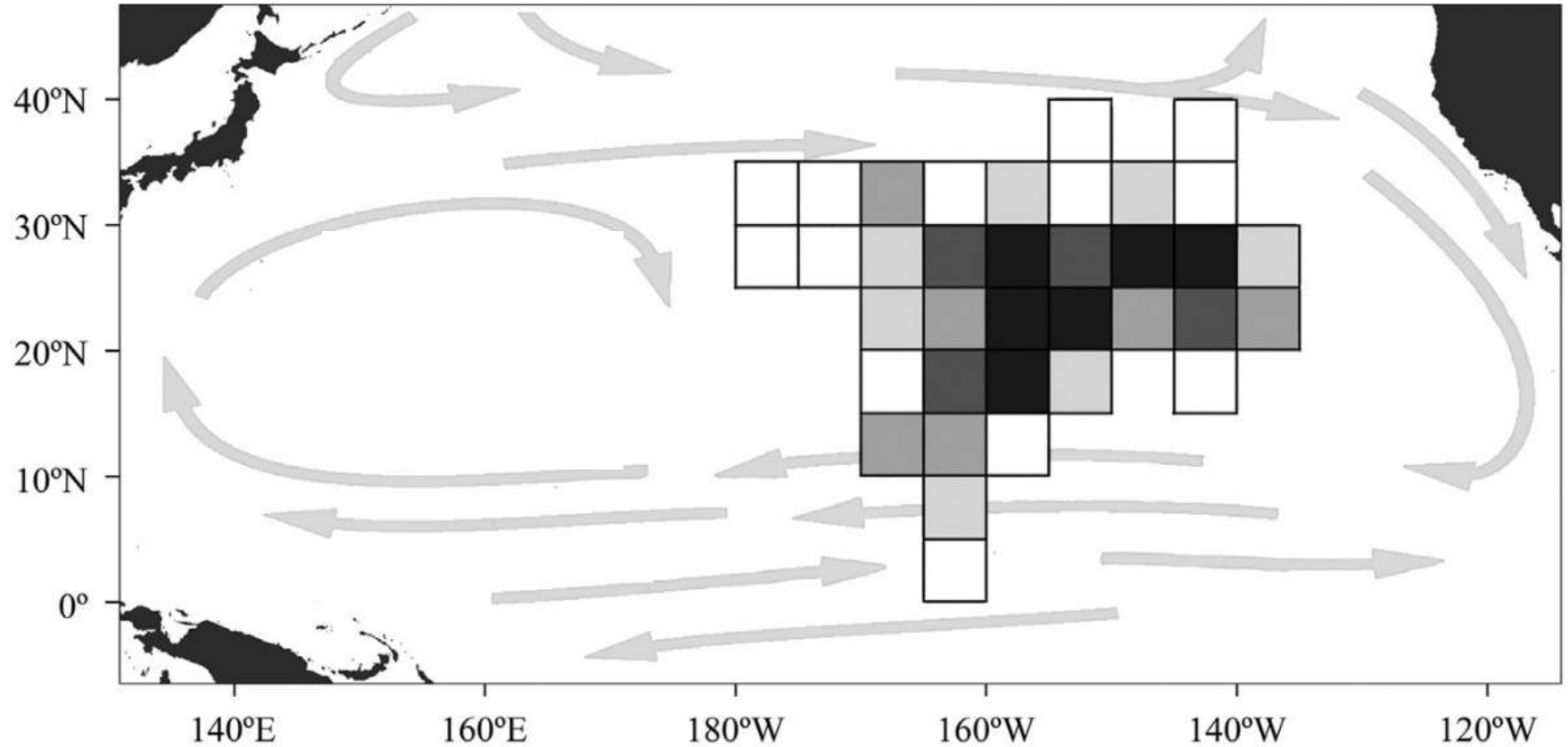


# Conventional Tagging



~5,000 – 70,000 records per species

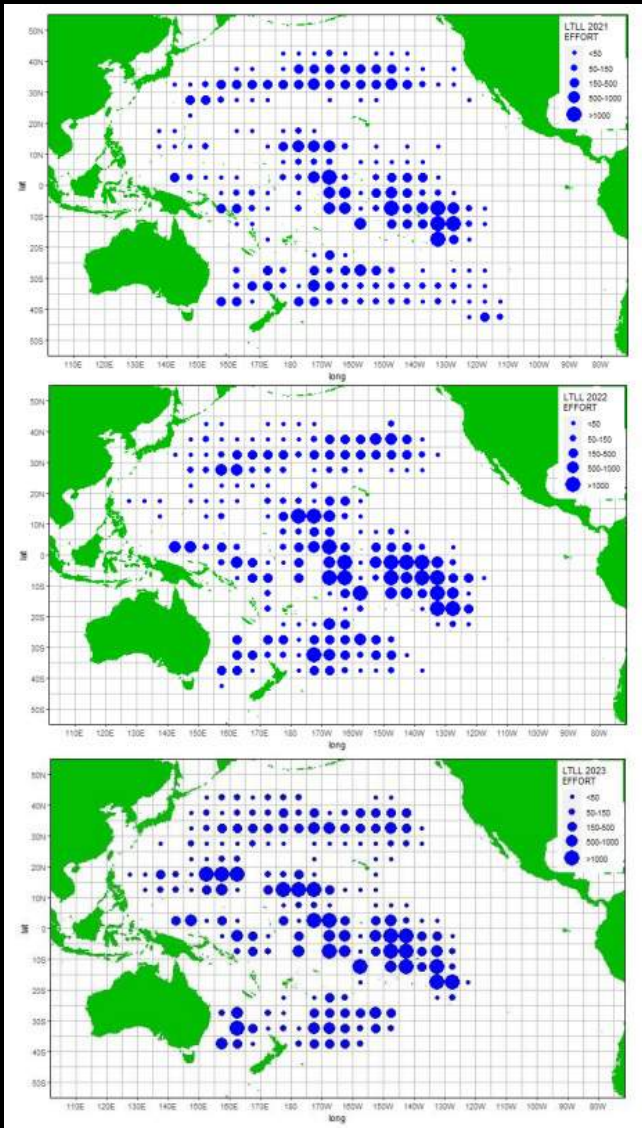
# Hawai`i Longline (Observer)



