## Spatial prediction of fisheries bycatch

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## Growing interest in spatial models

Spatial semiparametric models improve estimates of species abundance and distribution

Andrew Olar Shelton, James T. Thorson, Eric J. Ward, and Rlake E. Feist


Yields abundance estimates that are:

- More precise
- More biologically reasonable
- Extreme catch events
- Sampling locations


## Growing interest in spatial models

## Used by NWFSC assessment team



Static management


## Motivation

Static management


## Motivation

## Dynamic management



## Static management



Polovina et al. 2015

## Motivation

## Dynamic management


avoid fishing north of solid black $65.5^{\circ} \mathrm{F}$ line to reduce turtle interactions


## PIFSC

Polovina et al. 2015

## Research Questions

1. How well can we predict fisheries bycatch in space and time?

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Pr(some bycatch)
Binomial

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1. How well can we predict fisheries bycatch in space and time?

Pr(some bycatch)
E(bycatch | some bycatch)

Binomial
Positive

## Research Questions

2. What type of spatial model best predicts bycatch?

1) Shared
2) Constant

3) Multiple years

Parametric

- INLA-SPDE
- GAM

Non-parametric

- Random Forest
- SVM


## Research Questions

## 3. Does the answer depend on species traits?



Benthic
Benthic
Low
Low
Bycatch Rate:
29\%
18\%


Habitat:
Movement:
Benthic
Med


Habitat:


Pelagic
Movement:
Bycatch Rate:

Pelagic High
$0.15 \%$
$0.3 \%$


Pelagic High
0.18\%

## Research Questions

## 3. Does the answer depend on species traits?



Benthic
Benthic
Low
Low
Movement:
Bycatch Rate:


Habitat:
Pelagic


Pelagic
Movement:
Bycatch Rate:
High
High
High
89\%
$0.15 \%$
$0.18 \%$

## Research Questions

## 3. Does the answer depend on species traits?

| Habitat: | Benthic | Benthic | Benthic |
| ---: | :---: | :---: | :---: |
| Movement: | Med | Low | Low |
| Bycatch Rate: | $29 \%$ | $18 \%$ | $0.3 \%$ |
|  |  |  |  |
| Habitat: | Pelagic | Pelagic | Pelagic |
| Movement: | High | High | High |
| Bycatch Rate: | $89 \%$ | $0.15 \%$ | $0.18 \%$ |

## West Coast Groundfish

## Binomial

Positive
~ SSt + depth +
distance to rocky substrate + size of rocky patch + in Rockfish Conservation Area + gear type + predicted occurrence (survey) + spatial field

## Hawaii Longline

## Binomial

Positive
~ sst (observed) + target + spatial field

## Spatial field: INLA-SPDE

## Spatial Partial Differential Equation

- Discrete approximation of continuous spatial fields



## Results: Shared model

## Binomial



## Results: Multiple years

## Binomial



## Results: Shared model

## Binomial



## Results: Shared model

## Binomial



## Results: Shared model

## Binomial



## Results: ROC



## Results: ROC



## Results: ROC

## Binomial



Eliminate $20 \%$ of fishing
Reduction in bycatch:

-     - = $20 \%$
- $45 \%$
- $85 \%$
- 100\%


## Results: ROC

## Binomial



## Conclusions

1. How well can we predict fisheries bycatch in space and time?

Well enough to be useful for management
2. What type of spatial model best predicts bycatch?
3. Does the answer depend on species traits?

Depends on amount of data and bycatch rate

## Acknowledgements

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- Brice Semmens

NWFSC


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- West Coast Groundfish Observer Program (Jason Jannot)

SWFSC

- Tomo Eguchi

PIFSC

- Hawaii Longline Observer Program (Eric Forney)


## Results (preliminary)

Table 2. Probability of occurrence (binomial model, test data)

|  | $\begin{aligned} & \text { DBRK } \\ & \text { (18\%) } \end{aligned}$ |  | $\begin{aligned} & \text { PHLB } \\ & \text { (28\%) } \end{aligned}$ |  | $\begin{aligned} & \text { YEYE } \\ & \text { (0.4\%) } \end{aligned}$ |  | $\begin{aligned} & \text { LOGG } \\ & (0.15 \%) \end{aligned}$ |  | $\begin{aligned} & \text { LEATH } \\ & (0.18 \%) \end{aligned}$ |  | $\begin{aligned} & \text { BLUE } \\ & \text { (89\%) } \end{aligned}$ |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Method | AUC | F | AUC | F | AUC | F | AUC | F | AUC | F | AUC | F |
| INLA |  |  |  |  |  |  |  |  |  |  |  |  |
| Shared | . 843 |  | . 820 |  | . 775 |  | . 923 |  | . 795 |  | . 740 |  |
| Constant | . 849 |  | . 826 |  | . 774 |  | --- |  | --- |  | . 749 |  |
| Fixed | . 863 |  | . 790 |  | . 774 |  | --- |  | --- |  | --- |  |
| AR | . 862 |  | . 790 |  | . 774 |  | --- |  | --- |  | . 684 |  |
| GAM |  |  |  |  |  |  |  |  |  |  |  |  |
| Null (GLM) | . 799 |  | . 704 |  | . 762 |  | . 924 |  | . 797 |  | . 672 |  |
| Shared | . 845 |  | . 818 |  | . 766 |  | . 931 |  | . 847 |  | . 739 |  |
| Constant | . 851 |  | . 826 |  | . 776 |  | . 938 |  | . 820 |  | . 749 |  |
| Fixed | . 864 |  | . 848 |  | . 653 |  | . 947 |  | . 677 |  | . 762 |  |
| Random Forest |  |  |  |  |  |  |  |  |  |  |  |  |
| Constant | . 881 |  | . 874 |  | . 743 |  | . 592 |  | . 627 |  | . 780 |  |
| SMOTE | . 879 |  | . 871 |  | . 794 |  | . 953 |  | . 704 |  | . 781 |  |
| Downsample | . 874 |  | . 869 |  | . 788 |  | . 946 |  | . 836 |  | . 795 |  |

## Results: ROC

## Binomial



## Fisheries Observer Data

West Coast Groundfish

- 2002-2013
- 55,835 tows
- 1.7 million records


Hawaii Longline

- 1994-2014
- 70,297 sets
- 3.2 million records



## Results: RCAs

## Binomial

## 11\% of tows were in Rockfish Conservation Areas



## Q: What about the positive model?



## Q: What about effort?

## Results: ROC (survey)

## Binomial



## Proof of Concept

## Binomial



## Proof of Concept

## Positive



## Proof of Concept

## Positive



These extreme bycatch events are the most important to predict!

## Proof of Concept

## Positive



## Proof of Concept

## Positive



## Proof of Concept

## Positive



## Spatial models: INLA-SPDE

Integrated $\mathbb{N}$ ested $\llcorner$ Laplace $\mathbb{A} p$ proximation

- Alternative to MCMC for Bayesian inference
- Much faster

1. Find the posterior mode
2. Calculate local curvature
3. Use $N$ (mode, curvature)


## Spatial models: INLA-SPDE

Spatial Partial Differential Equation

- Discrete approximation of continuous spatial fields


# Preliminary results: ROC curves 



False positive rate 0.30
True positive rate 0.70

## Preliminary results: ROC curves



False positive rate 0.48
True positive rate
0.52

False positive rate 0.30
True positive rate 0.70

## Preliminary results: ROC curves

This PDF was later amended to make the document 508 compliant.


