



Comisión Interamericana del Atún Tropical
Inter-American Tropical Tuna Commission



Standardization including size composition data in VAST

Haikun Xu, Cleridy Lennert-Cody, Mark Maunder, Carolina Minte-Vera, and Keisuke Satoh

Model Structure (multiple categories (sizes))

VAST separately models encounter probability (p) and positive catch rate (λ) for each catch rate observation i :

$$\text{logit}(p_i) = \beta_1(c_i, t_i) + \sum_{f=1}^{n_{\omega_1}} L_{\omega 1}(c_i, f) \omega_1(s_i, f) + \sum_{f=1}^{n_{\varepsilon_1}} L_{\varepsilon 1}(c_i, f) \varepsilon_1(s_i, t_i, f) + \dots$$

$$\log(\lambda_i) = \beta_2(c_i, t_i) + \sum_{f=1}^{n_{\omega_2}} L_{\omega 2}(c_i, f) \omega_2(s_i, f) + \sum_{f=1}^{n_{\varepsilon_2}} L_{\varepsilon 2}(c_i, f) \varepsilon_2(s_i, t_i, f) + \dots$$

$\beta(t_i)$: intercept for category c_i in year t_i

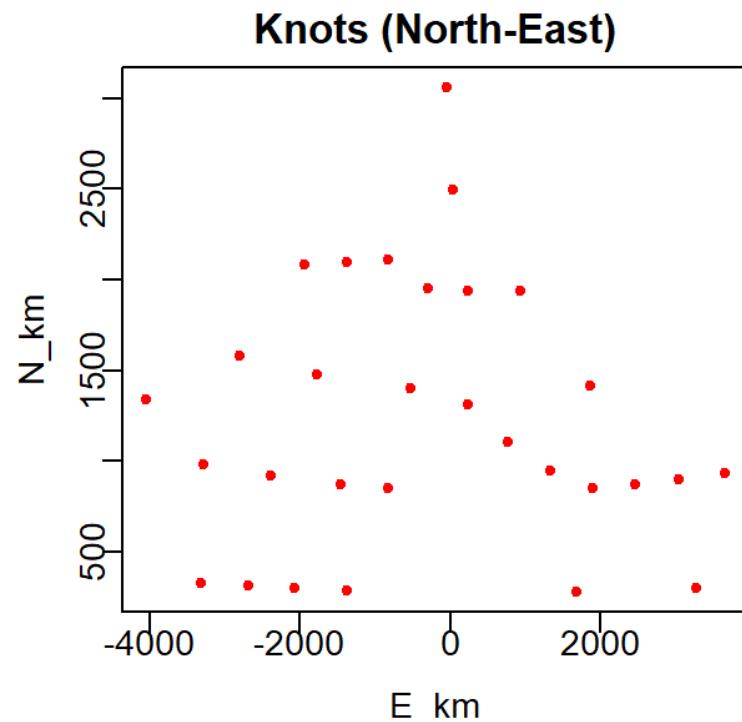
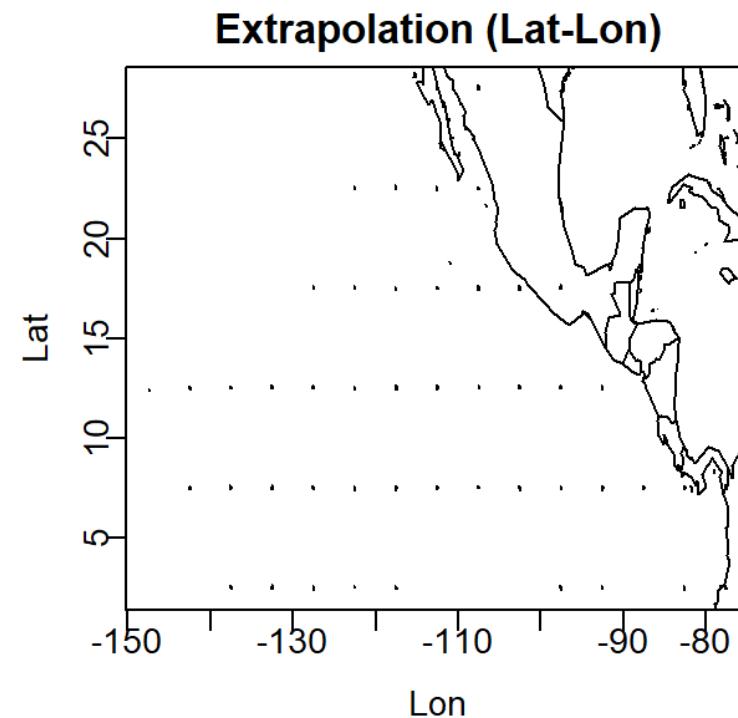
$\omega(s_i, f)$: spatial variation for factor f at location s_i ; L_ω : loading matrix

