



Research update: BBRKC bycatch distribution models

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Motivation

- BBRKC distribution is poorly understood outside the NMFS summer survey season
 - Environmental and biological drivers
 - Better understanding is needed to support management and conservation



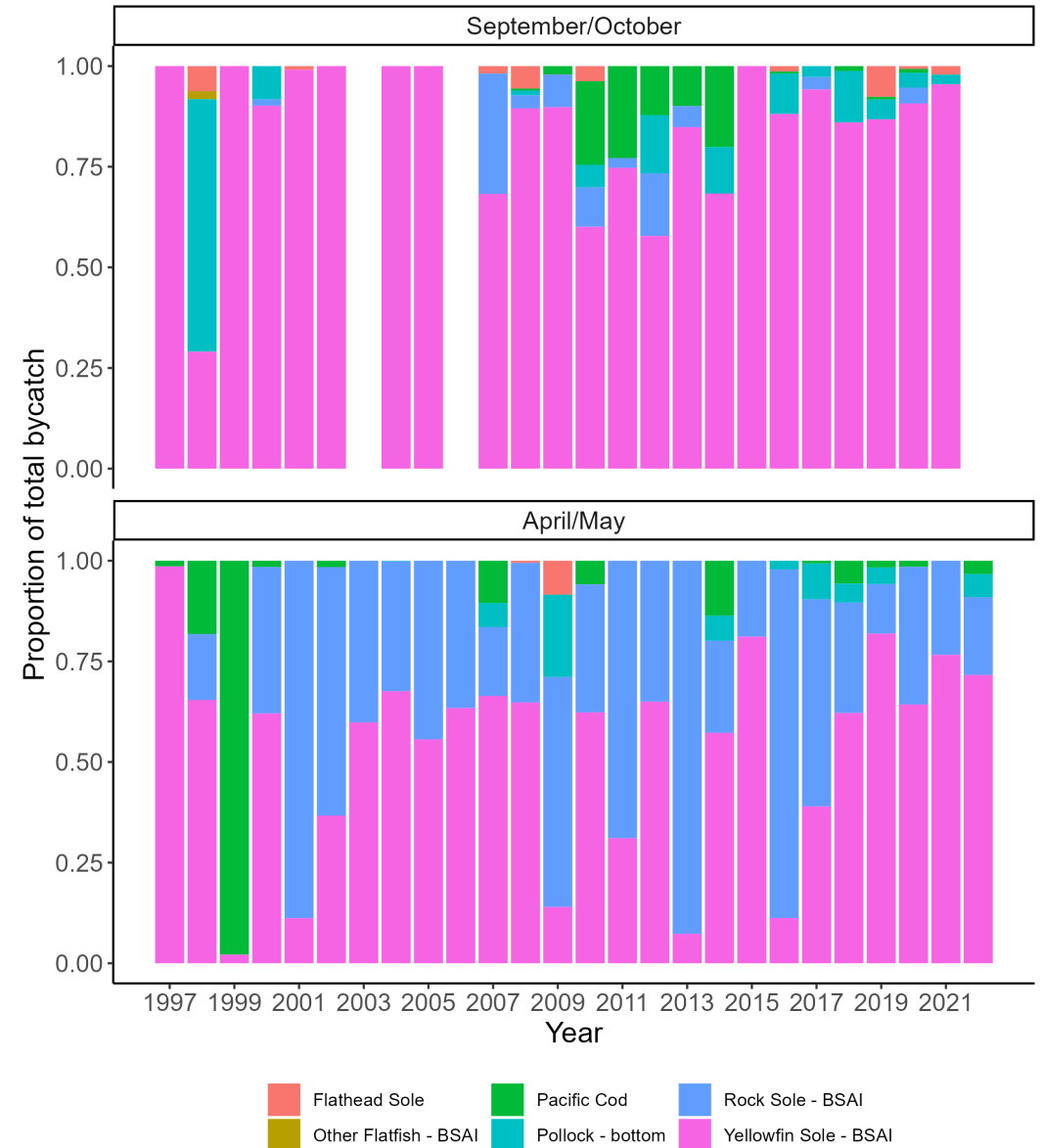
Objectives

1. Create models to predict the distribution of BBRKC bycatch in fall/spring bottom trawl fisheries
2. Identify covariates that drive changes in bycatch distribution from year to year



Data

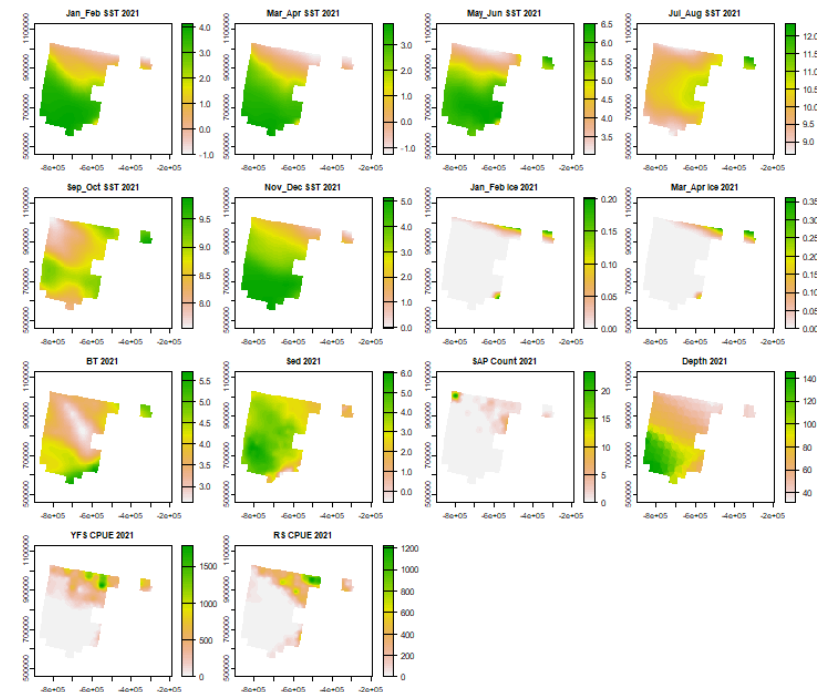
- Response: legal male BBRKC bycatch in two trawl fisheries:
 1. Yellowfin sole (September/October and April/May)
 2. Northern rock sole (April/May)
- Prediction years: 1997-present



Data

- Covariates:

- Surface temperature (°C; two-month increments)¹
- Ice area fraction (% cover; Jan/Feb, Mar/Apr)¹
- Bottom temperature (°C; summer months)²
- Depth (m; constant)²
- BBRKC survey abundance (summer months)²
- Target fish survey abundance (summer months)²
- Sediment grain size (phi; constant)³
- Target fishery quota, TAC (by year)⁴



