Spatial Variability in Bigeye Vertical Behaviour: Environmental Influences and Impacts on Fisheries

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Objectives:

Are there spatial variations in fish depth distributions?

Can we link that variations with the environment?

What are the factors affecting the probability of association to FADs?

Can we use the information provided from electronic tags to improve stock assessment and management?

Spatial Observations of Bigeye CPUE

Purse Seine



Spatial Observations of Bigeye CPUE Longline



0.2 0.4 0.6 0.8 1 1.2 1.4 max~75

Data sources

- Archival tag data: used for depth, thermocline depth, attenuation coefficients...
- Moonphase US naval observatory)
- Forage: Tuna forage model-Seapodym
- Oxygen: Climatological data from the WOA
- Chla and PP: Oregon State University.
- Cloud cover to correct moon illumination and explain night depth (OMI-Aura-L3 product)
- IRD model data
- DSL depth (based in attenuation coefficient and solar irradiation)
- Solar elevation (proxy for season)
- sea surface height, surface currents, salinity and temperature at 45 m from GODAS
- Land, seamounts, bathymetry
- Thermal gradient from REMSS.
- Fish length

300	Corr	Corr:	Corr.	Corr:	Corr.	Corr:	Corr:	Corr.	Corr:
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.	3 2 meso 1 0 1 2 3	Corr: 0.2	-0.351	-0.444	-0.45	Corr: -0.195	Corr: -0.355	Corr: 0.401	Corr. 0.38
46	6	mmeso 0.5	Corr: -0.519	Corr: 	Corr: -0.47	Corr: 0.428	Corr: -0.293	Corr: 0.586	Corr: 0.36
Ø ·	.		300 -200 ₀ x3 -100-200300	Corr: 0.8	Corr: 0.762	Corr: 0.662	Corr: 0.444		Corr: -0.554
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Archival data







Phase of day











datab\$datetime

Behavioral classification (Schaefer & Fuller, 2010)





Behavioral type

Lon

Daytime. Nominal depths



Nighttime. Nominal depths

Modelling

- Split data into daytime and nighttime
- Classify according to three behavioral types: Associated, Characteristic, Other
- Model with GAMMs using tag as random effect.
- Data sources:
 - Fish length: length at release and time at liberty
 - Thermocline depth (D20): Tag data
 - Temperature stratification (D18-D20): Tag data
 - Moon phase: US Naval observatory
 - Rest: interpolation based on geolocation estimates of satellite/model products

What influences the behavioral type?

Daytime characteristic depth vs oxygen

Residual distribution daytime characteristic model

E.g, nighttime associated, 70 cm fish

E.g, daytime characteristic, 120 cm fish

Nominal CPUE vs predicted depths

Nominal CPUE vs predicted depths

LL Standardized CPUE

Some potential discussions:

Are the spatial trends in the species composition in PS and LL CPUE driven by catchability?.

Can the patterns observed in 2015 in LL and PS CPUE be explained with these results?

Can it be used, and how, in future stock assessments?

Can this be a tool for the management of the species?

THANK YOU