$\varepsilon(s_i, t_i, f)$: spatiotemporal variation for factor f at location s_i in year t_i ; L_ε : loading matrix

Users can specify the number of factors (principle components) for the spatial (n_{ω_1} and n_{ω_2}) and spatiotemporal components (n_{ε_1} and n_{ε_2})

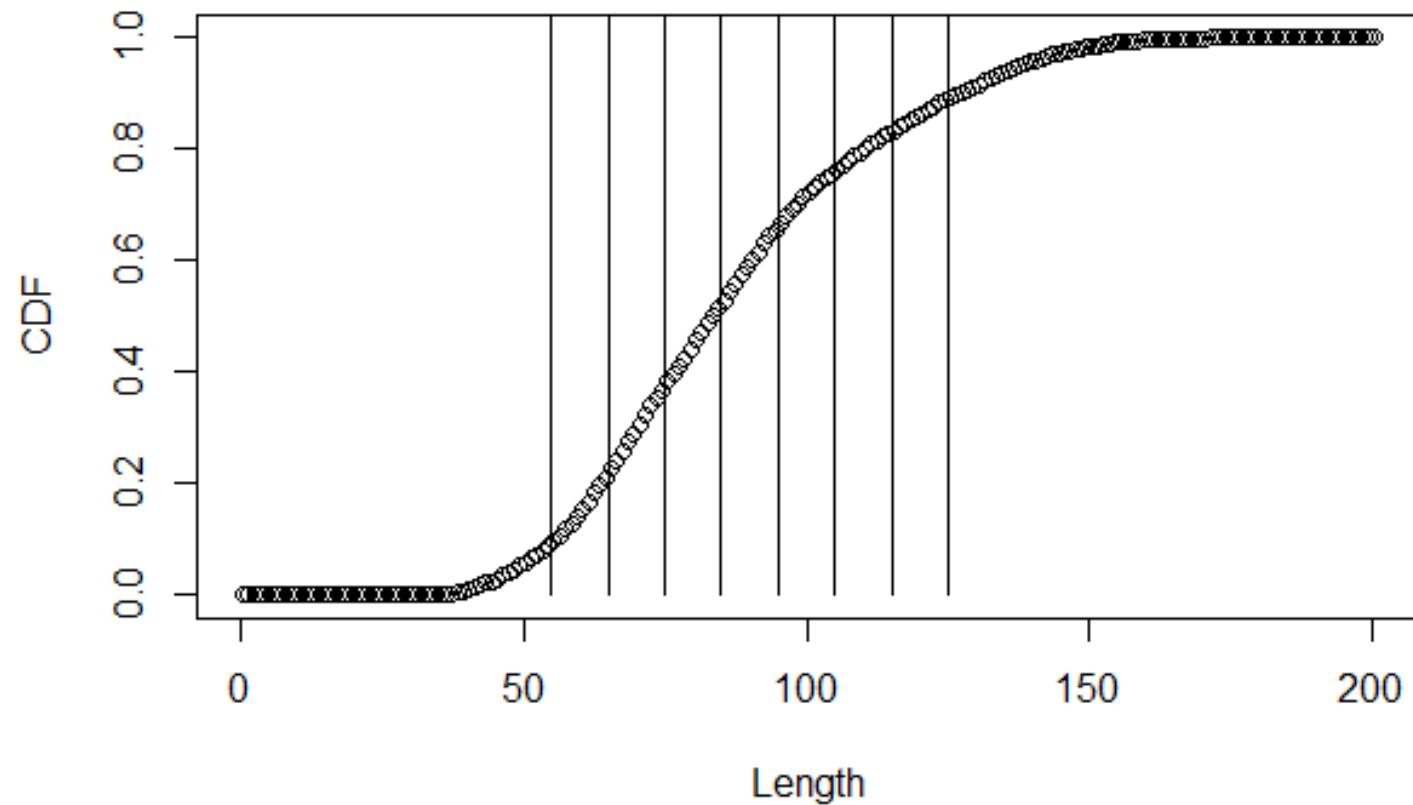
Case study: dolphin-associated EPO yellowfin fishery