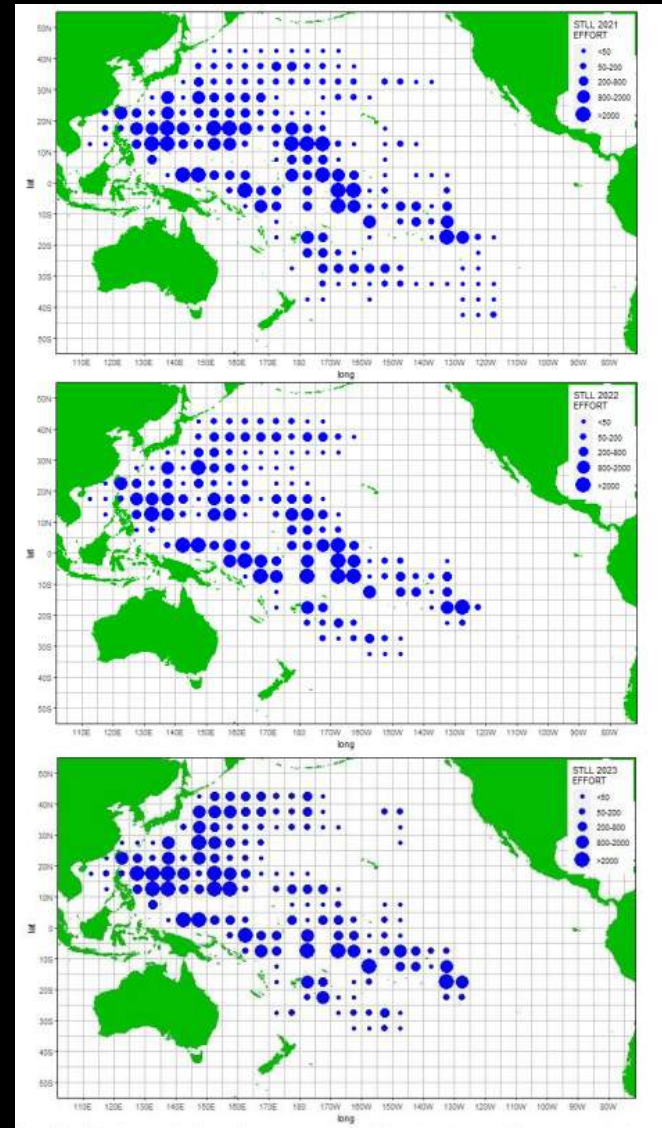
~19,000 – 42,000 records per species

# Taiwan (Chinese Taipei) Longline (Logbook)

Large-Scale

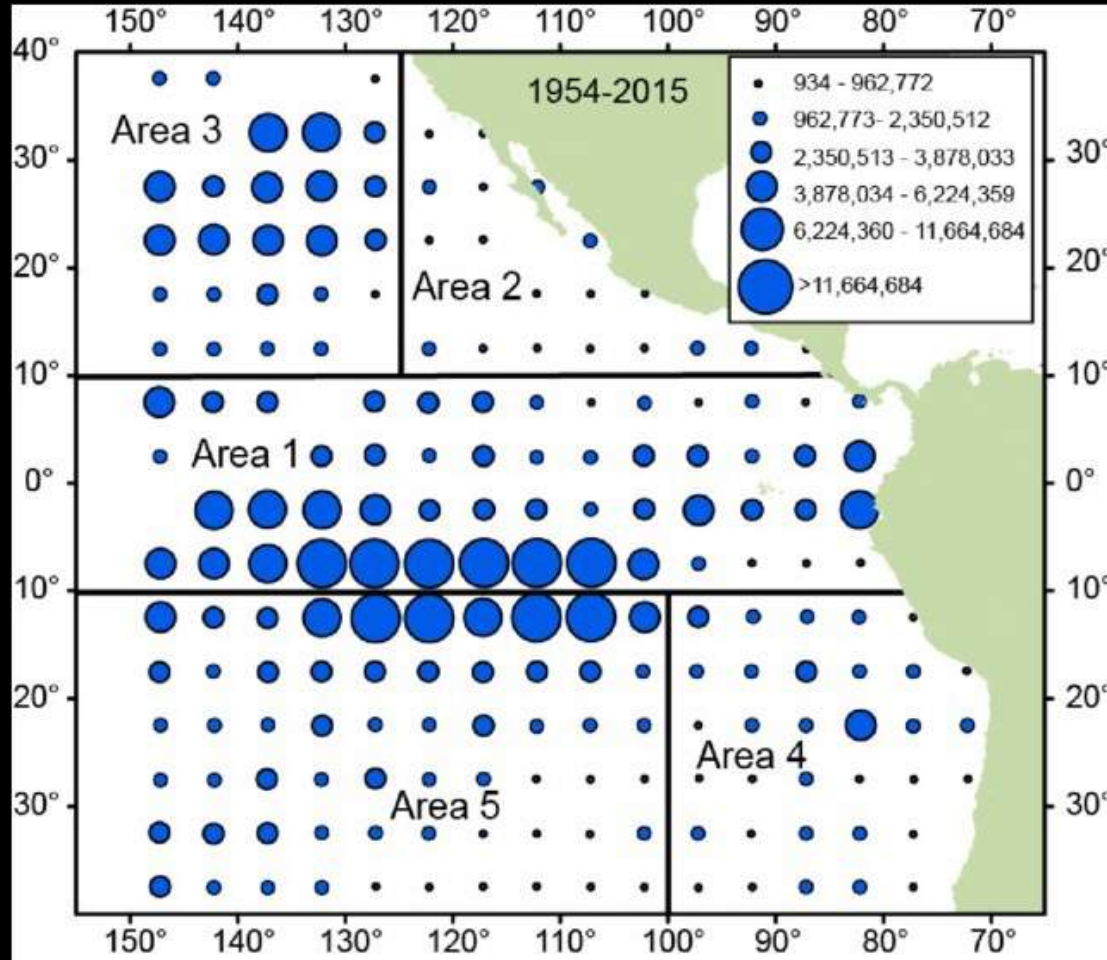


Small-Scale



~61,000 – 486,000 records per species

# IATTC Longline & Purse-Seine (Observer)



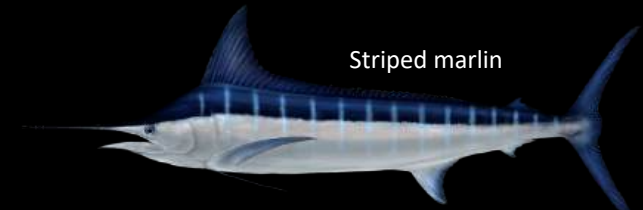
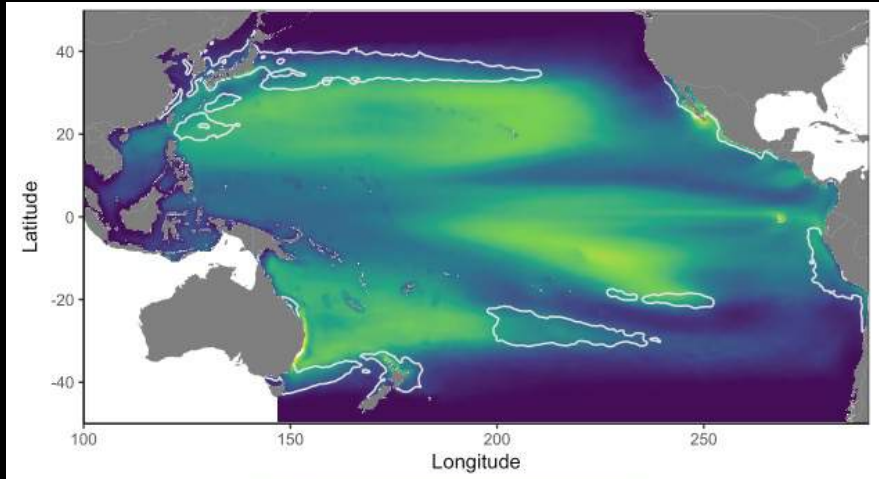
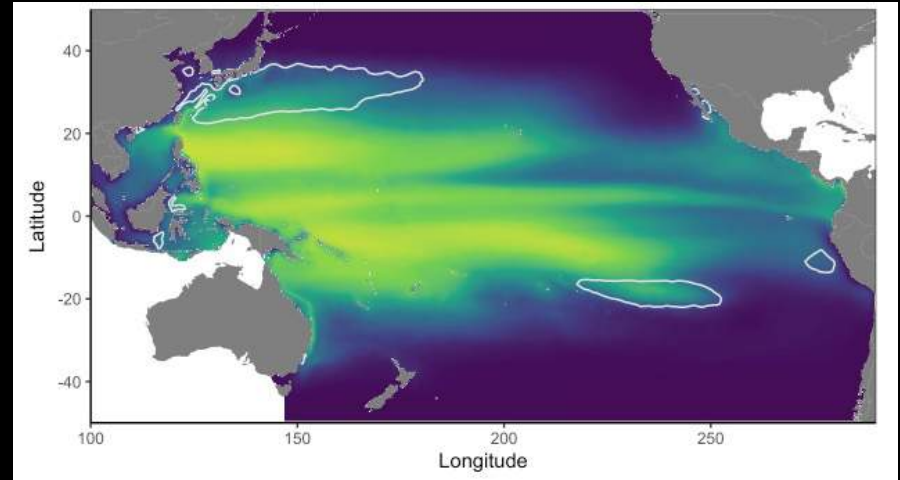
~13,000 – 50,000 records per species