¹ERA5

²NMFS-AFSC bottom trawl survey

³Smith & McConnaughey 1999

⁴NPFMC SAFEs

Species distribution models (SDMs)

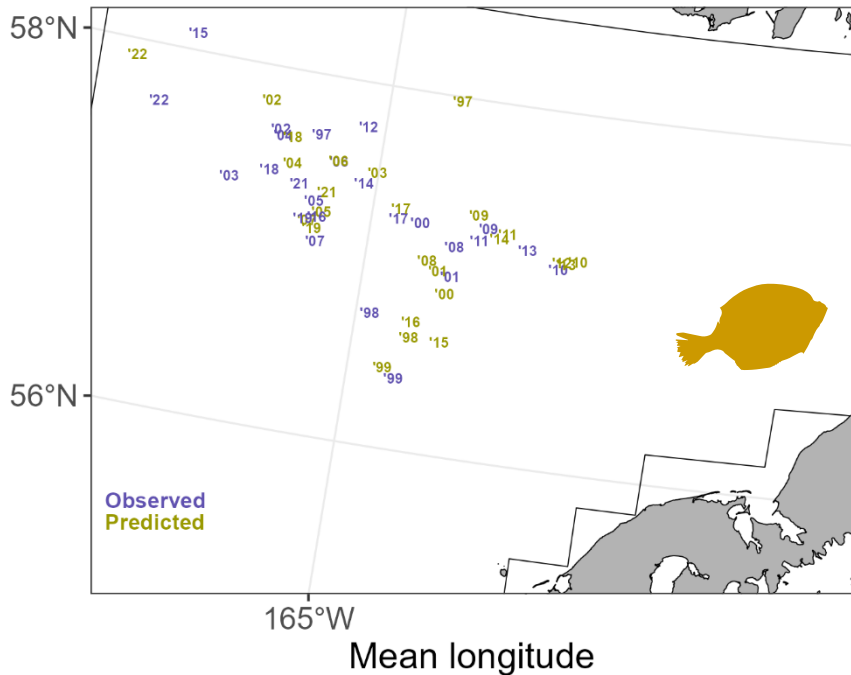
- Algorithm: Boosted Regression Trees
 - Machine learning
 - Estimate non-linear responses through binary splits (“regression trees”)
 - A top-performing SDM algorithm
- Framework: Delta models
 - Two components:
 1. Occurrence (binomial presence/absence, all data)
 2. Abundance (Poisson, only positive catch data)
 - Suitable for zero-inflated data (e.g., survey data)
 - Components evaluated separately

General approach

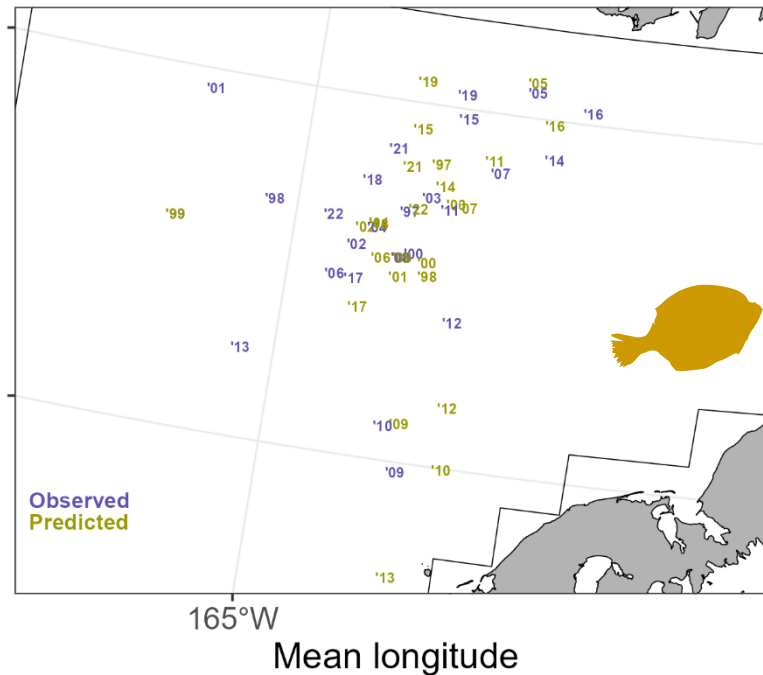
1. Evaluate covariate collinearity
2. Randomly split data into training (80%) and testing (20%)
3. Fit models with training data
4. Evaluate model performance with testing data
 - Occurrence component: AUC-ROC, values ≥ 0.8 indicate excellent predictive ability
 - Abundance component: Spearman's ρ , higher values indicate better predictive ability

Centers of distribution (CODs)