- $5^{\circ} \times 5^{\circ}$ catch at length (1cm bin) and effort data from the dolphin-associated yellowfin fishery in the northern hemisphere in Quarter 2



Case study: dolphin-associated EPO yellowfin fishery

Yellowfin are grouped into 9 length bins



Model Structure (multiple categories (sizes))

VAST separately models encounter probability (p) and positive catch rate (λ) for each catch rate observation i :

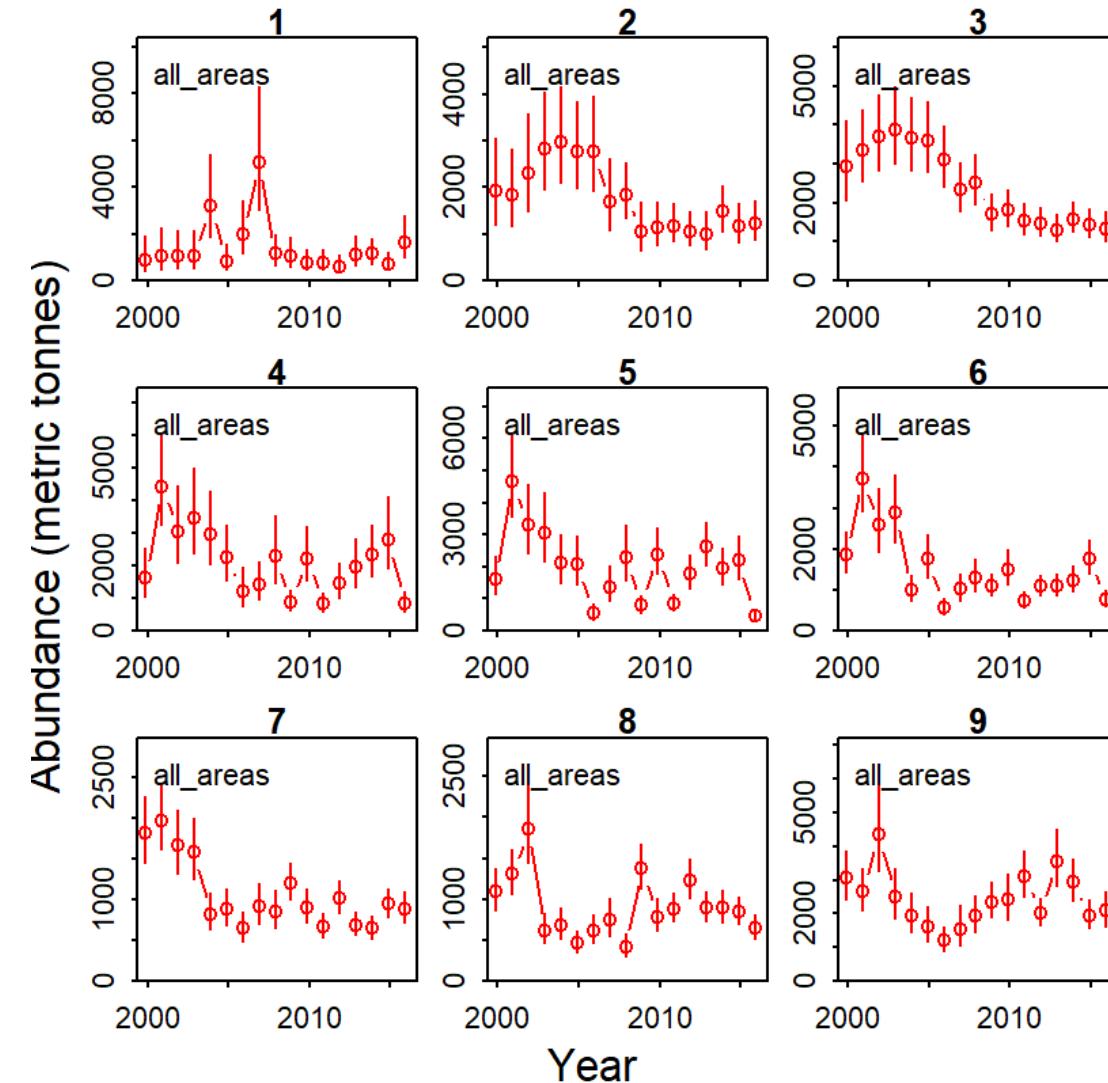
$$\text{logit}(p_i) = \beta_1(c_i, t_i) + \sum_{f=1}^{n_{\omega_1}} L_{\omega_1}(c_i, f) \omega_1(s_i, f) + \sum_{f=1}^{n_{\varepsilon_1}} L_{\varepsilon_1}(c_i, f) \varepsilon_1(s_i, t_i, f) + \dots$$