# Billfish SDMs



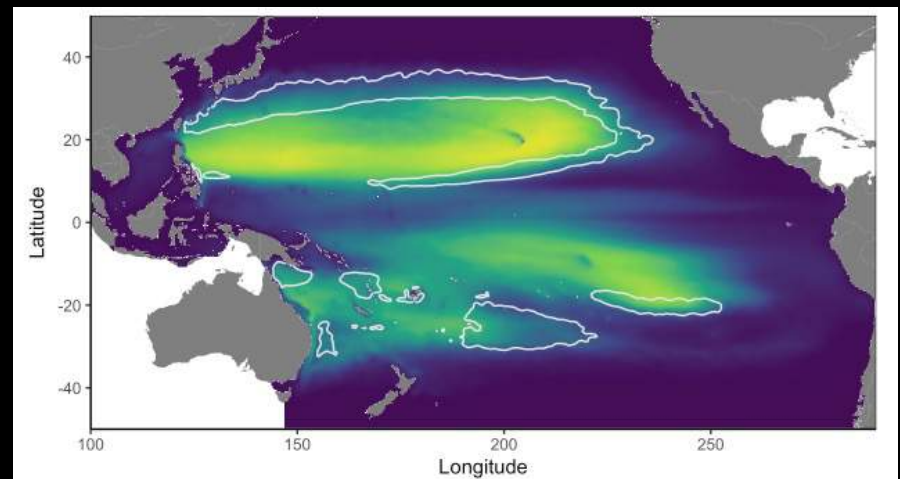
Blue marlin

$R^2 = 0.33$ , AUC = 0.84, TSS = 0.55  
Vars: SST (74%), SSH (10%), SSS (8%)



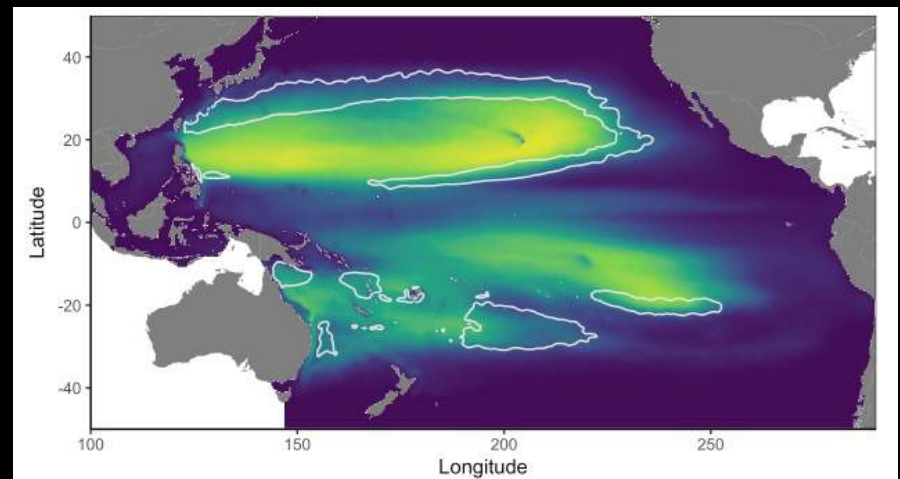
Striped marlin

$R^2 = 0.30$ , AUC = 0.84, TSS = 0.50  
Vars: SST (46%), SSH (15%), CHL (13%)



Shortbill spearfish

$R^2 = 0.44$ , AUC = 0.90, TSS = 0.66  
Vars: SSH (53%), SSS (16%), CHL (14%)



# Data – Longline VDMs

## Vessel Data Sources

- Global Fishing Watch
- Functional fleet assignment
  - Frawley et al. 2022 *One Earth*
- 5 pre-existing VDMs
  - Frawley et al. 2024 *Fish and Fisheries*

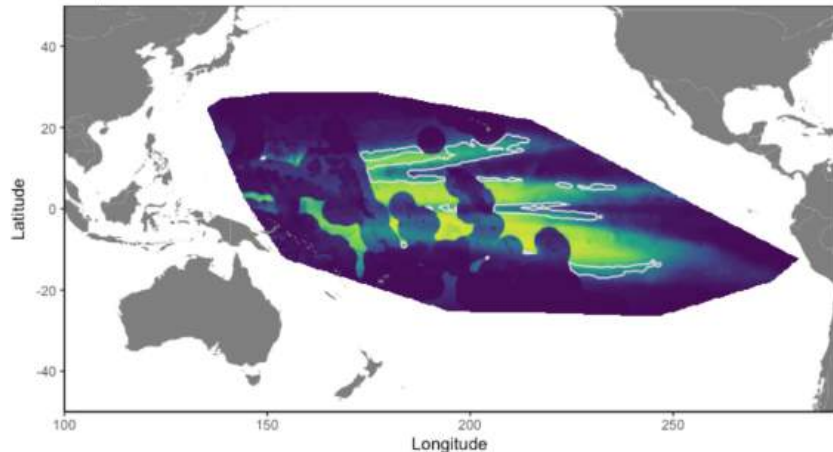
## Modeling Approach

- Boosted Regression Tree (BRT)
- 12 variables: daily, 0.25° resolution
- 3 years of training data (2017-2019)
- N = 14,000 – 184,000 records per fleet



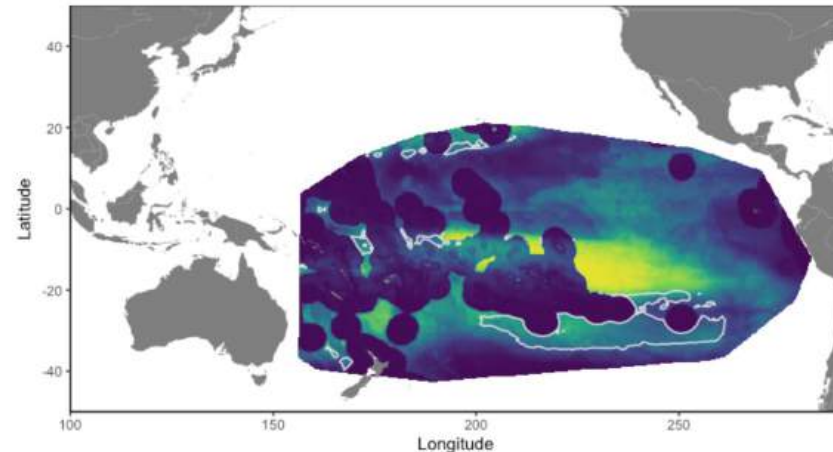
# Distant Water Fleets

Tropical Distant Water



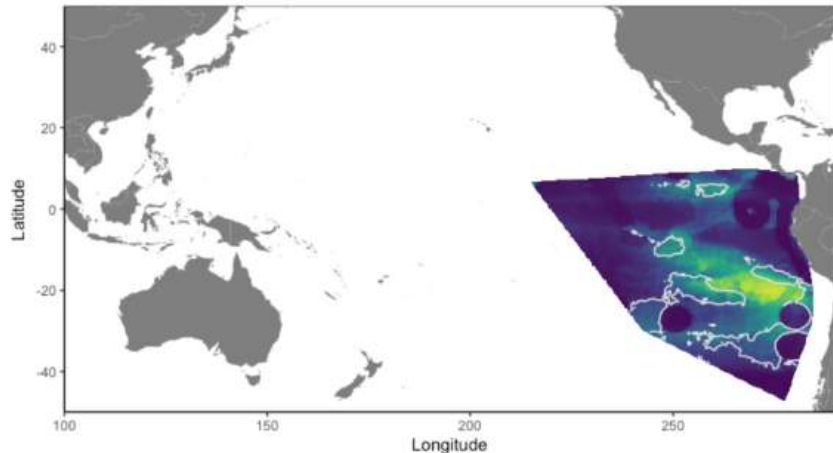
$R^2 = 0.53$ ,  $AUC = 0.93$ ,  $TSS = 0.74$   
Vars: Dist (31%), DO (23%), NPPV (13%)