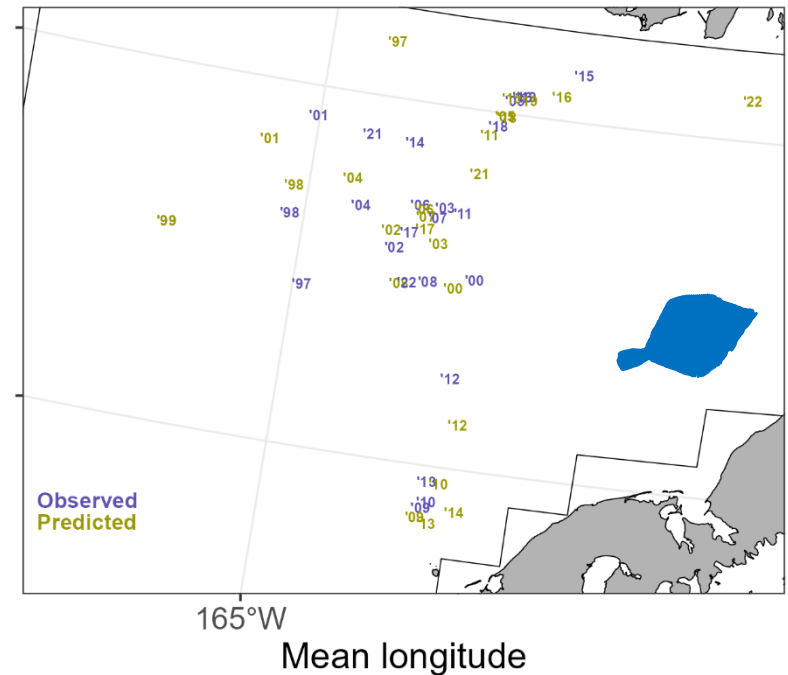
Period: Sep/Oct, Target: Yellowfin Sole



Period: Apr/May, Target: Yellowfin Sole



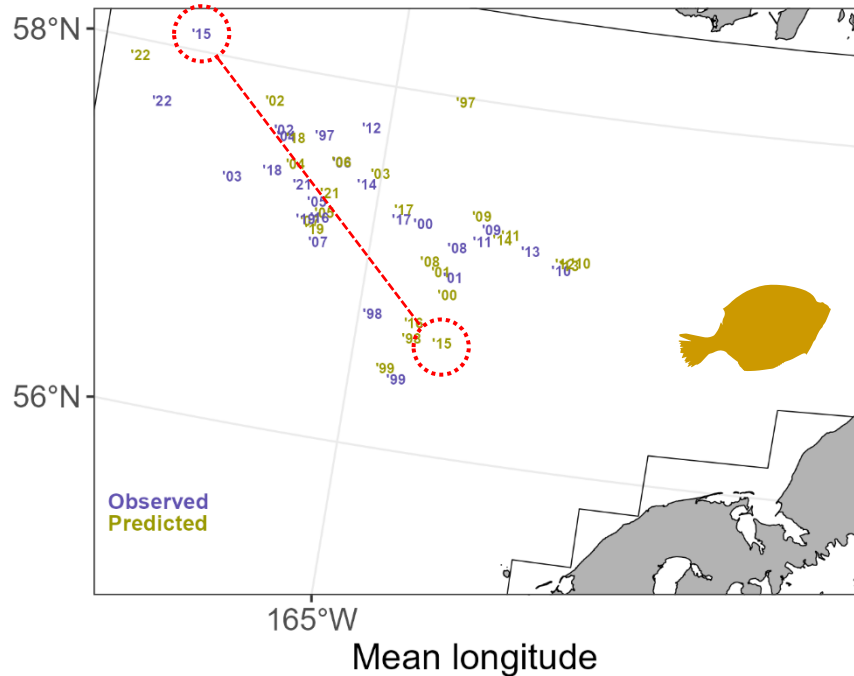
Period: Apr/May, Target: Rock Sole



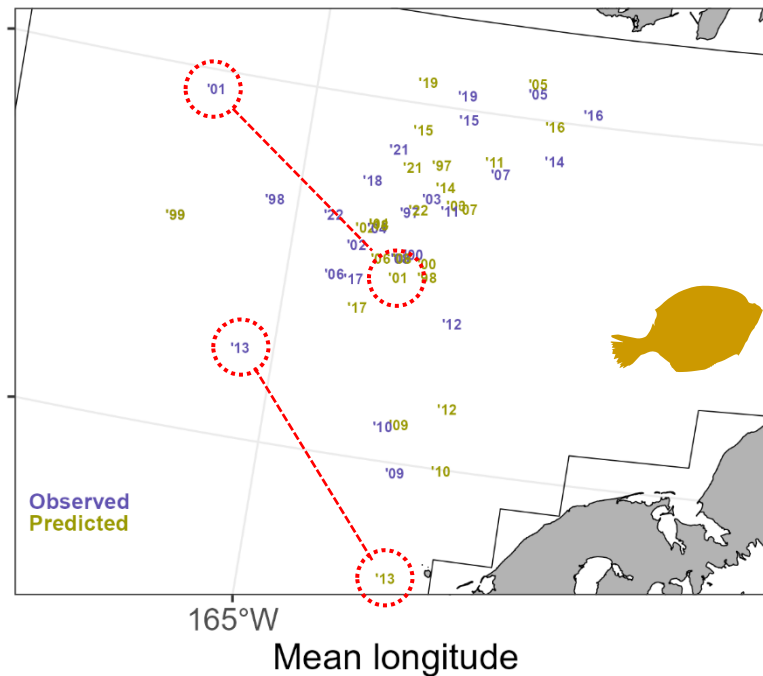
