Spatial prediction of fisheries bycatch

Brian Stock, Eric Ward, Tomo Eguchi SIO/UCSD, NWFSC, SWFSC



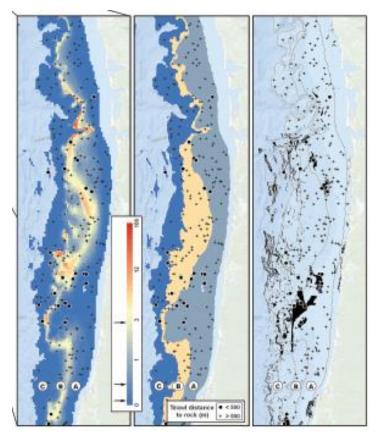


Growing interest in spatial models

Spatial semiparametric models improve estimates of species abundance and distribution

Andrew Olaf Shelton, James T. Thorson, Eric J. Ward, and Blake E. Feist





Yields abundance estimates that are:

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

Shelton et al. (2014)

Growing interest in spatial models

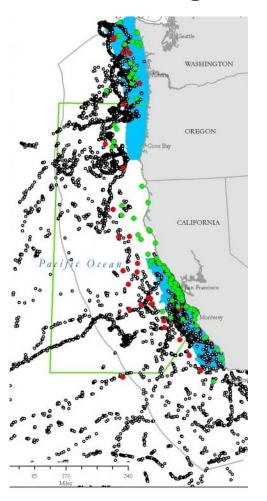
Used by NWFSC assessment team

GitHub This repository Search	Explore	Features	Enterprise	orise Pricing				
nwfsc-assess / geostatistical_delta-GLMM				• Watch	9			

Tool for geostatistical analysis of survey data, for use when estimating an index of abundance

🕝 145 commits	ဖို 1 branch	🟷 8 releases	ses 😳 2 contributors		
ी Branch: master -	geostatistical_delta-GLMM / +		E		
Hames-Thorson fixed	bug in mean_D_tl computation		Latest commit 6c99fa7 11 hours ago		
🖿 R	fixed bug in V3i		a day ago		
ata data	adding South African grid		7 days ago		
examples	added V3i		a day ago		

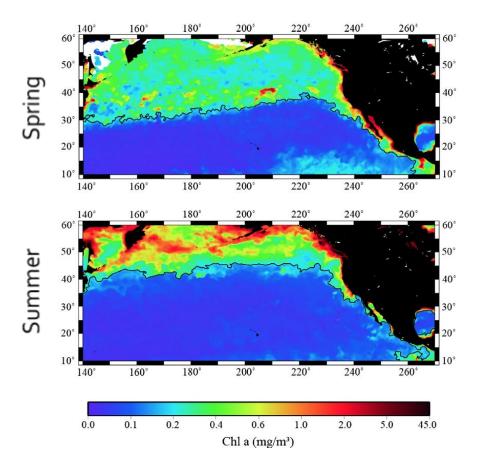
Static management



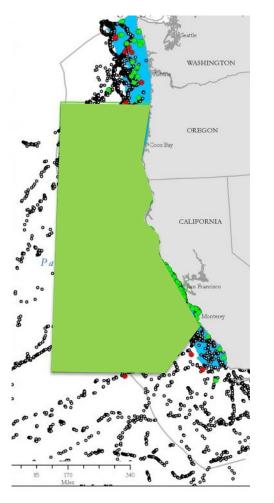
Static management



Dynamic management

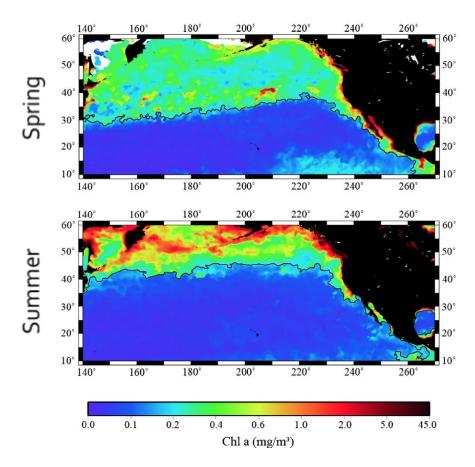


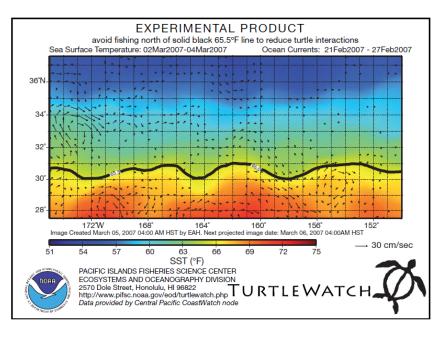
Static management



Polovina et al. 2015

Dynamic management





PIFSC

Polovina et al. 2015

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

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Pr(some bycatch)

Binomial

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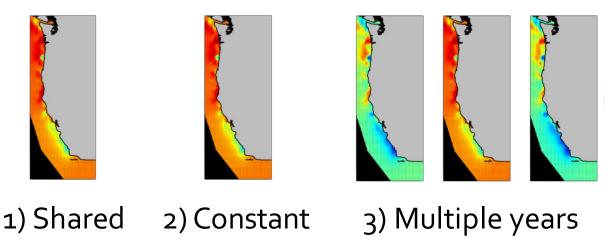
Pr(some bycatch)

E(bycatch | some bycatch)