$$\log(\lambda_i) = \beta_2(c_i, t_i) + \sum_{f=1}^{n_{\omega_2}} L_{\omega_2}(c_i, f) \omega_2(s_i, f) + \sum_{f=1}^{n_{\varepsilon_2}} L_{\varepsilon_2}(c_i, f) \varepsilon_2(s_i, t_i, f) + \dots$$

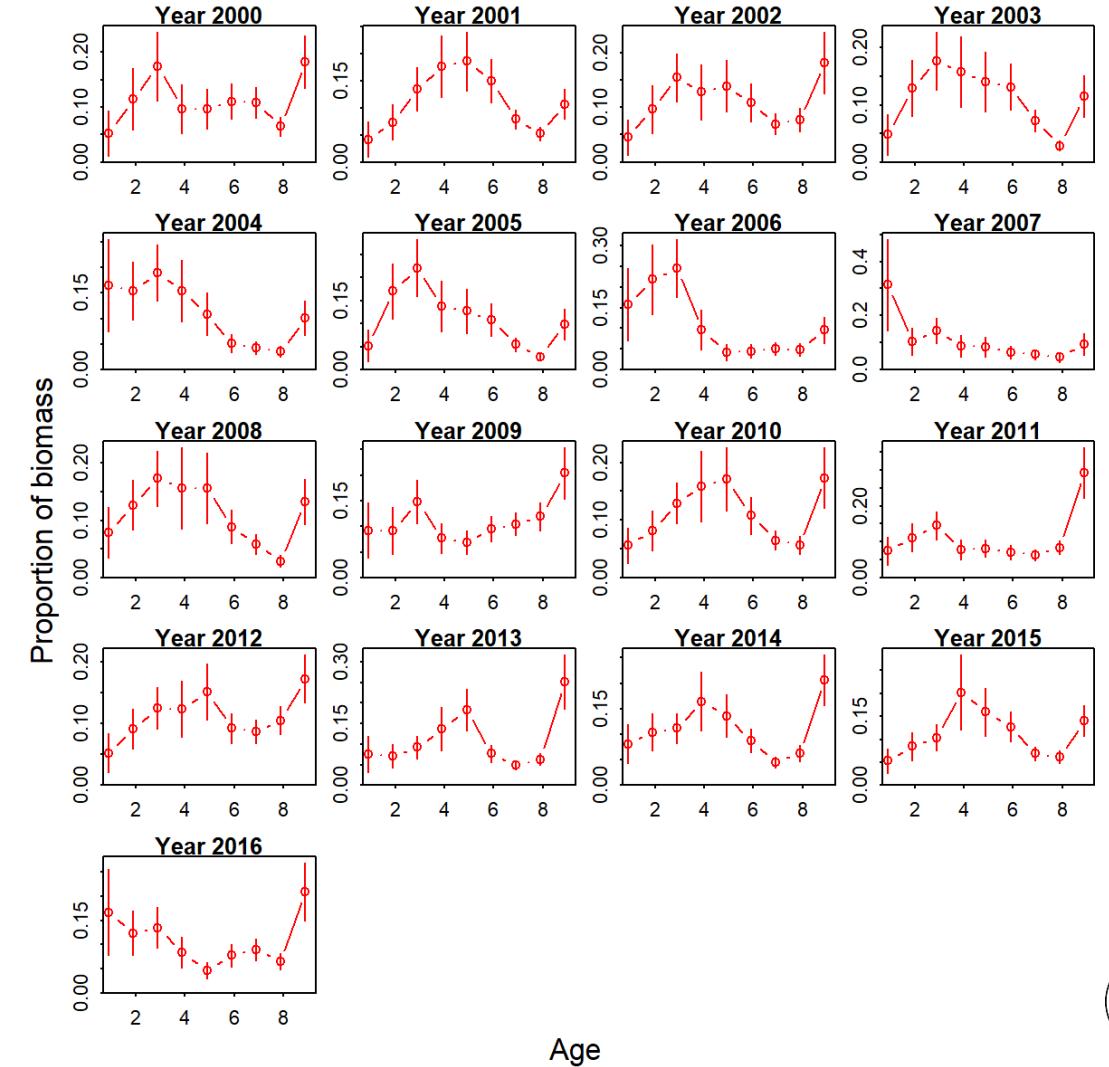
- c_i : length bin number (1-9)
- 2 factors (principle components) for the spatial (n_{ω_1} and n_{ω_2}) and spatiotemporal components (n_{ε_1} and n_{ε_2})