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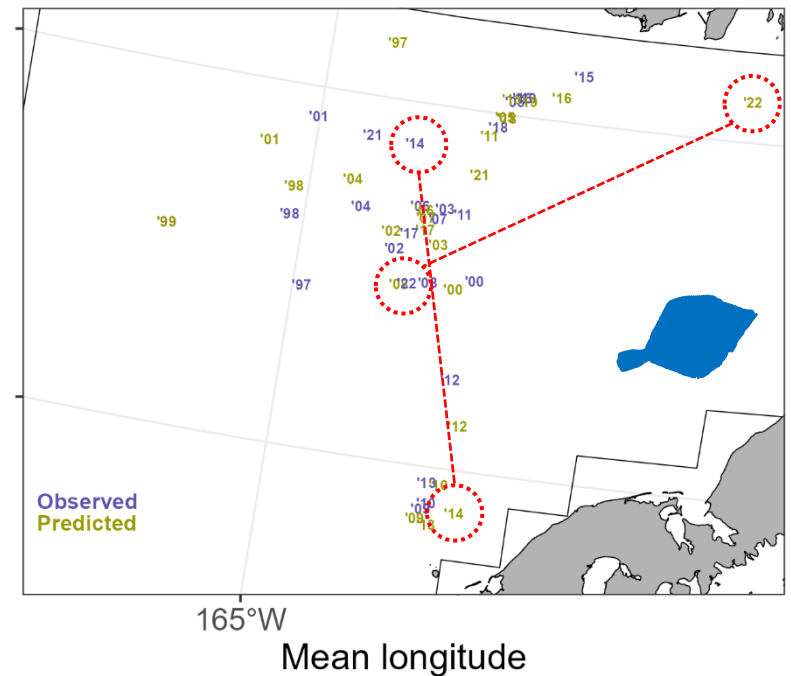
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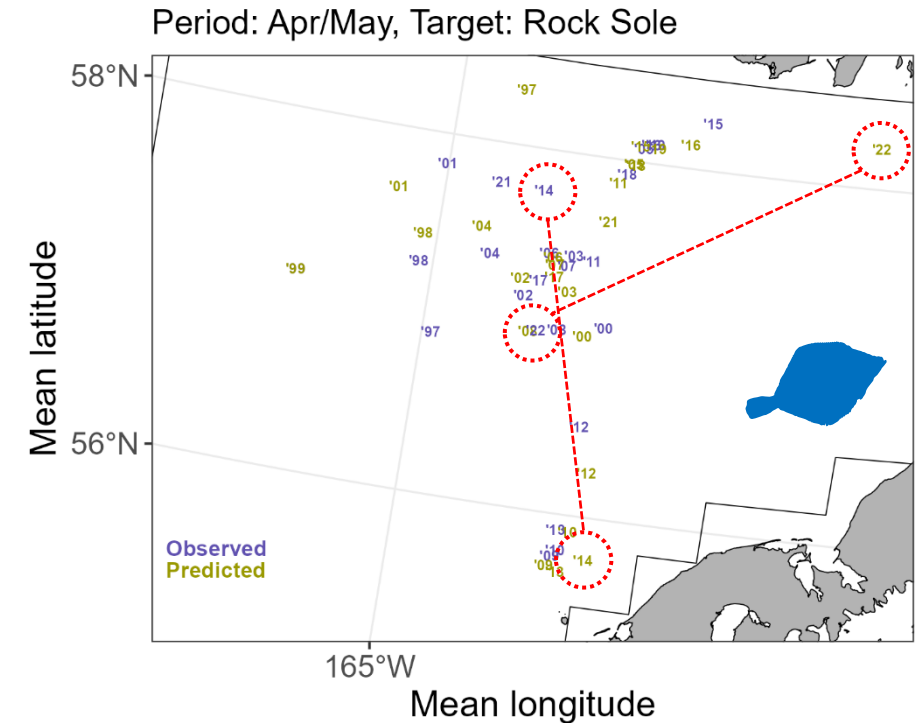
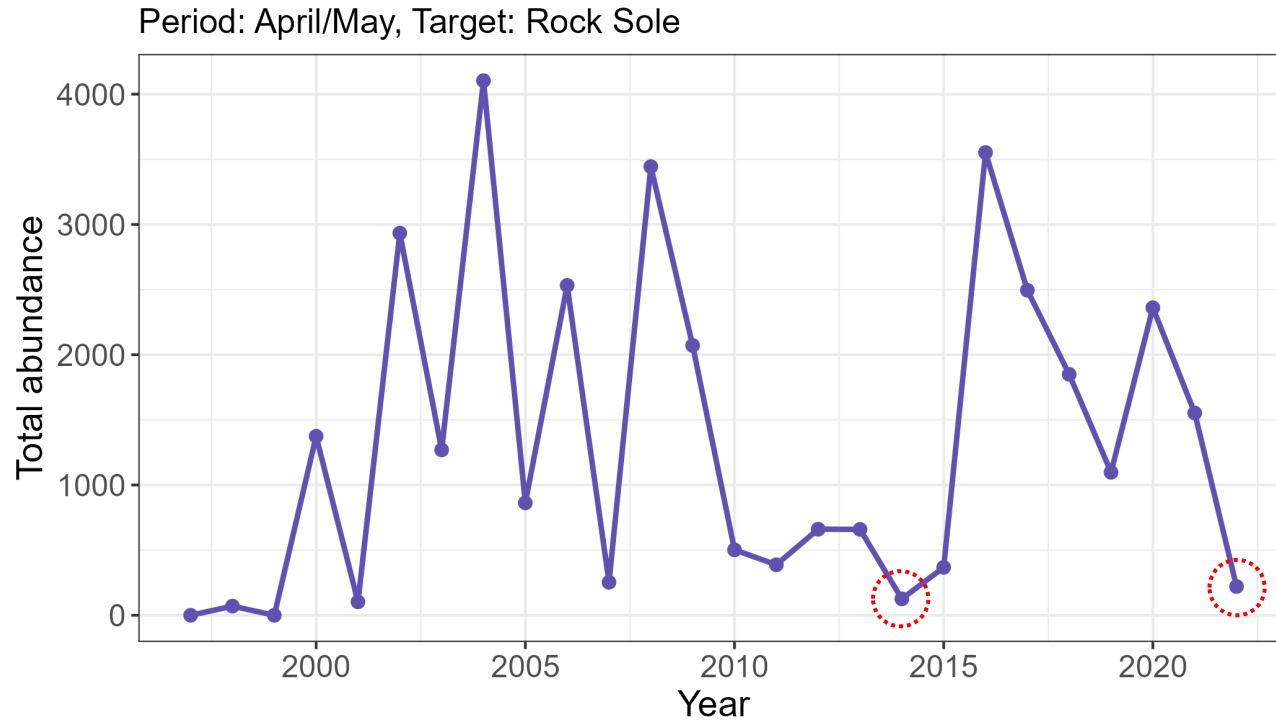