Binomial

Positive

2. What *type* of spatial model best predicts bycatch?



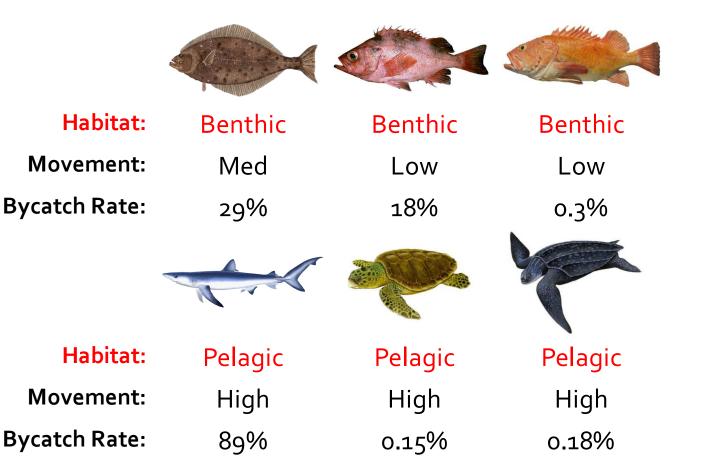
Parametric

- INLA-SPDE
- GAM

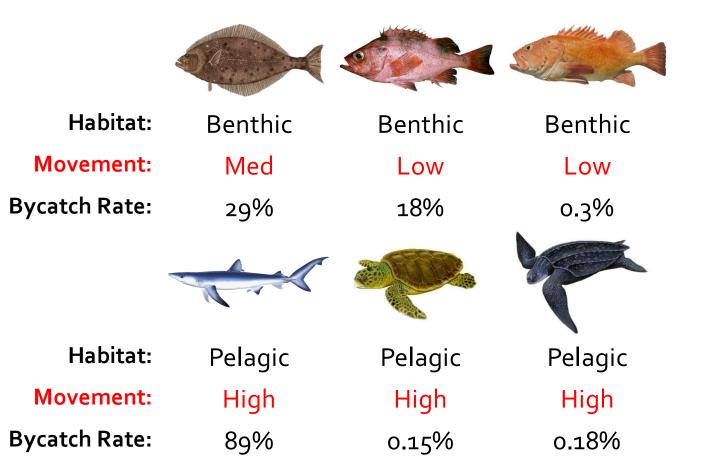
Non-parametric

- Random Forest
- SVM

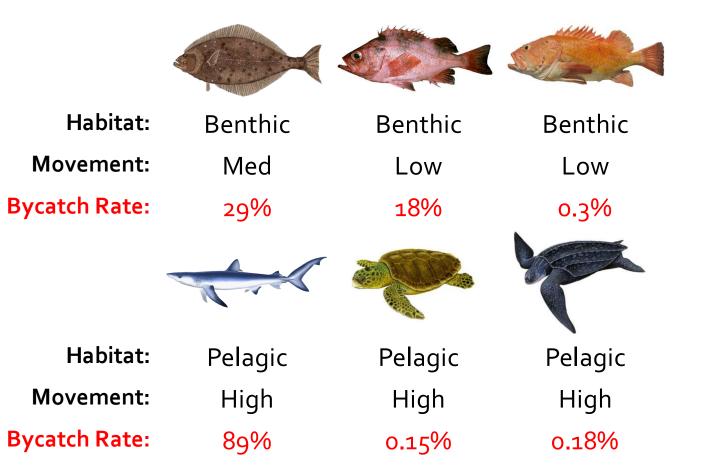
3. Does the answer depend on *species traits*?



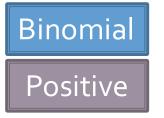
3. Does the answer depend on *species traits*?