Results: index of abundance and length frequency

Index of abundance by length bin

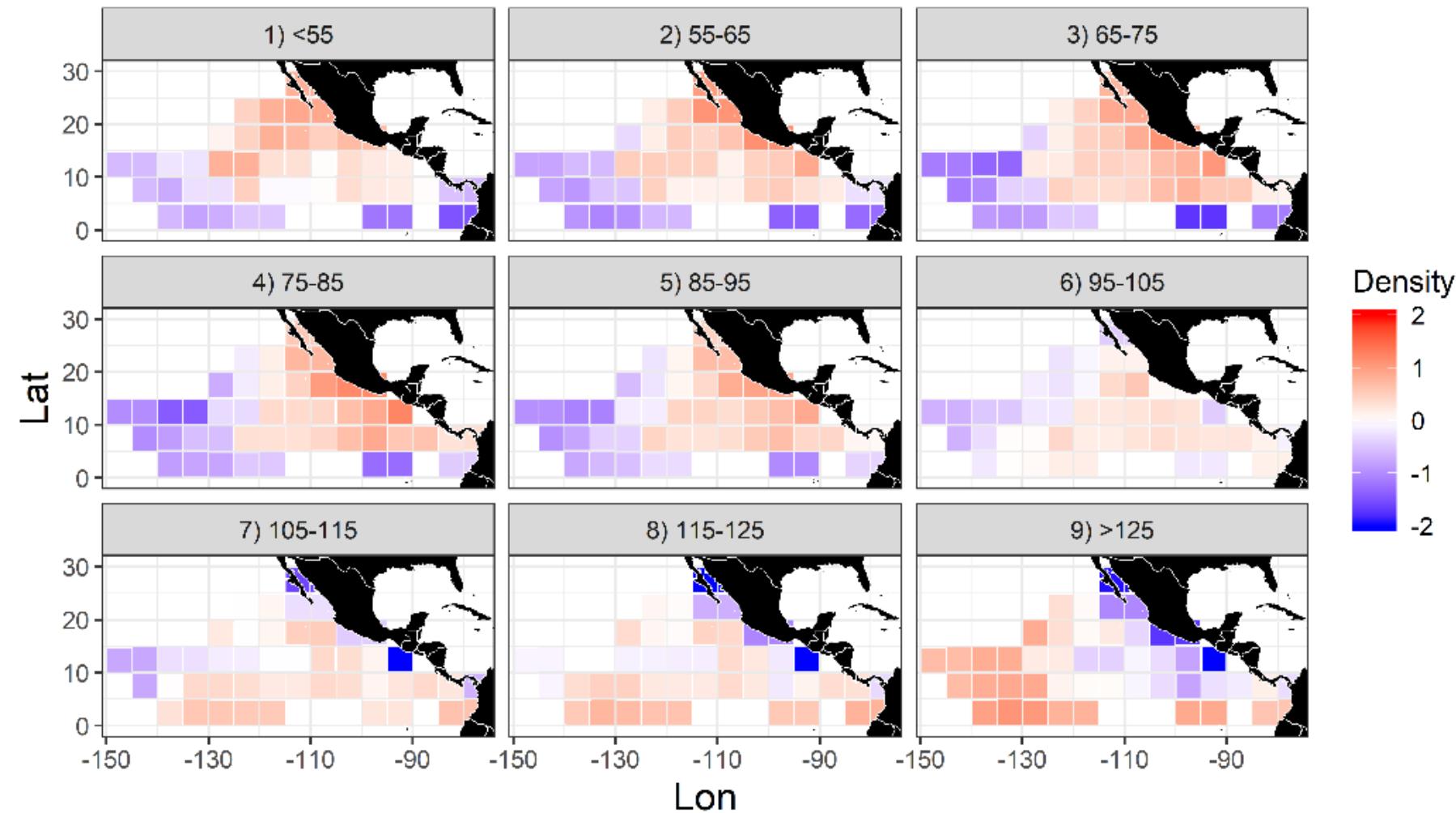


Length frequency by year

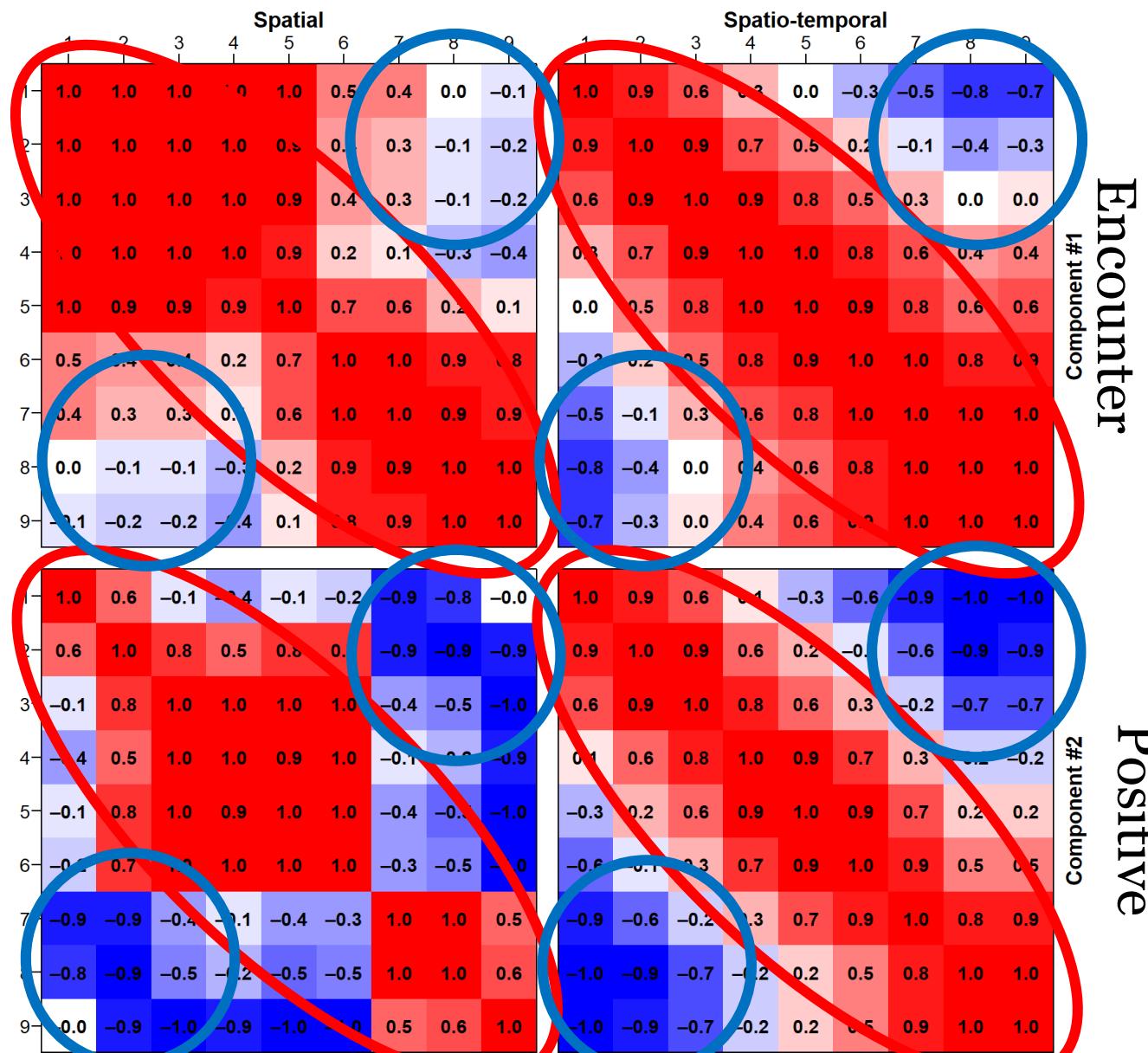


Results: mean log(density) by length bin

Spatial segregation by length



Results: correlation matrix informs segregation



Spatial segregation by size:

- positive correlations between adjacent length bins: density of adjacent length bins tends to change in the same direction