Southeast Distant Water



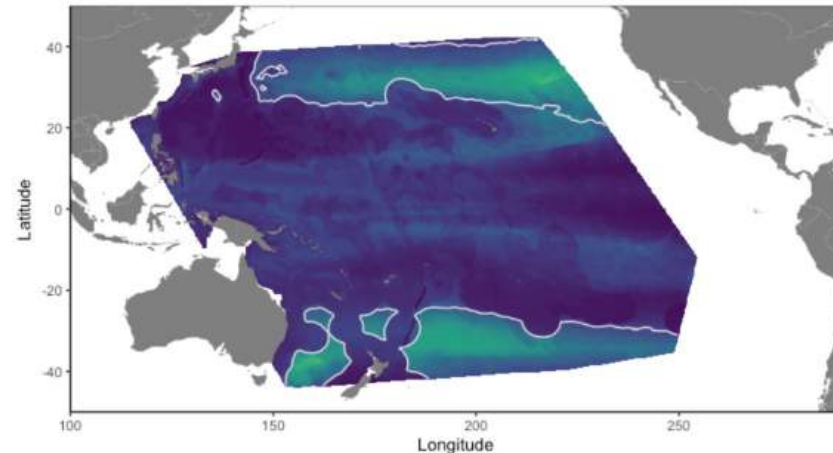
$R^2 = 0.50$ ,  $AUC = 0.92$ ,  $TSS = 0.69$   
Vars: Dist (30%), DO (22%), ADT (14%)

East Distant Water



$R^2 = 0.53$ ,  $AUC = 0.93$ ,  $TSS = 0.72$   
Vars: NPPV (20%), Bathy (16%), SST (15%)

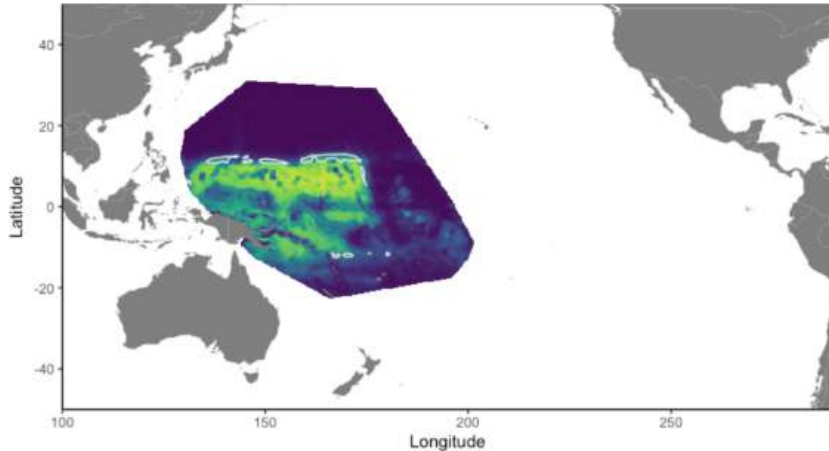
Dual-Hemisphere Distant Water



$R^2 = 0.53$ ,  $AUC = 0.94$ ,  $TSS = 0.75$   
Vars: SST (58%), MLD (20%), Dist (7%)

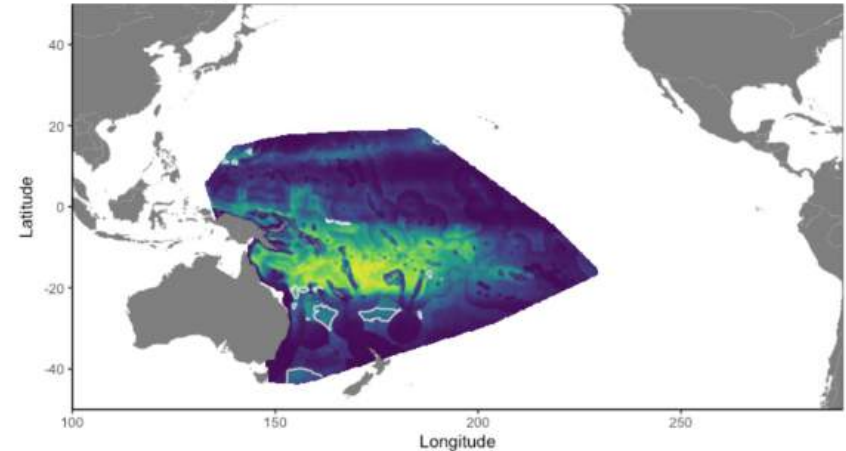
# Foreign & Domestic Fleets

West Tropical Foreign



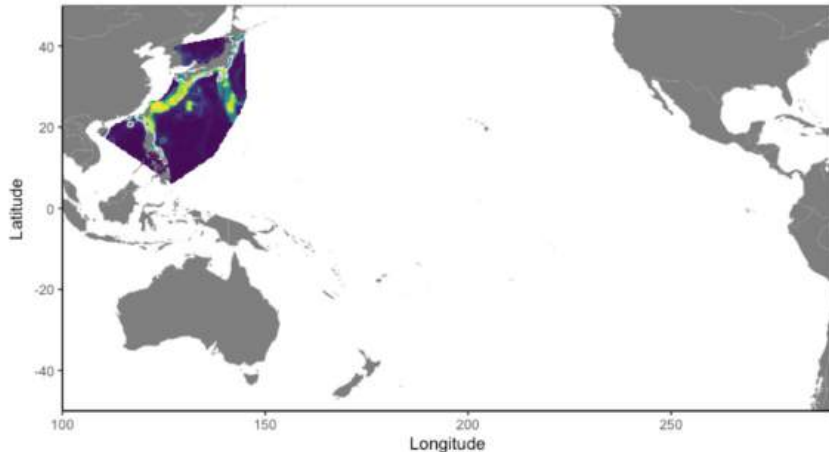
$R^2 = 0.47$ , AUC = 0.91, TSS = 0.67  
Vars: DO (53%), Dist (21%), Bathy (11%)

Southwest Foreign



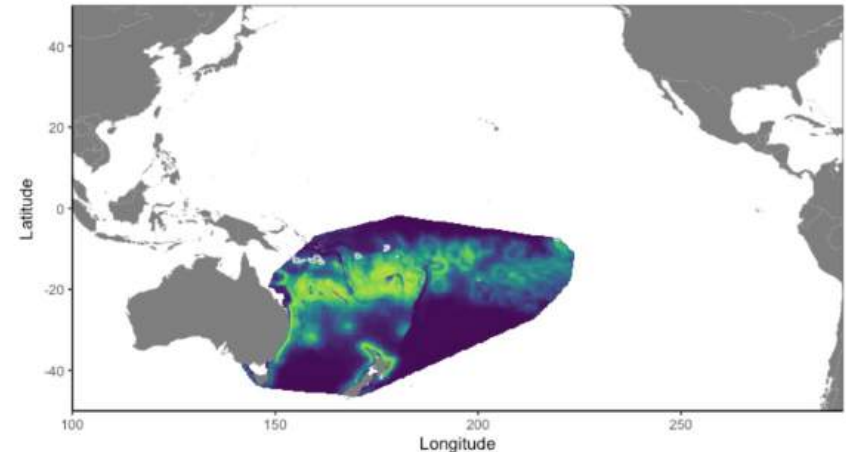
$R^2 = 0.48$ , AUC = 0.92, TSS = 0.69  
Vars: DO (28%), ADT (27%), Bathy (18%)

Northwest Domestic



$R^2 = 0.35$ , AUC = 0.87, TSS = 0.58  
Vars: Dist (55%), DO (20%), Bathy (13%)