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



West Coast Groundfish



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

Hawaii Longline

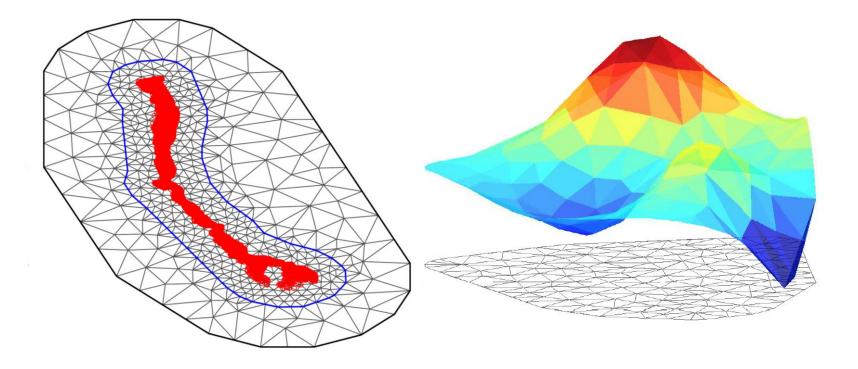


~ sst (observed) + target + spatial field

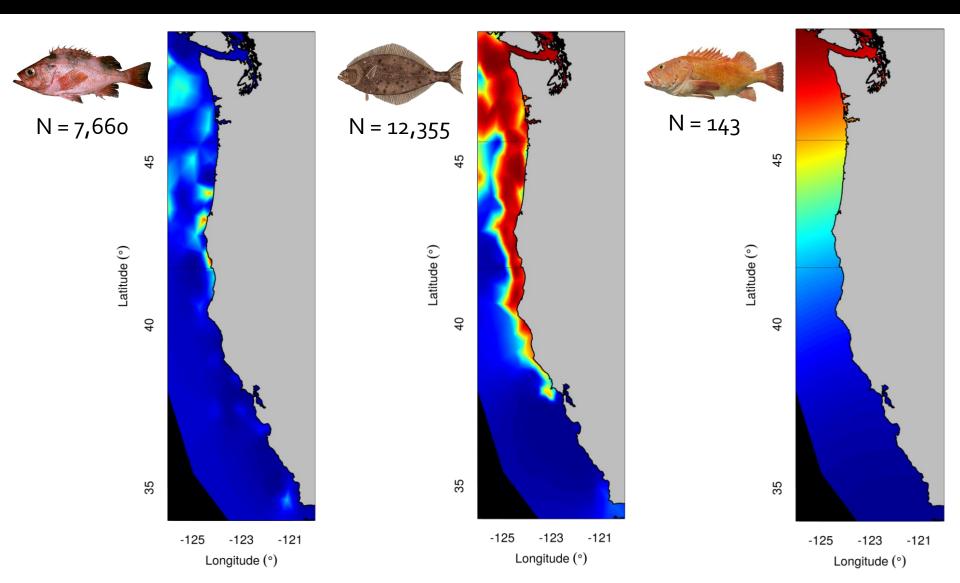
Spatial field: INLA-SPDE

Spatial Partial Differential Equation

Discrete approximation of continuous spatial fields