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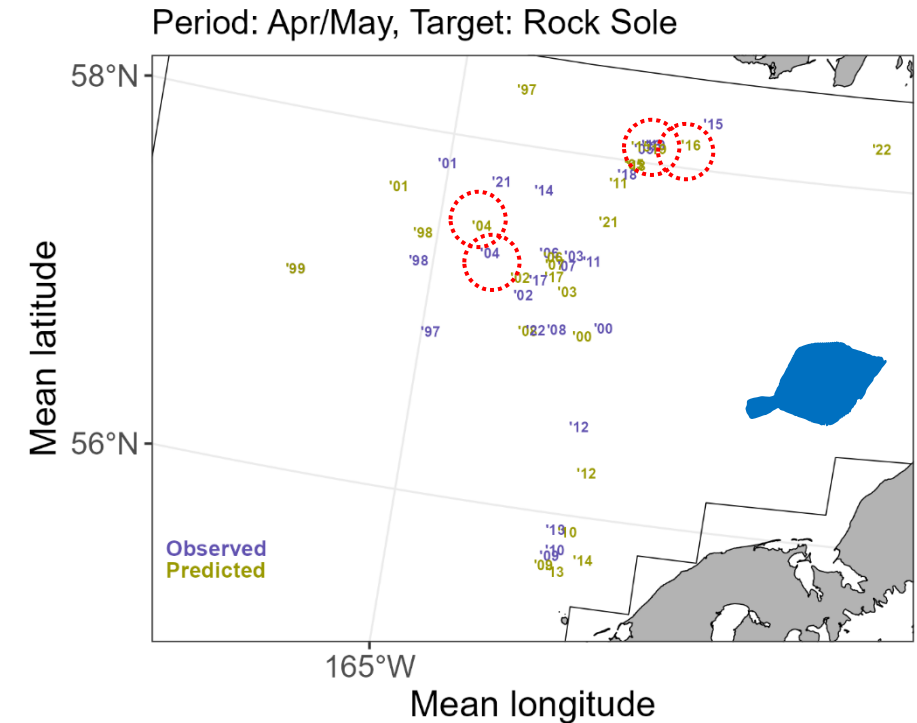
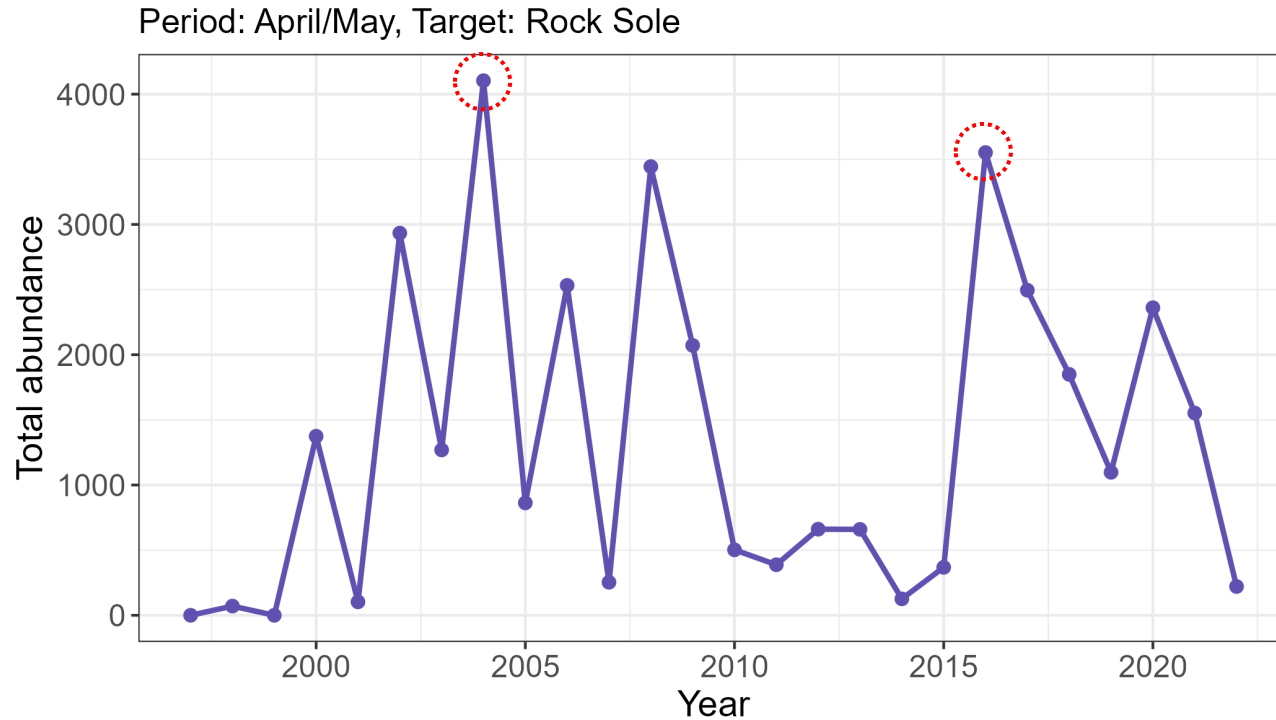
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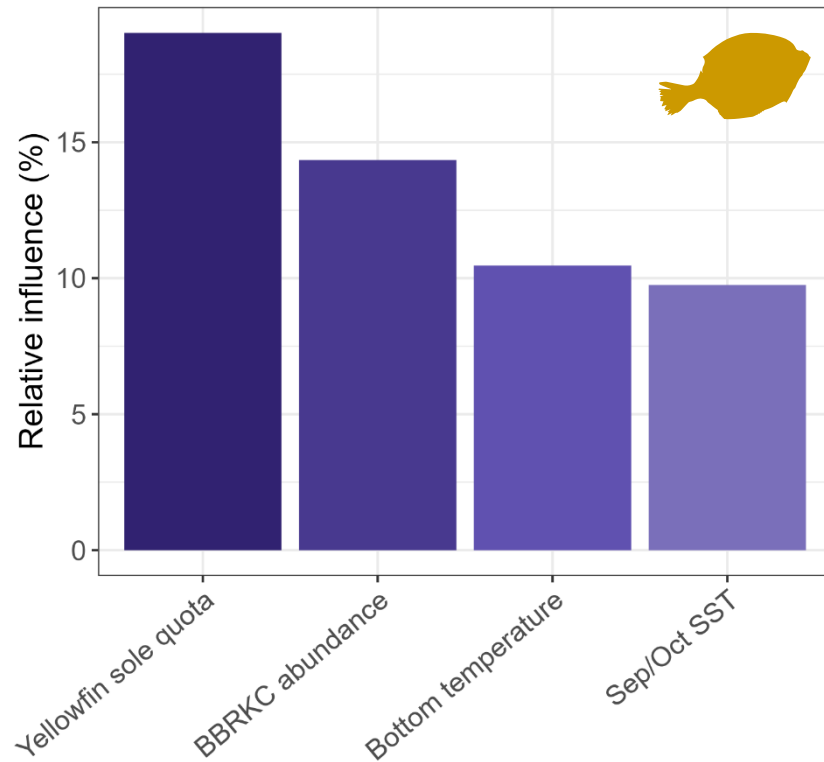
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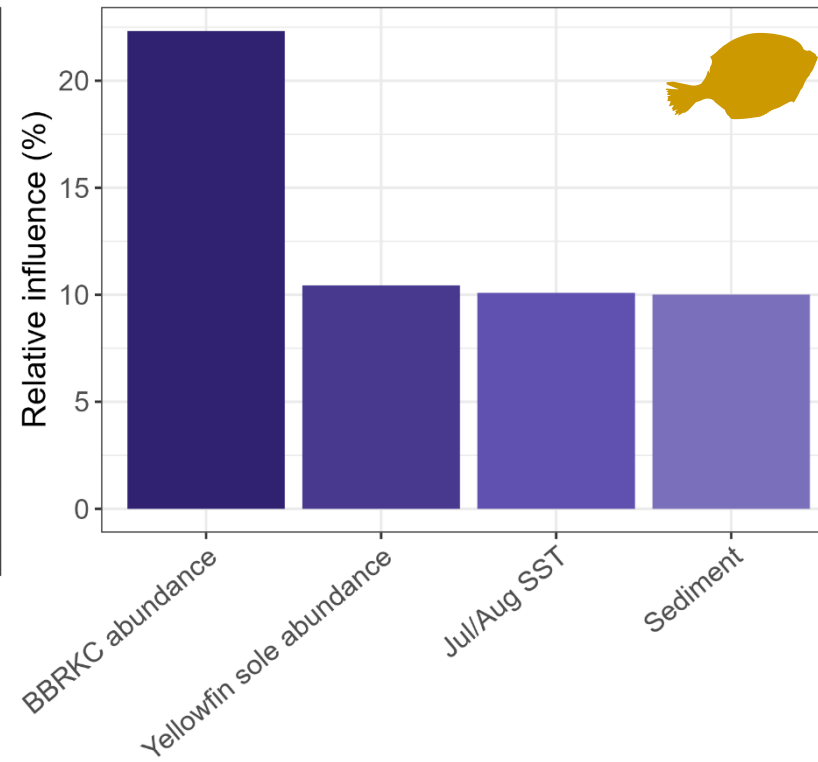
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Leading covariates