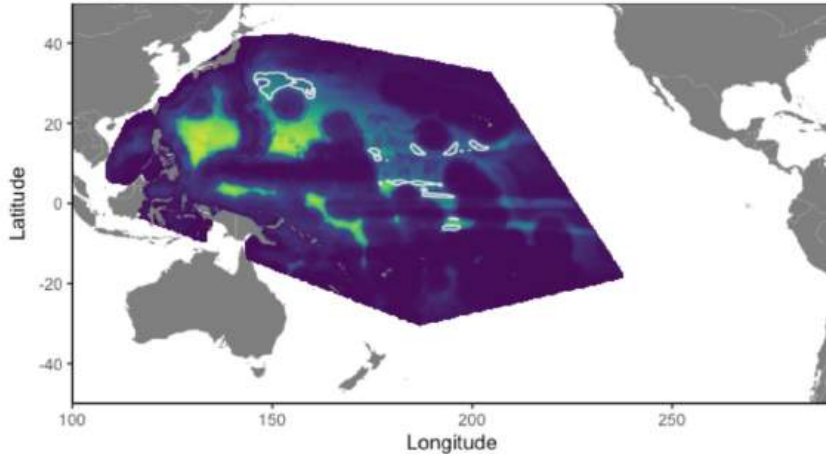
Southwest Domestic



$R^2 = 0.43$ , AUC = 0.89, TSS = 0.64  
Vars: DO (28%), Dist (24%), Bathy (22%)

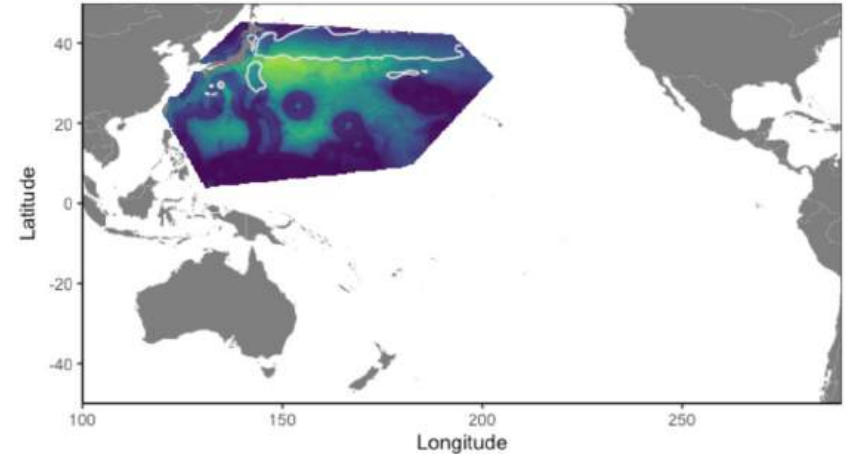
# Primarily Single-Nation Fleets

Taiwan (Chinese Taipei) Offshore



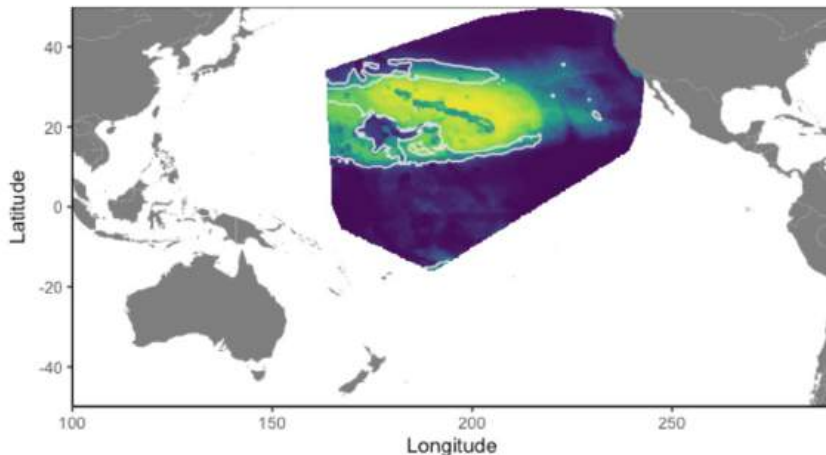
$R^2 = 0.51$ ,  $AUC = 0.93$ ,  $TSS = 0.74$   
Vars: Dist (31%), DO (23%), NPPV (13%)

Japan Offshore



$R^2 = 0.35$ ,  $AUC = 0.87$ ,  $TSS = 0.58$   
Vars: Dist (19%), SST (18%), DO (16%)

USA Offshore



$R^2 = 0.61$ ,  $AUC = 0.95$ ,  $TSS = 0.79$   
Vars: NPPV (36%), Dist (29%), ADT (14%)