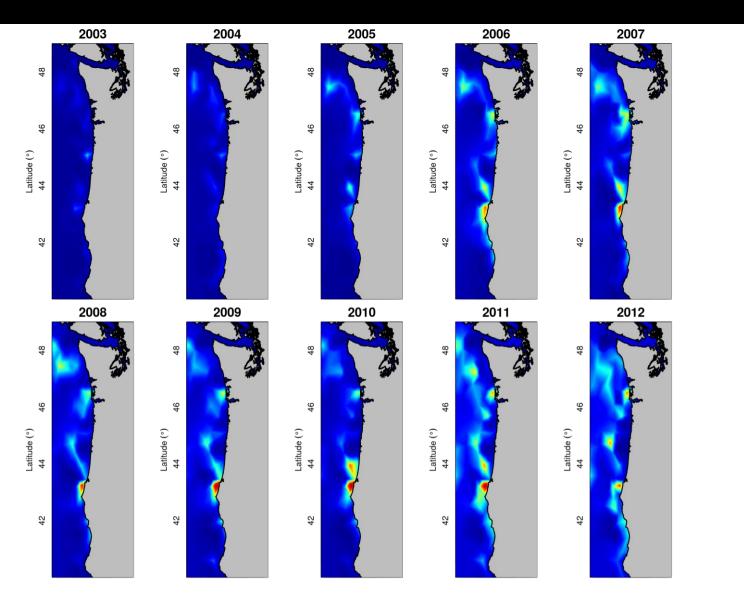


Binomial

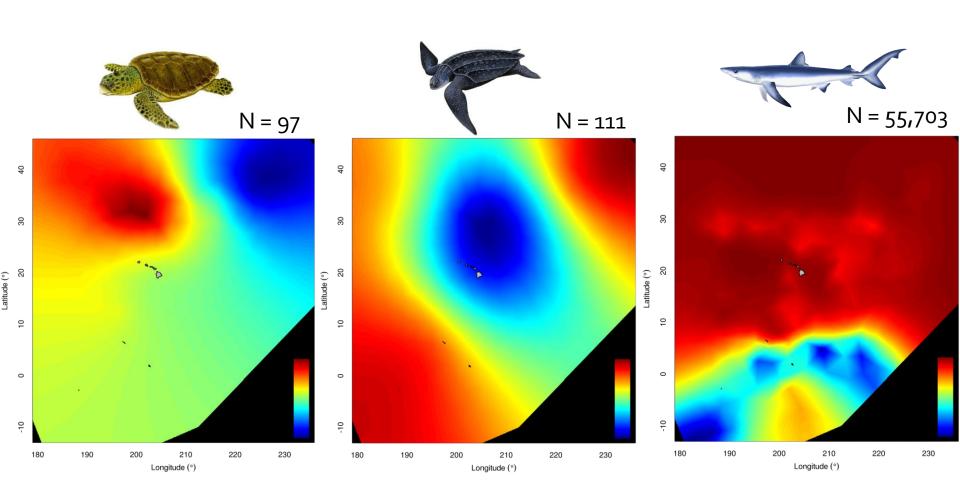


Results: Multiple years

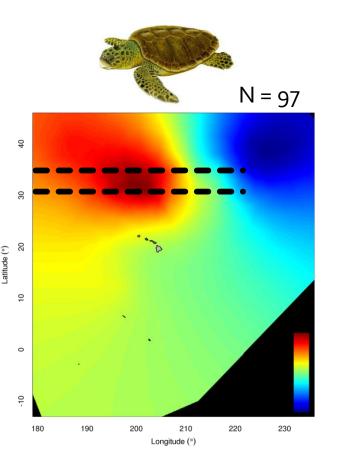


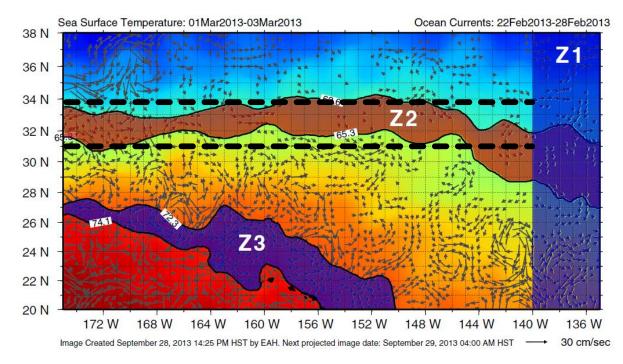




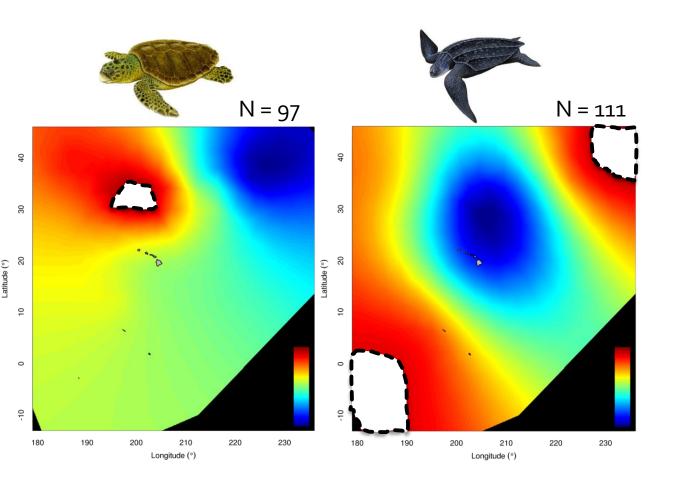


Binomial



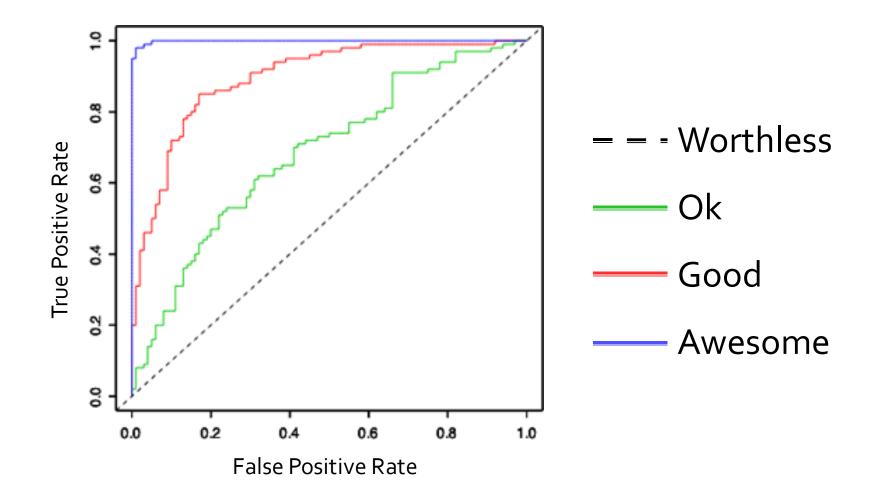




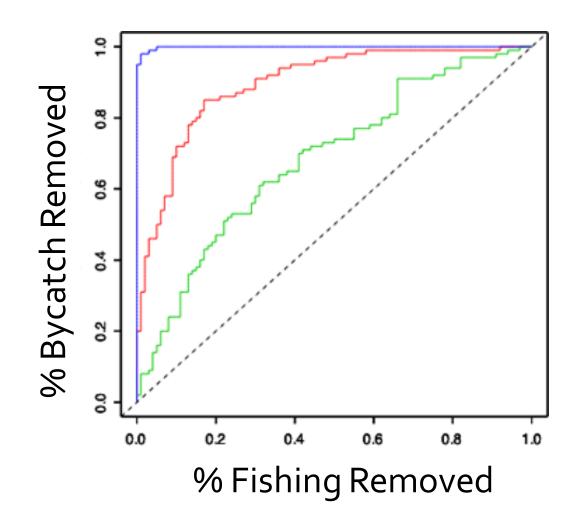


So what?

