# **Modeling larval northern anchovy (***Engraulis mordax***)** abundance distributions using fine-resolution imagery data

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## Motivation

- Modeling anchovy abundance distributions will aid in anticipation of climate change effects on populations
- Models must adequately capture mechanistic drivers that underpin response to environment
- Larval anchovy abundance distributions are difficult to predict under novel environmental conditions<sup>1</sup>

## Results

- CV RF r = 0.35, OOB RF r = 0.36
- Depth allowed greatest reduction in node impurities (Fig. 5)
- RF was more likely to predict high anchovy abundance at:
  - Shallow depths
  - High dissolved oxygen
  - High chlorophyll a



Fine-scale abundance data may allow more accurate characterization of mechanistic drivers of abundance



**Fig. 1.** Coupled 1-m<sup>2</sup> and 4-m<sup>2</sup> Multiple Opening/Closing Net and Environmental Sensing System (MOCNESS)



Fig. 2. The In Situ Ichthyoplankton Imaging System (ISIIS), which uses two shadowgraph cameras and environmental sensors to measure fine-scale plankton abundance and physical parameters.



Fig. 3. Clupeiformes larvae imaged by ISIIS, which cannot be distinguished to species in *IS*IIS images.

- Extreme salinities
- High temperatures
- Offshore and extreme inshore locations

Fig. 5. Variable importance plot of predictors included in model fit to randomly selected 75% of ISIIS larval anchovy abundance observations. Bars indicate the total increase in node purity (Gini index) averaged across all trees for each predictor.







### **Data Collection**

- Winter and summer 2018– 2019
- MOCNESS to collect larvae (25-m depth strata)
- In Situ Ichthyoplankton Imaging System (ISIIS) to collect fine-scale (1-m) larval clupeiform abundance in top 100 m across 17 transects (>120 km each)



### **Use MOCNESS to distinguish between ISIIS clupeiform larvae**

• Separate random forests (RFs) were fit to summer 2019 MOCNESS anchovy and sardine abundance data



Longitude

Fig. 6. Partial effect plots for a subset of predictors included in RF fit to randomly selected 75% of ISIIS larval anchovy abundance observations. Hash marks indicate deciles of predictors. Note: Y-axis scales differ among plots.

## **Conclusions and Future Directions**

- Similar performance of CV and OOB RFs suggest negligible influence of autocorrelation on model performance
- Partial effect plots (Fig. 6) support hypothesized effects of most important predictors (*e.g.*, depth, oxygen, chlorophyll)
  - Unexpected effect of salinity
- Low predictive performance may indicate that other ecological factors (*e.g.*, currents, predators) are also important • Time lags may obscure relationships between larval anchovy abundance and environmental conditions
- Predictors: temperature, salinity, depth, latitude, longitude
- Applied ratio of predicted anchovy to sardine abundance during co-occurrence to ISIIS clupeiform abundance

#### **Predict ISIIS larval anchovy abundance**

- RF fit to random 75% of ISIIS larval anchovy (corrected clupeiform) abundance data
  - Out-of-bag (OOB) testing to assess predictive performance
- 4-fold spatially and temporally blocked cross-validation on withheld 25% of data
  - Assess influence of autocorrelation on model performance
- Predictors: temperature, salinity, chlorophyll a, dissolved oxygen, depth, latitude, longitude, season, year

- Zero inflation may be addressed using a two-step model
- Other modeling frameworks (*e.g.*, SDMTMV) may be useful

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#### Reference

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