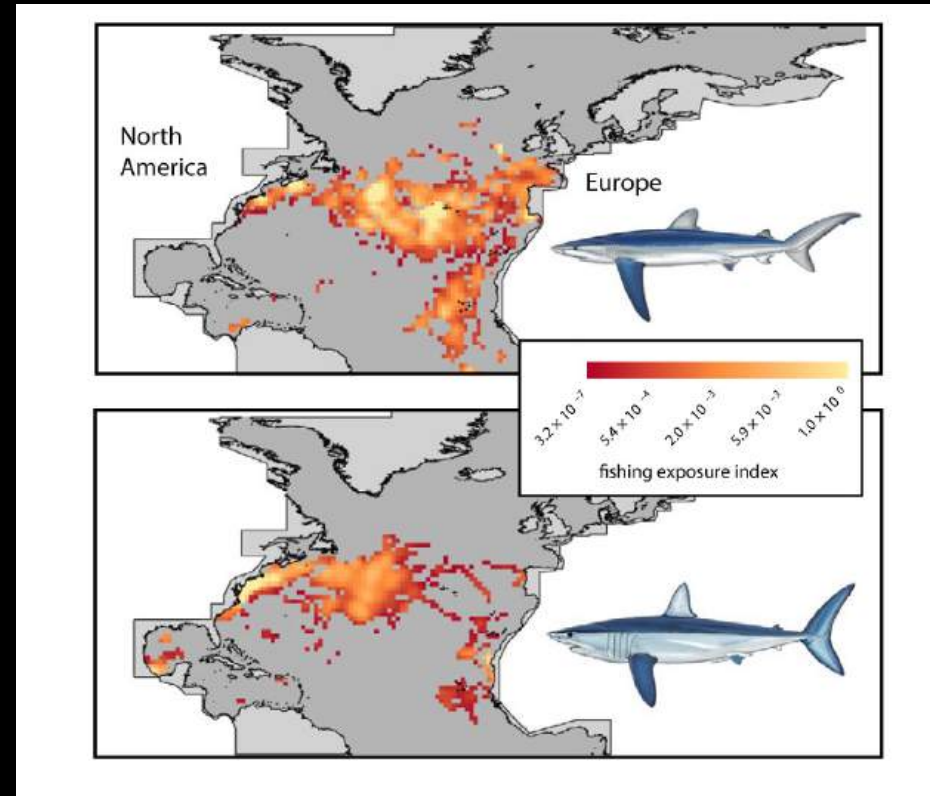
# Quantifying Risk & Persistence

## Risk

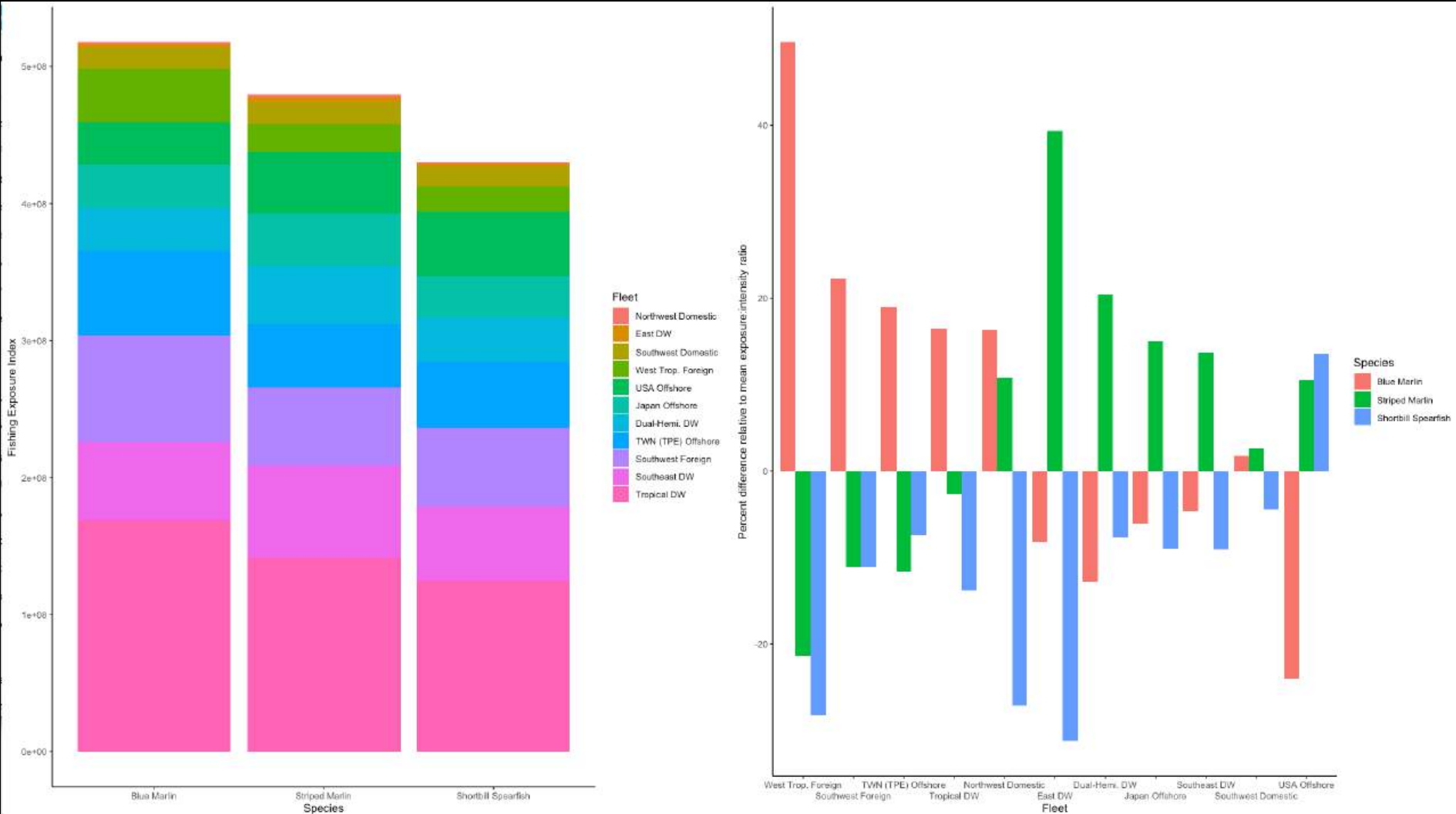
- Fleet-specific scaling
  - Overall effort (vessel days)
  - Intra-annual effort variation (monthly proportion of vessel days)
- Summed across fleets
- Hotspot: > 90<sup>th</sup> risk quantile

## Persistence

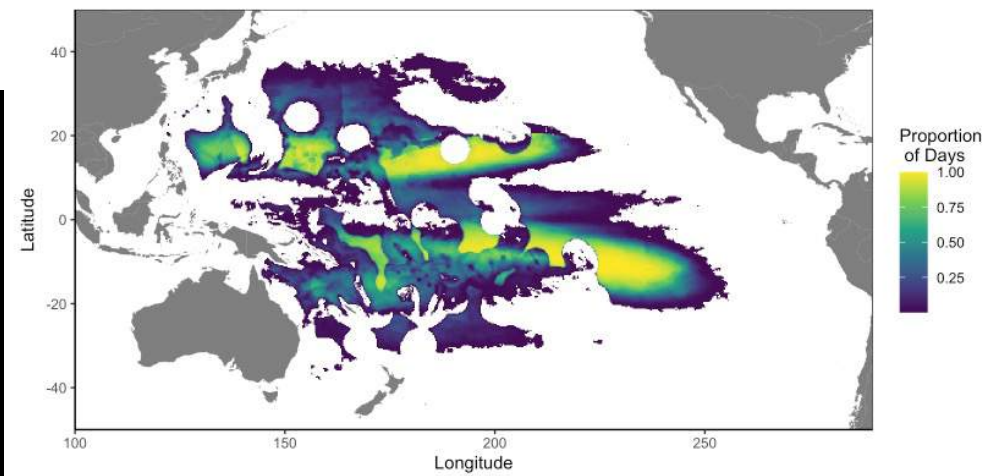
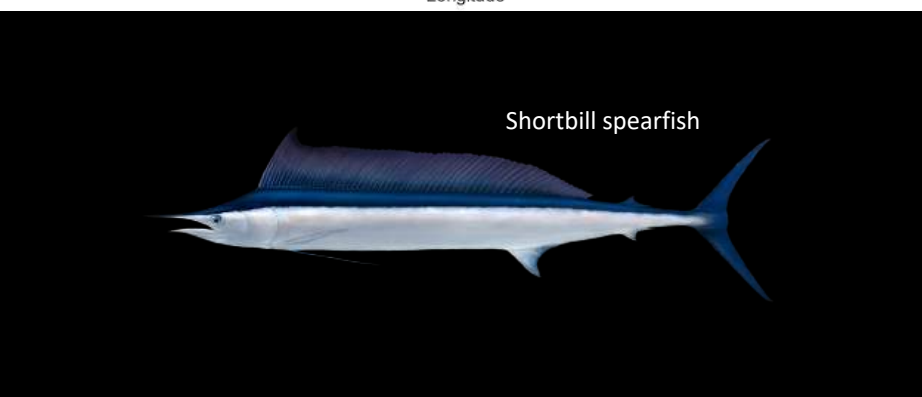
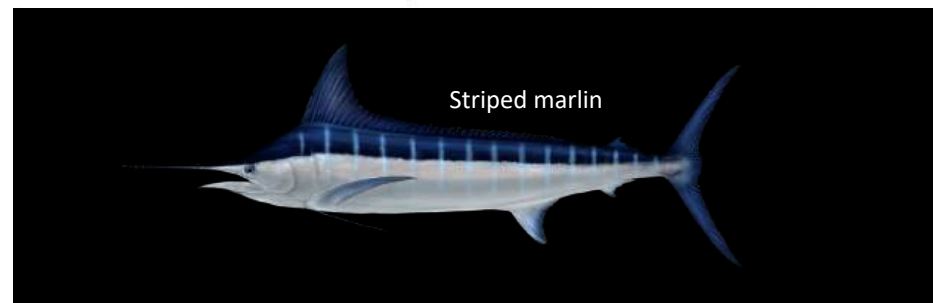
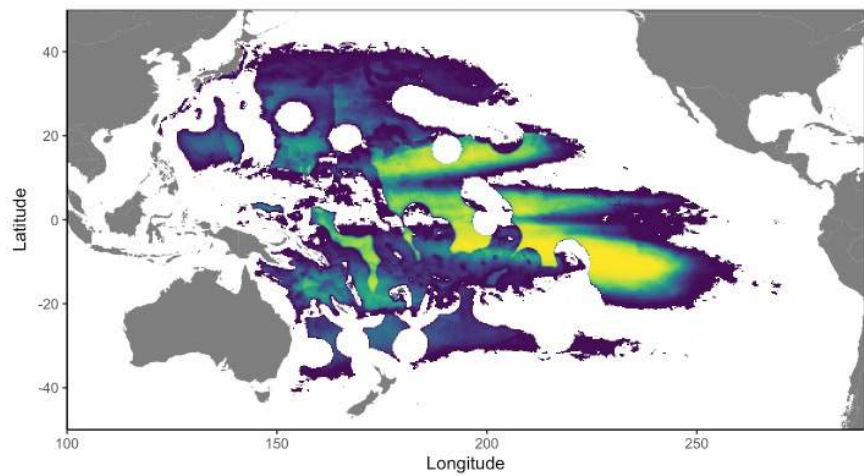
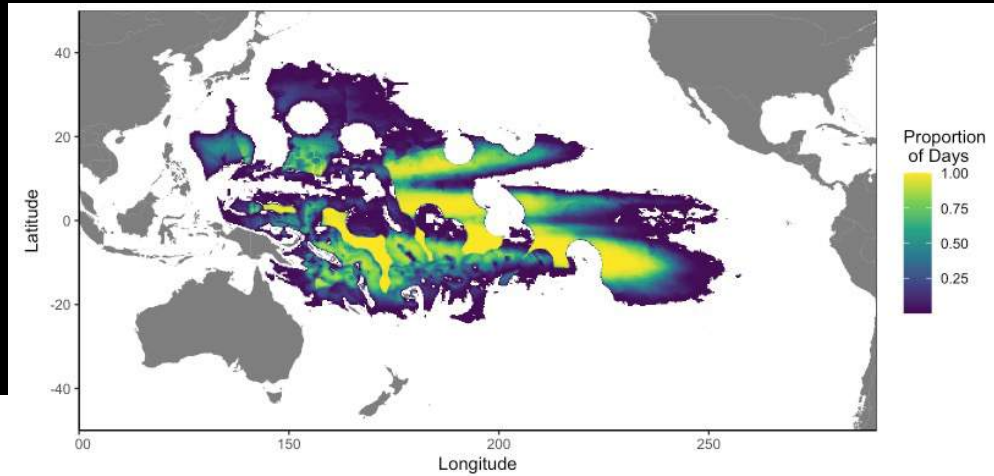
- Persistent:  $\geq 50\%$  of days
- Ephemeral:  $> 0$  but  $< 50\%$
- 10-year, seasonal, interannual, ENSO, PDO