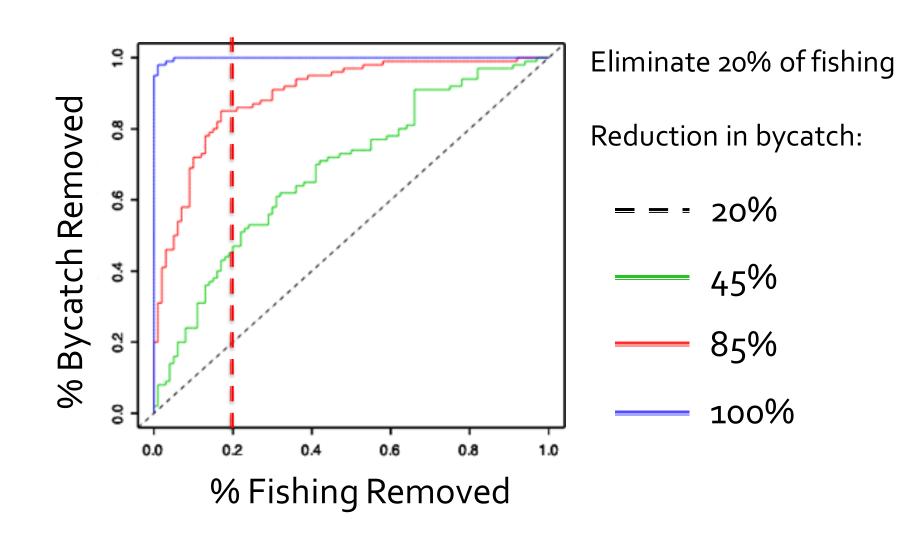




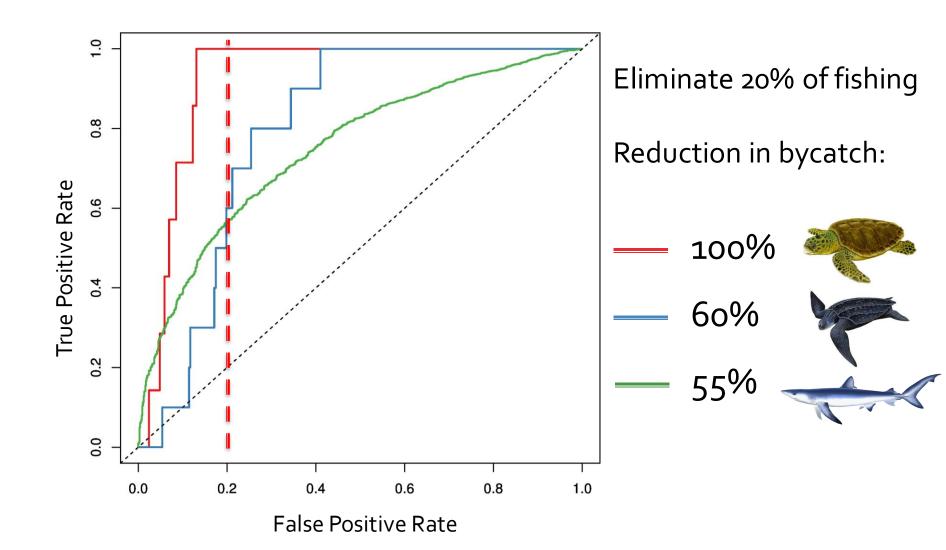












Conclusions

- How well can we predict fisheries bycatch in space and time? Well enough to be useful for management
- What *type* of spatial model best predicts bycatch?
 Does the answer depend on *species traits*?
 Depends on amount of data and bycatch rate

Acknowledgements

SIO

Brice Semmens
 NWFSC

UCSD ATMOSPHERIC TO MUSERATION OF COMMERCIAN

- Eric Ward
- Essential Fish Habitat (Blake Feist)
- West Coast Groundfish Observer Program (Jason Jannot)
 SWFSC
 - Tomo Eguchi
- PIFSC
 - Hawaii Longline Observer Program (Eric Forney)

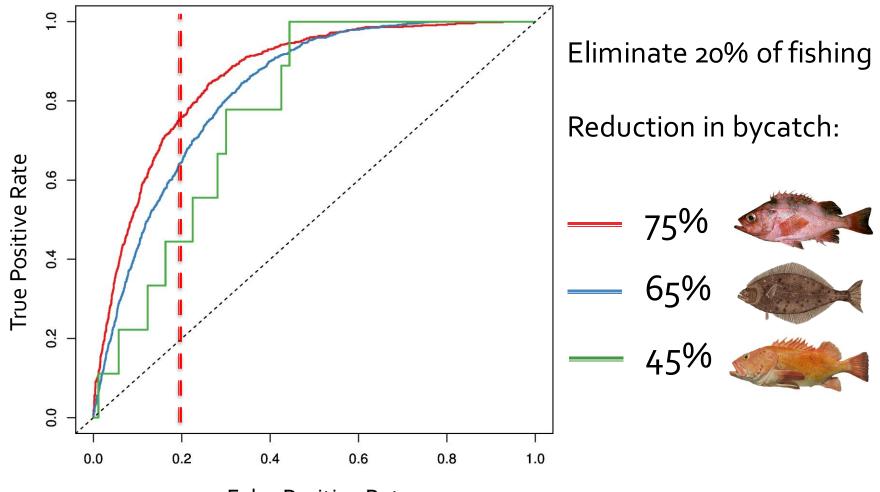
Results (preliminary)



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

	DIZ				VE		~~		ATL1	DI		
1 1											BLUE (89%)	
-	-		•									
AUC	F	AUC	F	AUC	F	AUC	F	AUC	F	AUC	F	
.843		.820		.775		.923		.795		.740		
.849		.826		.774						.749		
.863		.790		.774								
.862		.790		.774						.684		
.799		.704		.762		.924		.797		.672		
.845		.818		.766		.931		.847		.739		
.851		.826		.776		.938		.820		.749		
.864		.848		.653		.947		.677		.762	_	
											1	
.881		.874	1	.743		.592		.627		.780		
.879		.871		.794		.953		.704		.781		
.874		.869		.788		.946		.836	1	.795		
	(18 AUC .843 .849 .863 .862 .862 .851 .851 .851 .864 .881 .879	.843 .849 .863 .862 .862 .862 .851 .851 .851 .851 .864 .851 .864	(18%) (28 AUC F AUC AUC F AUC .843 .820 .820 .843 .820 .820 .843 .820 .820 .843 .790 .826 .863 .790 .790 .862 .790 .790 .862 .790 .704 .845 .818 .818 .851 .826 .826 .864 .826 .848 .864 .848 .848 .864 .848 .848 .881 .874 .871	(18%) $(28%)$ AUCFAUCFAUCFAUCF.842.820.820.843.820.826.863.790.790.862.790.704.862.704799.704845.818851.826864.848864.848864.848879.871.	(18%) $(28%)$ (0.4) AUCFAUCAUCFAUC AUC FAUC.843.820.775.849.826.774.863.790.774.862.790.774.862.790.774.862.790.774.862.704.762.863.818.766.851.826.776.864.826.776.864.848.653.881.874.743.879.871.794	(18%) $(28%)$ $(0.4%)$ AUCFAUCFAUCFAUCF AUC FAUCF 843 .820.775.775.843.826.774.774.863.790.774.774.862.790.774.714.862.790.774.714.862.790.774.714.862.790.774.714.862.790.774.714.862.818.762.114.845.818.766.114.864.826.776.114.864.848.653.114.881.874.743.143.879.871.794.794	$(18 \times)$ $(28 \times)$ $(0.4 \times)$ (0.1) AUCFAUCFAUCIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII	(18%)(28%)($0.+\%$)($0.1-5\%$)AUCFAUCFAUCFAUCFAUCFAUCF.843.820.775.923.923.849.826.774863.790.774862.790.774862.790.774862.790.774862.790862862862862862862862862863864864881881879879	(18 \times) (28 \times) (0.4 \times) (0.1 \times) (0.1 \times) AUC F AUC F AUC F AUC F AUC F AUC AUC F AUC	(1 \mathbb{B}) (2 \mathbb{B}) (0. \mathbb{H}) (0. \mathbb{H}) (0. \mathbb{H}) (0. \mathbb{H}) AUC F AUC </td <td>(18\times) (0.4\times) (0.1\times) <th< td=""></th<></td>	(18 \times) (0.4 \times) (0.1 \times) <th< td=""></th<>	