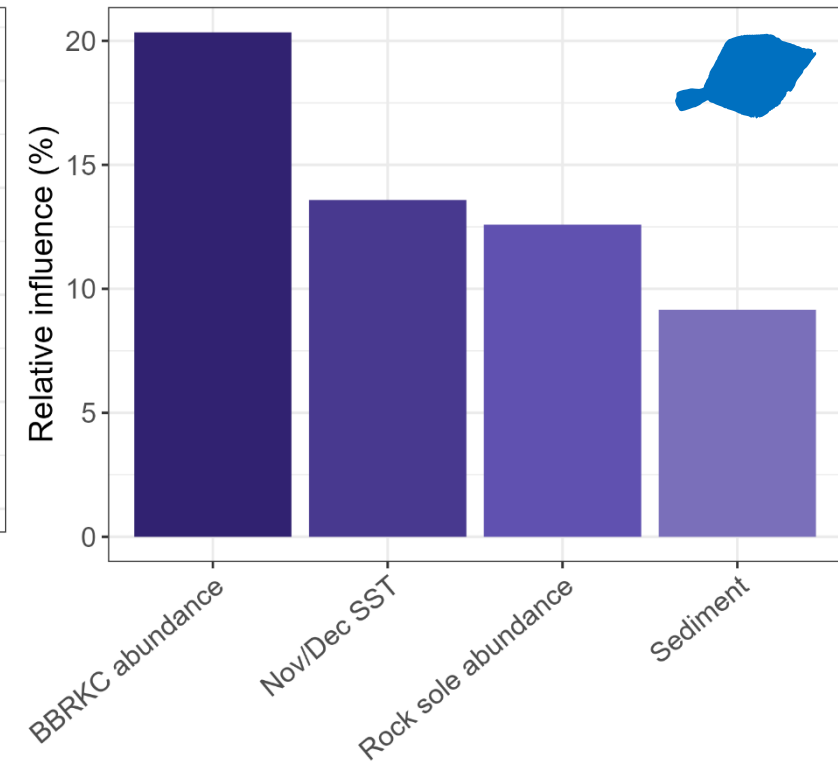
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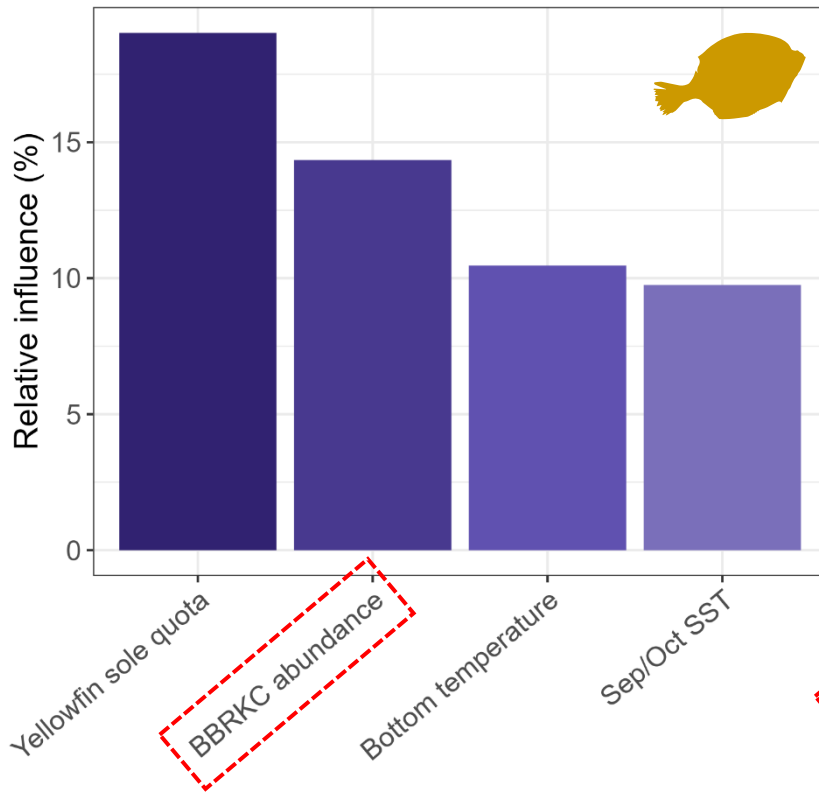
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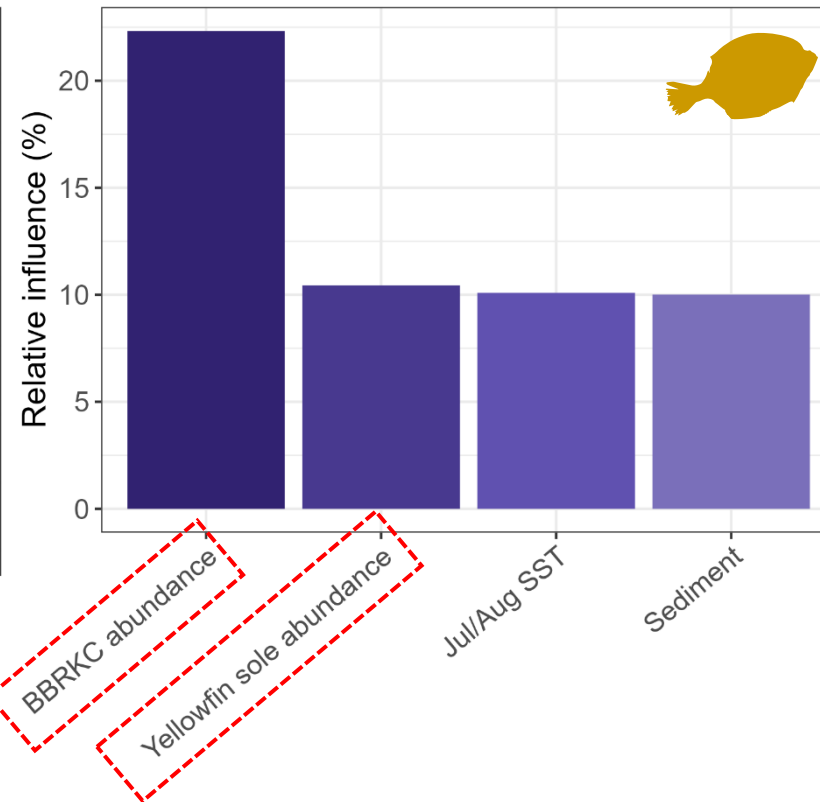
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- Target fish survey abundance and sediment also highly influential for April/May models
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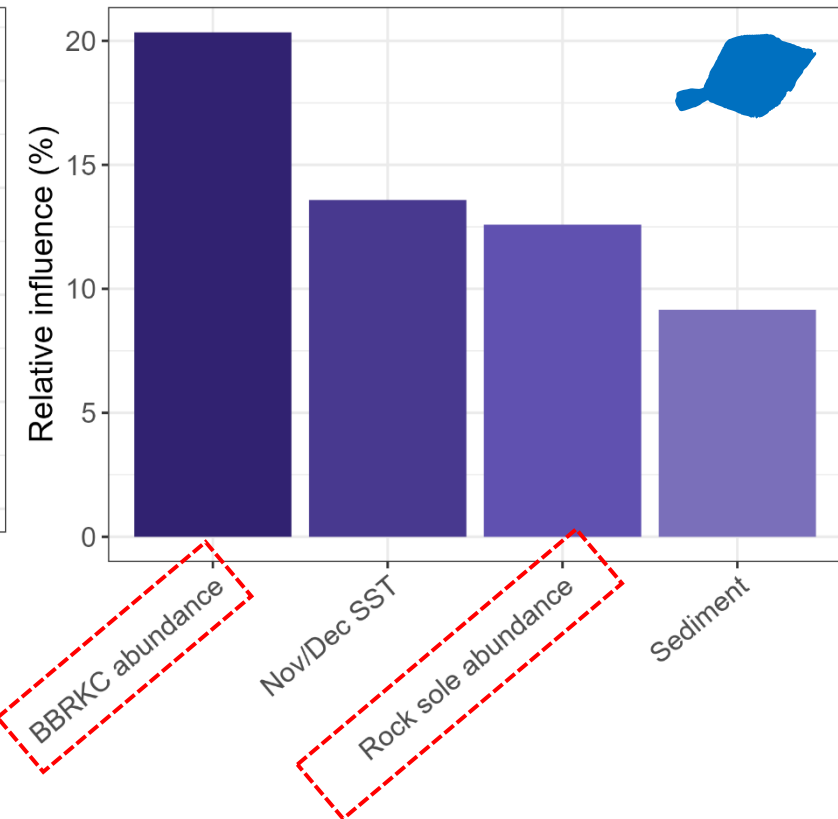
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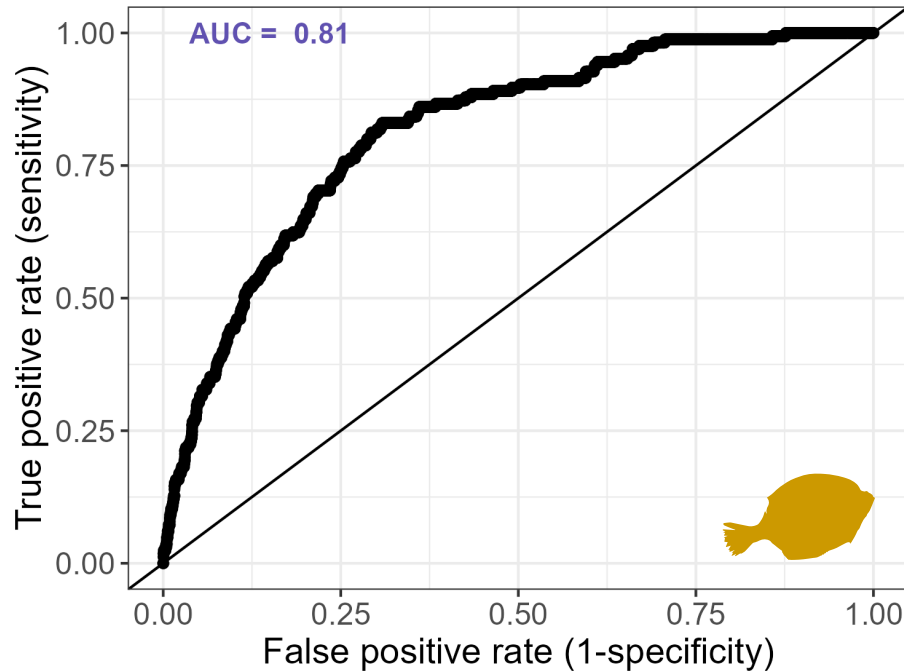
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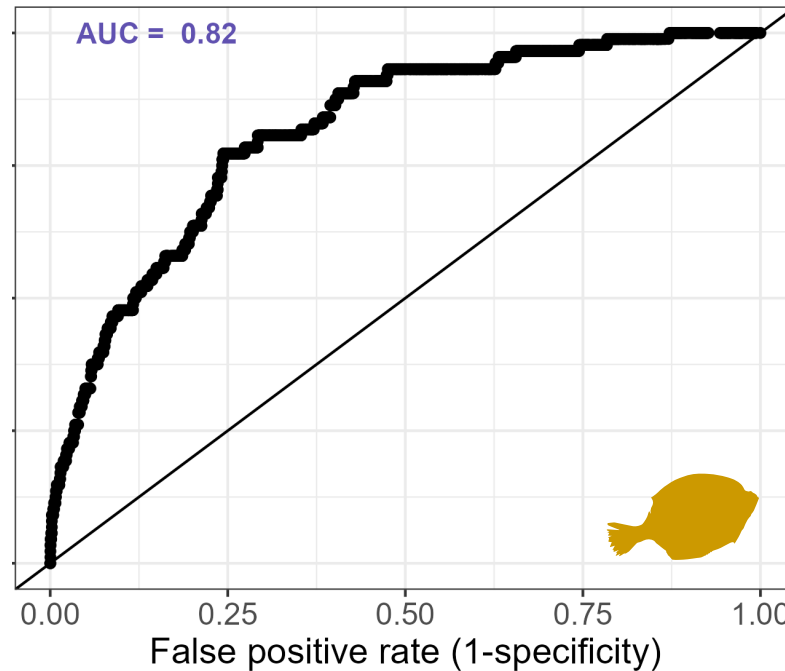
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Occurrence model evaluation