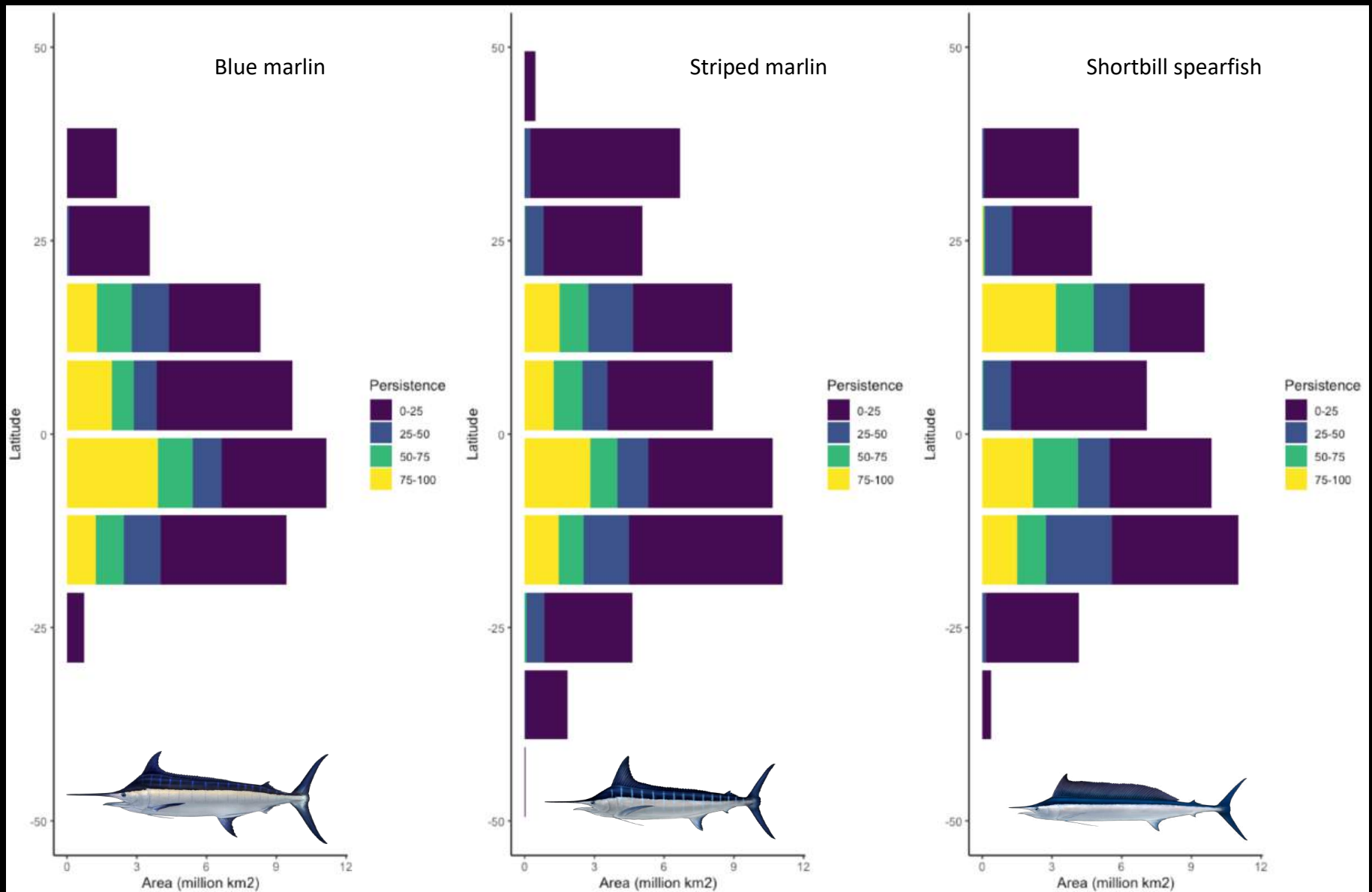
# Cumulative and Fleet-Specific Exposure