- negative correlations between distant length bins: density of distant length bins tends to change in the opposite direction

Case study: JPN LL fleet in A1

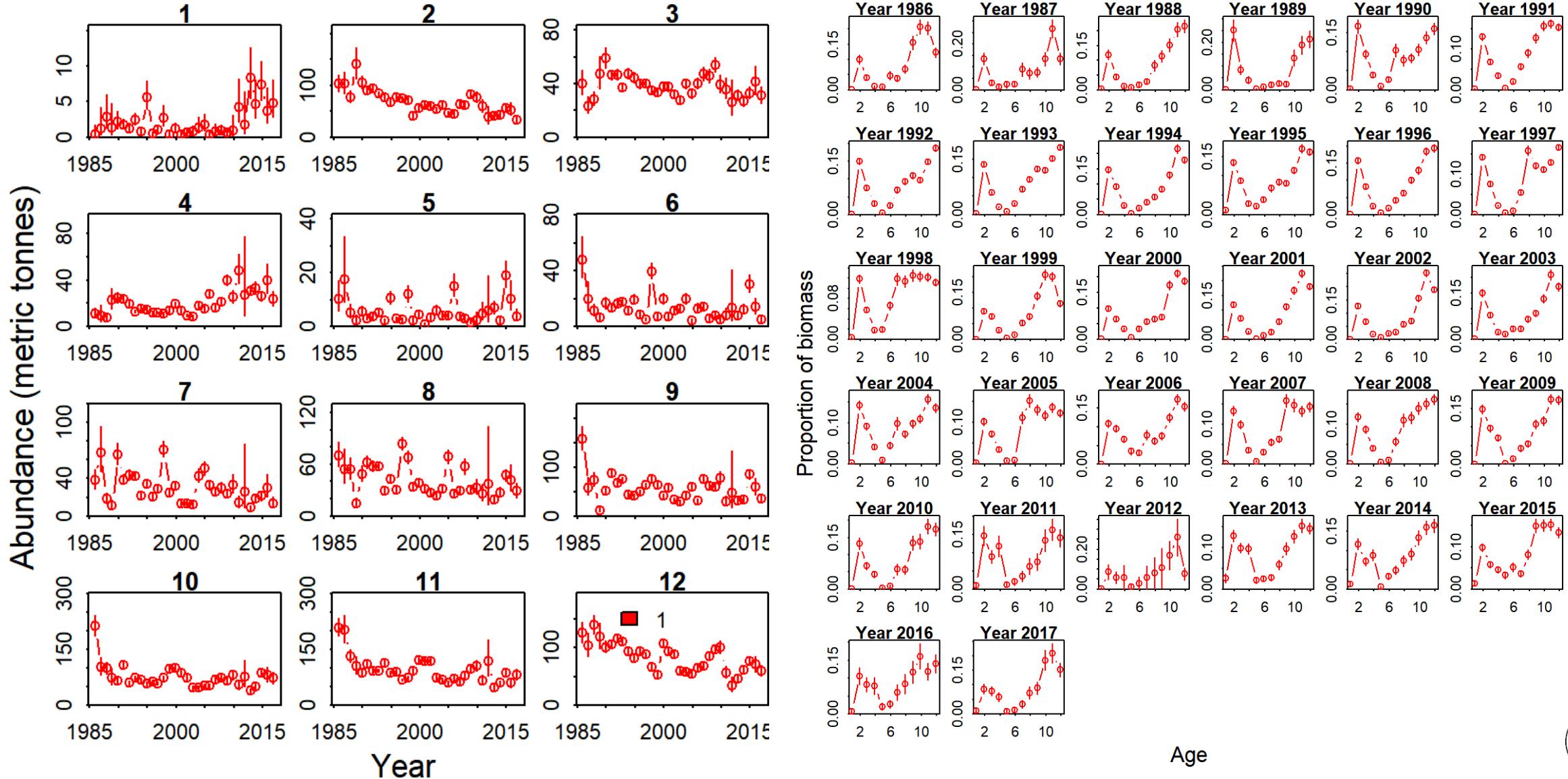
```
spp[which(Data_Geostat_raw$length<80)] <- 1  
spp[which(Data_Geostat_raw$length>=80&Data_Geostat_raw$length<90)] <- 2  
spp[which(Data_Geostat_raw$length>=90&Data_Geostat_raw$length<100)] <- 3  
spp[which(Data_Geostat_raw$length>=100&Data_Geostat_raw$length<110)] <- 4  
spp[which(Data_Geostat_raw$length>=110&Data_Geostat_raw$length<120)] <- 5  
spp[which(Data_Geostat_raw$length>=120&Data_Geostat_raw$length<130)] <- 6  
spp[which(Data_Geostat_raw$length>=130&Data_Geostat_raw$length<140)] <- 7  
spp[which(Data_Geostat_raw$length>=140&Data_Geostat_raw$length<150)] <- 8  
spp[which(Data_Geostat_raw$length>=150&Data_Geostat_raw$length<160)] <- 9  
spp[which(Data_Geostat_raw$length>=160&Data_Geostat_raw$length<170)] <- 10  
spp[which(Data_Geostat_raw$length>=170&Data_Geostat_raw$length<180)] <- 11  
spp[which(Data_Geostat_raw$length>=180)] <- 12
```