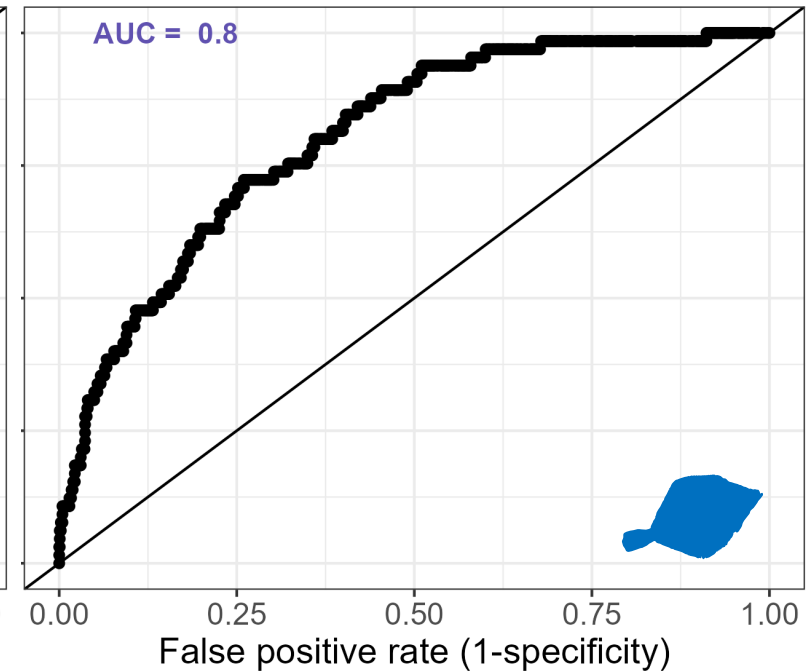
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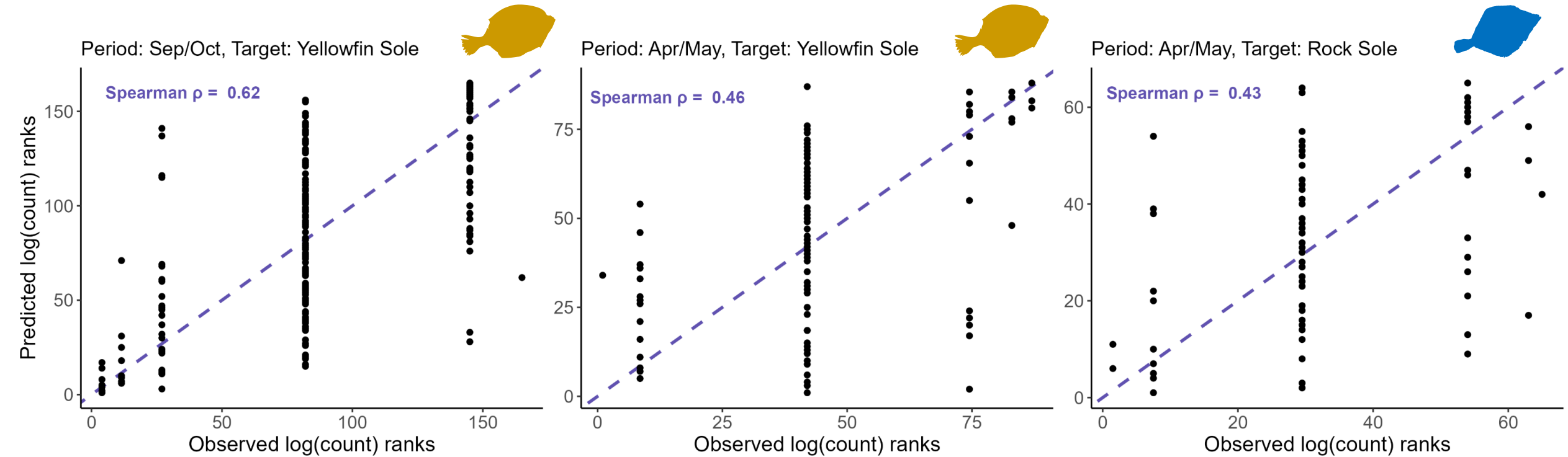