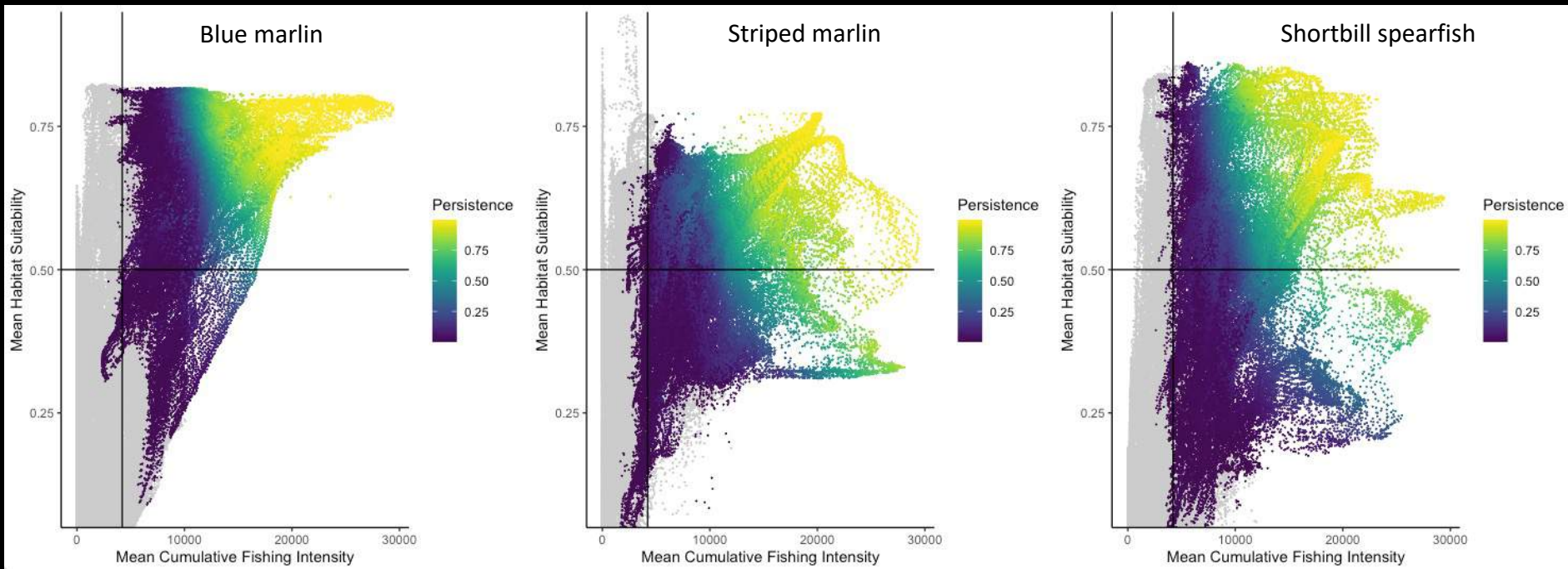
# Billfish (By)catch Hotspots



# Billfish (By)catch Hotspots



# What Makes a Hotspot



# WCPFC

# IATTC

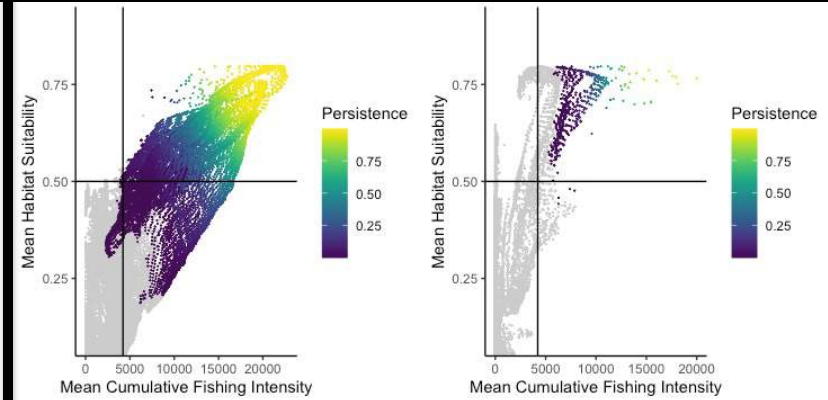
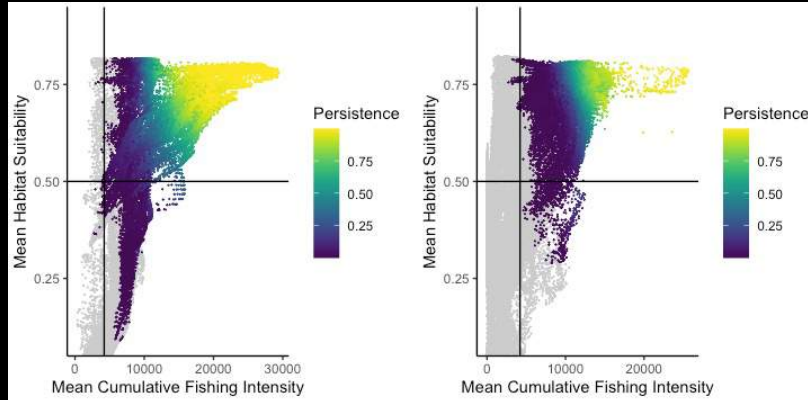
High Seas

EEZs

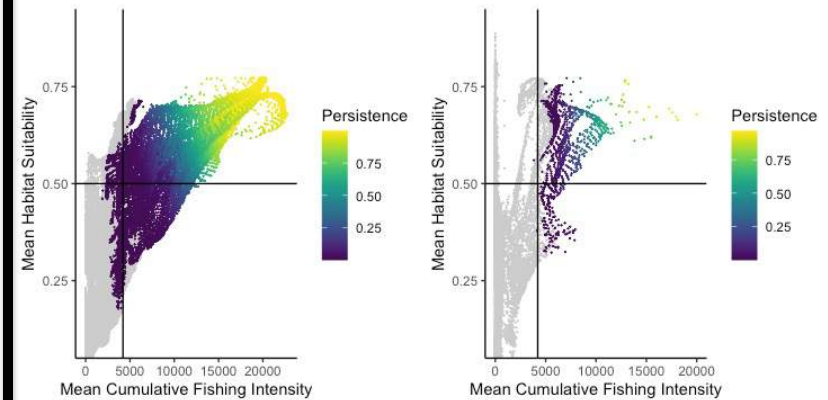
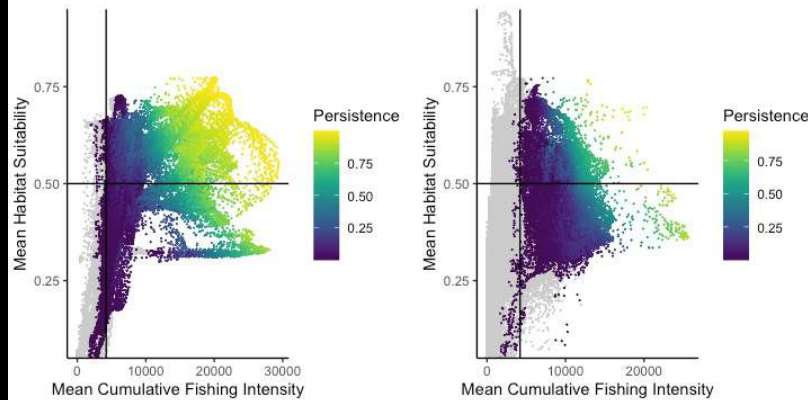
High Seas

EEZs

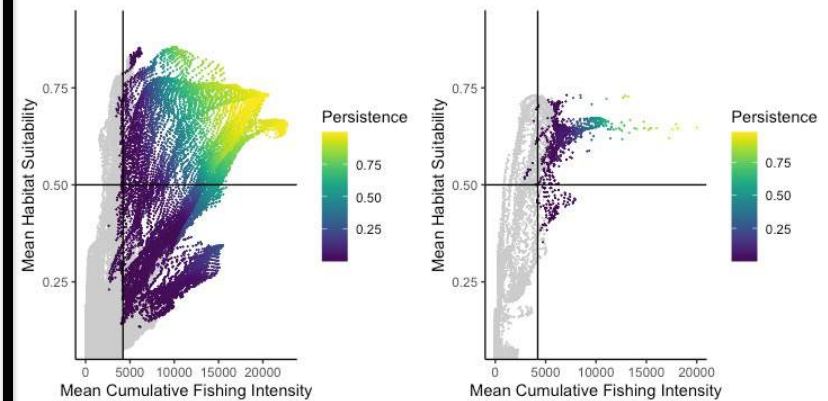
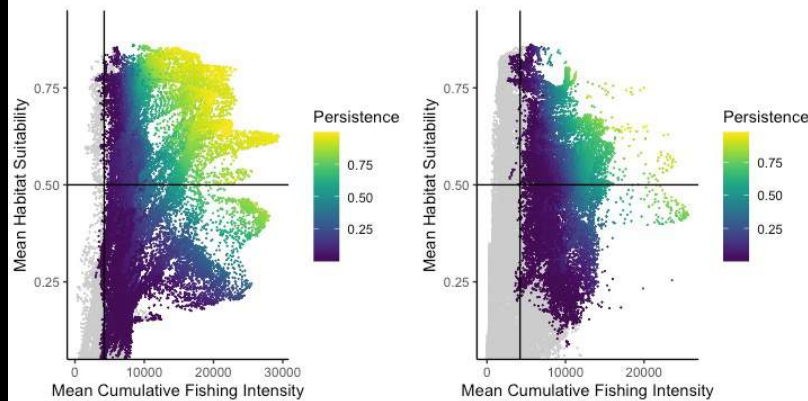
Blue marlin



Striped marlin



Shortbill spearfish



# Feedback For Actionable Insight

- What analyses would be most informative to management Pacific-wide and within your jurisdiction?
- What context might we want to emphasize to better situate these analyses in the broader issues of (by)catch and management across EEZs and ABNJ?