False Positive Rate

Fisheries Observer Data

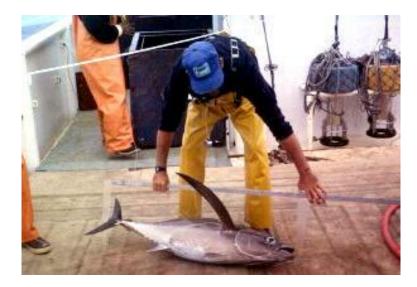
West Coast Groundfish

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



Hawaii Longline

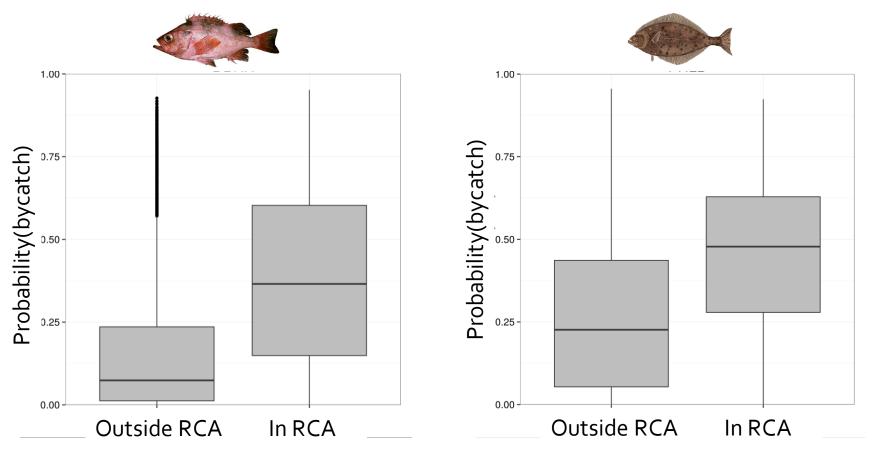
- **1994-2014**
- 70,297 sets
- 3.2 million records



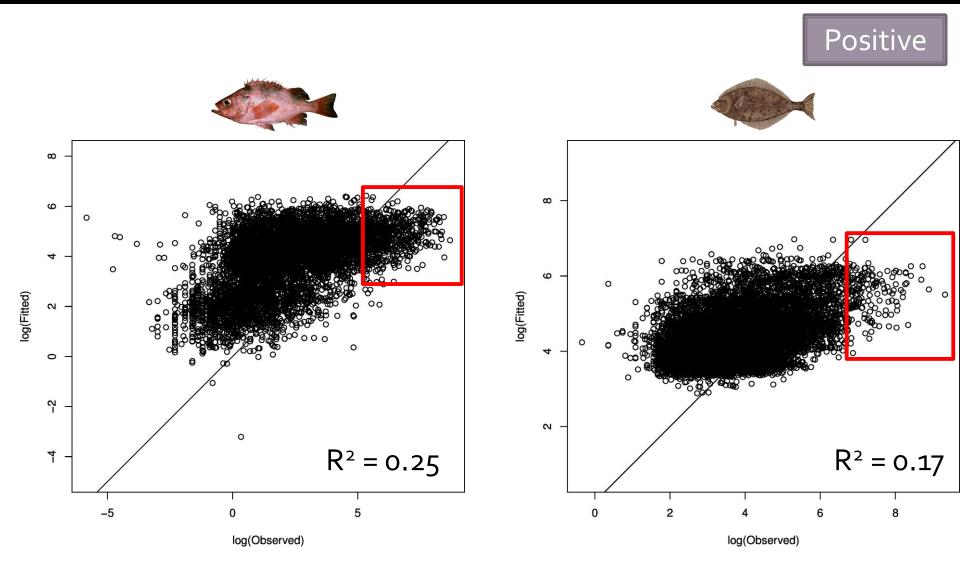
Results: RCAs



11% of tows were in Rockfish Conservation Areas



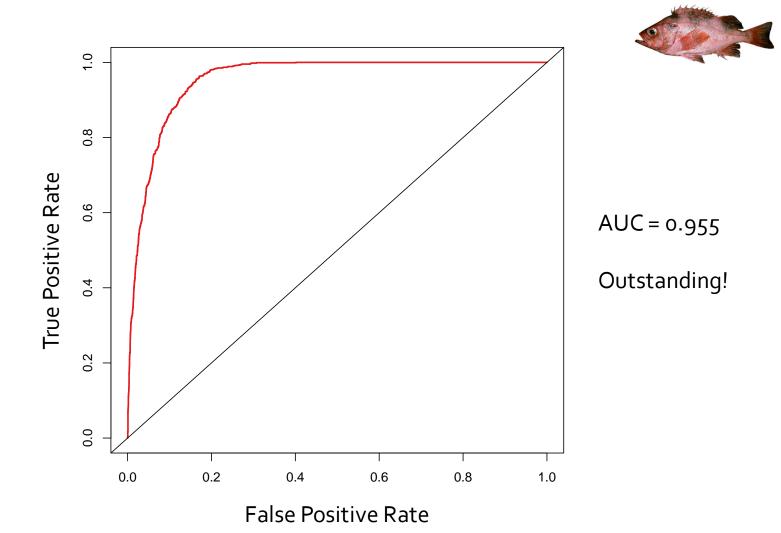
Q: What about the positive model?