Case study: JPN LL fleet in A1

A brief look at the size catch rate data used in this analysis

| Year | Month | Lat | Lon | Catch_Rate | hbf | L1 | L2 | ... | ... | L11 | L12 |
|-------------|--------------|------------|------------|-------------------|------------|-----------|-----------|------------|------------|------------|------------|
| 1986 | | 1 | | 23.38166 | 13 | 0 | 0 | 0 | 0 | 0 | 9 |
| 1986 | | 1 | | 22.1202 | 12.5 | 0 | 0 | 0 | 0 | 0 | 7 |
| 1986 | | 1 | | 26.79236 | 13 | 0 | 0 | 0 | 0 | 5 | 20 |

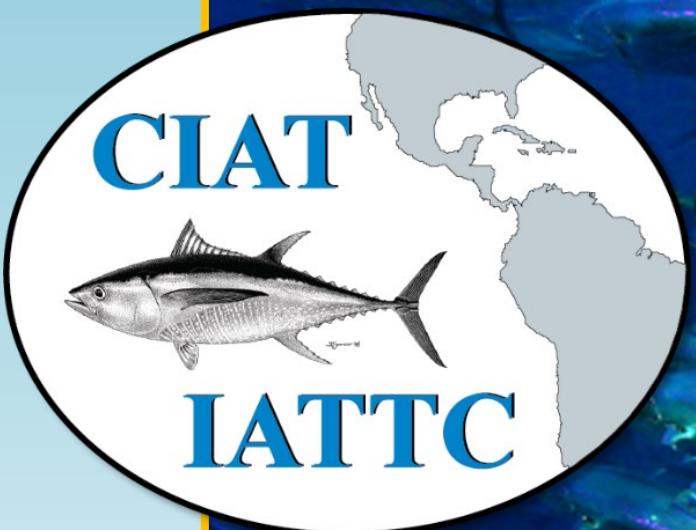
Results: index of abundance and length frequency



Summary

The importance of spatiotemporal modelling using size-specific catch rate observations:

1. Provide size-specific index of abundance (CPUE + LF)
2. Estimate the spatiotemporal distribution for different life stages (e.g., juvenile vs. mature fish)
3. Evaluate the existence of spatial segregation by size



Thank you!