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- Models exhibit excellent ability to correctly predict species occurrence (AUCs ≥ 0.8)
- Yellowfin sole models slightly better than the rock sole model

Abundance model evaluation



- The Sept/Oct yellowfin sole performed better at predicting abundance, though all models performed relatively well

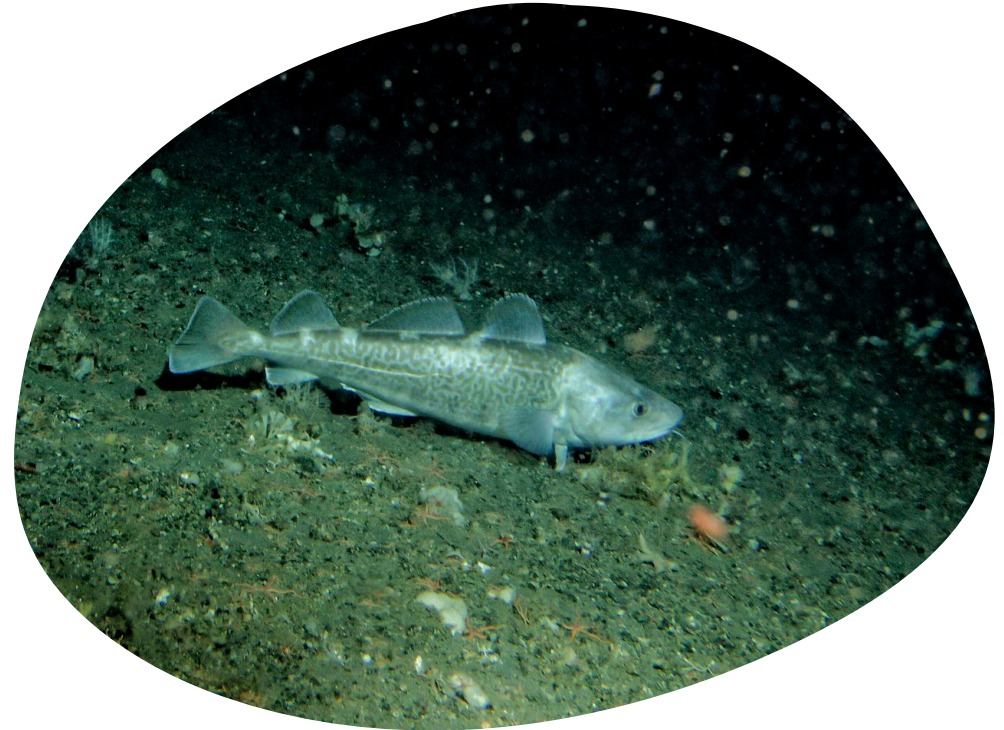
Conclusions

- SDMs may be useful tools for predicting BBRKC bycatch
 - Models performed well in predicting observed bycatch magnitude and spatial distribution
- Survey data is important
 - Survey-estimated BBRKC and target fish abundance highly influential covariates
- Differences in covariate importance for bycatch in different seasons/fisheries should be explored



Next steps

- Develop bycatch prediction models for BBRKC mature females and the pot cod fishery
- Develop SDMs to predict BBRKC distribution in data-poor fall/winter seasons





Thank you!

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