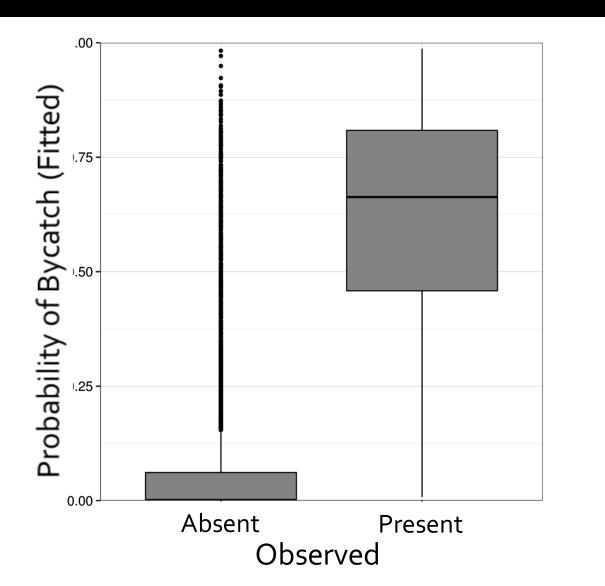
Q: What about effort?

Results: ROC (survey)

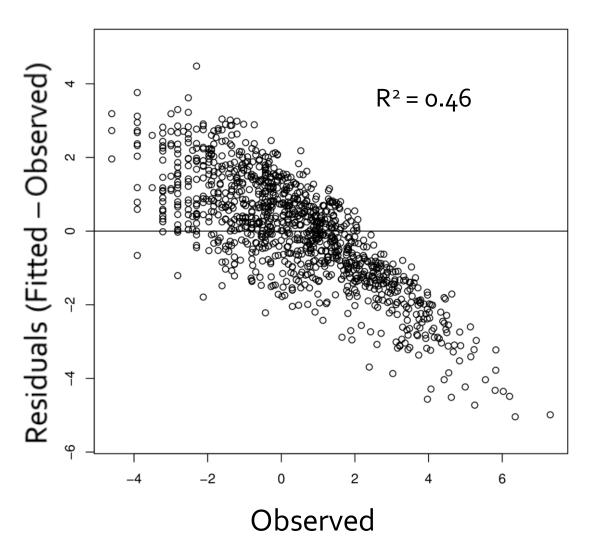




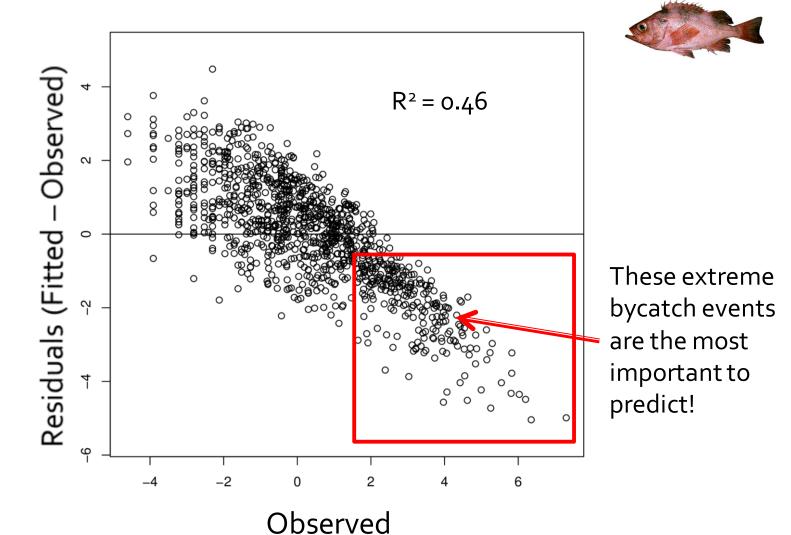






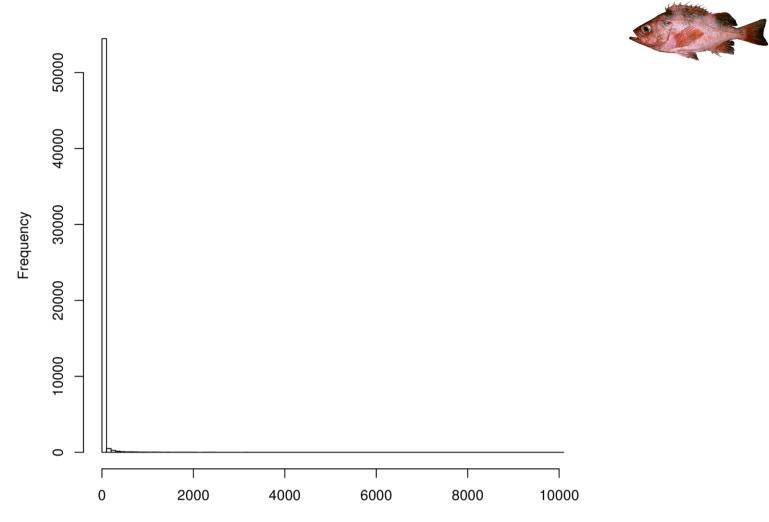


Positive



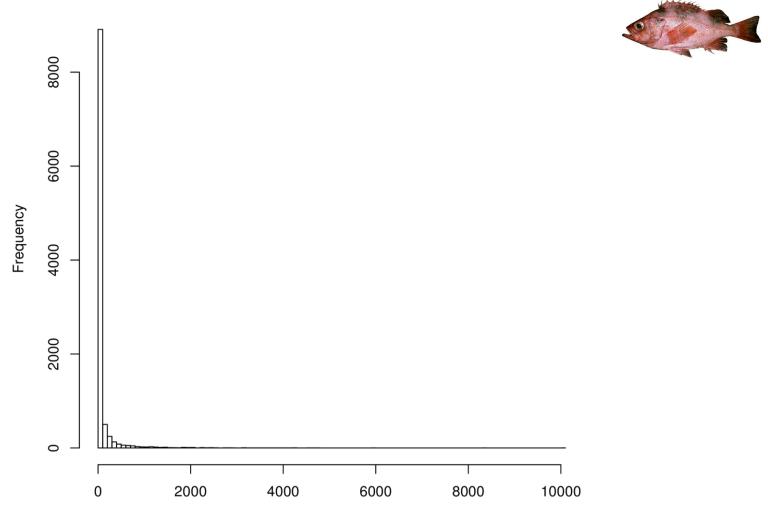
Positive





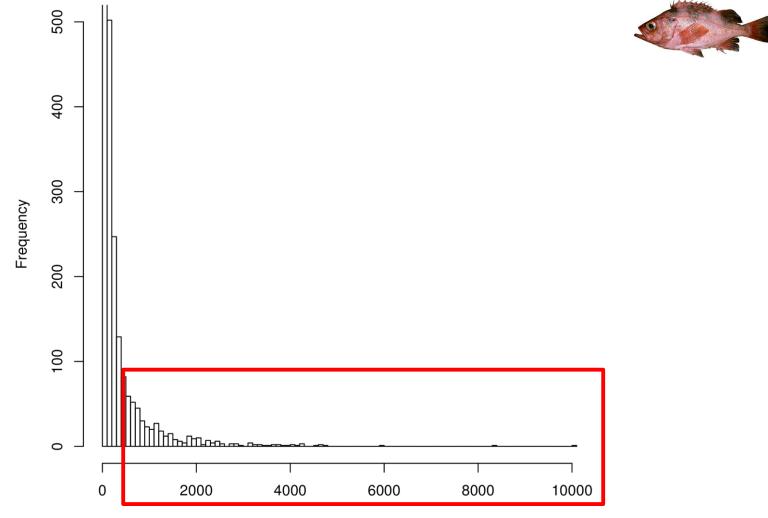
dat\$DBRK

Positive



DBRKpos

Positive



DBRKpos

Spatial models: INLA-SPDE

Integrated Nested Laplace Approximation

Alternative to MCMC for Bayesian inference

0.2

0.4

0.6

0.8

1.0

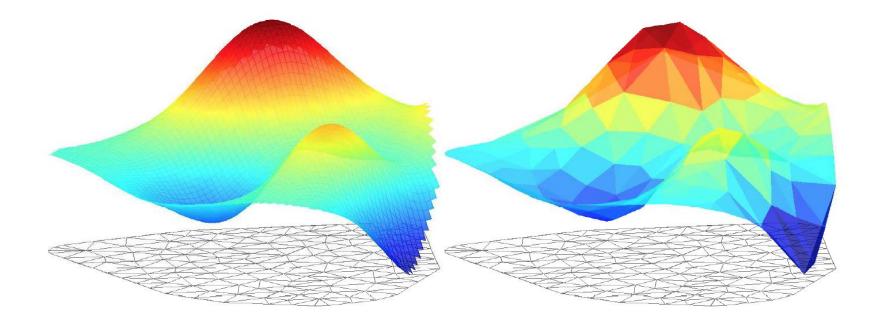
0.0

- Much faster
- 1. Find the posterior mode
- 2. Calculate local curvature
- 3. Use *N*(mode, curvature)

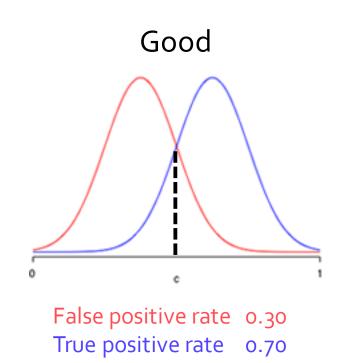
Spatial models: INLA-SPDE

Spatial Partial Differential Equation

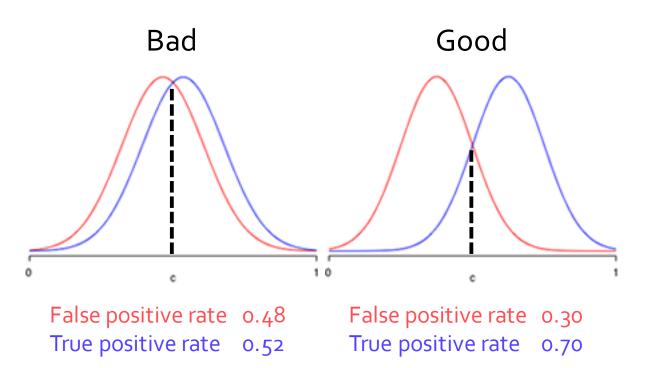
Discrete approximation of continuous spatial fields



Preliminary results: ROC curves



Preliminary results: ROC curves



Preliminary results: ROC curves

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

