

Characterizing Potential Distributions of Deep-Sea Corals and Sponges Offshore the US West Coast through Spatial Predictive Modeling

Matthew Poti^{1,2}, Laurie Bauer^{1,2}, Joseph J. Bizzarro^{3,4}, M. Elizabeth Clarke⁵, Michael Coyne^{1,2}, Meredith V. Everett⁶, Lisa Gilbane⁷, Thomas F. Hourigan⁸, Thomas E. Laidig³, Abigail Powell⁶, Curt E. Whitmire⁵, Arliss Winship^{1,2}, Mary M. Yoklavich³

- 1. CSS, Inc., USA
- 2. NOAA, NOS, National Centers for Coastal Ocean Science (NCCOS), USA
- 3. NOAA, NMFS, Southwest Fisheries Science Center, USA
- 4 University of California, Santa Cruz, USA
- 5. NOAA, NMFS, Northwest Fisheries Science Center, USA
- 6. Lynker Technologies under contract to NOAA, NMFS, Northwest Fisheries Science Center, USA
- 7. BOEM, Pacific OCS Region, USA
- 8. NOAA, NMFS, Deep Sea Coral Research and Technology Program, USA

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Background

Pacific OCS Region:

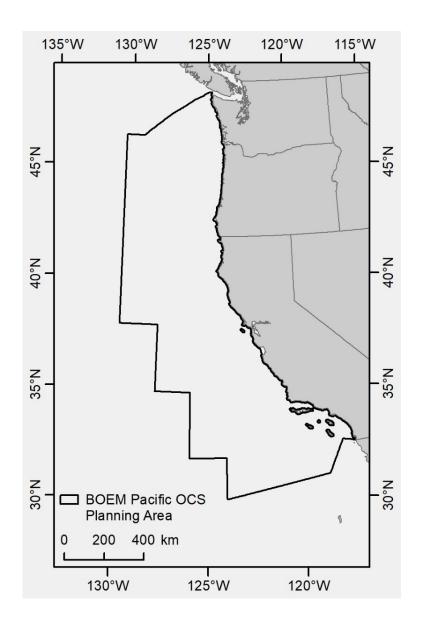
 BOEM oversees responsible development of energy and mineral resources for an extensive area offshore
 California, Oregon, Washington



Credit: Sarah Henkel, Oregon State University



Credit: NOAA SWFSC, Advanced Survey Technologies Group





Objectives

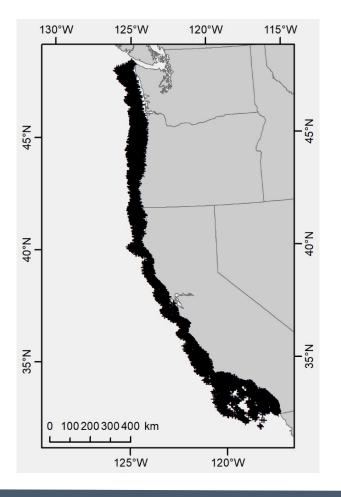
- Compile observations of deep-sea corals and sponges (DSC&S)
- Identify potential environmental covariates
- Predict and map spatial patterns of habitat suitability
- Evaluate model performance
- Support management and exploration priorities



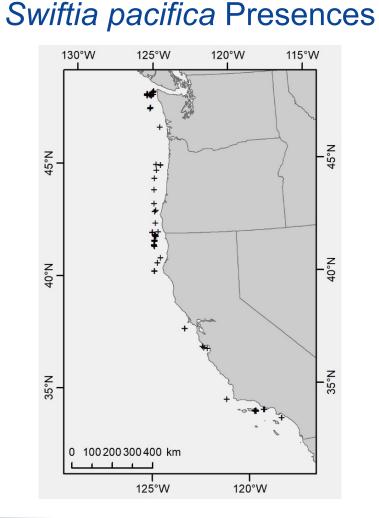
Occurrence Data

NOAA DSCRTP National Database

DSC&S Presences



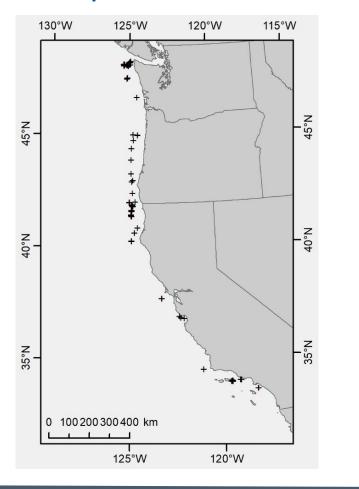
QA/QC Subset by Taxonomy Spatial Thinning



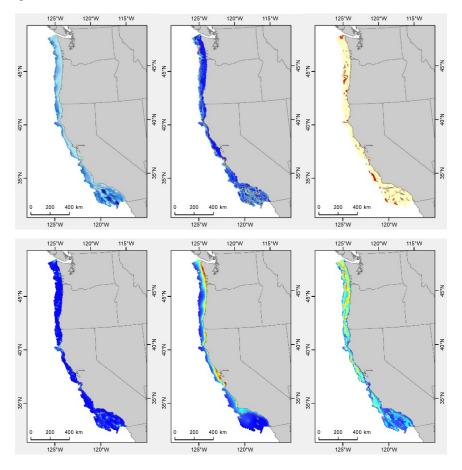


Step 1: Data preparation

Swiftia pacifica Presences



Spatial Environmental Predictors







Step 2: Model fitting

- Models fit using 'maxnet' package in R
- Presence/background data



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Opening the black box: an open-source release of Maxent

Steven J. Phillips, Robert P. Anderson, Miroslav Dudík, Robert E. Schapire and Mary E. Blair

S. J. Phillips (http://orcid.org/0000-0002-6991-608X) (mrmaxent@gmail.com) and M. E. Blair, Center for Biodiversity and Conservation, American Museum of Natural History, New York, NY, USA. – R. P. Anderson, Dept of Biology, City College of New York, City Univ. of New York, New York, NY, USA, and Program in Biology, Graduate Center, City Univ. of New York, New York, NY, USA, and Div. of Vertebrate Zoology (Mammalogy), American Museum of Natural History, New York, NY, USA. – M. Dudík and R. E. Schapire, Microsoft Research, New York, NY, USA.

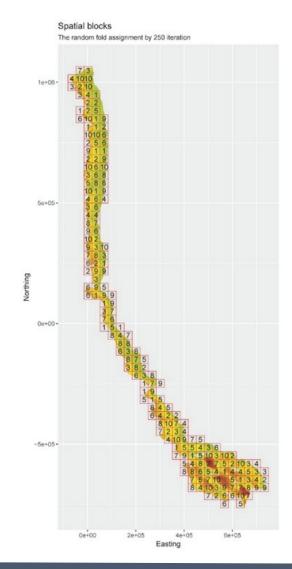


Step 2: Model fitting (continued)

Cross-validation using spatial blocking

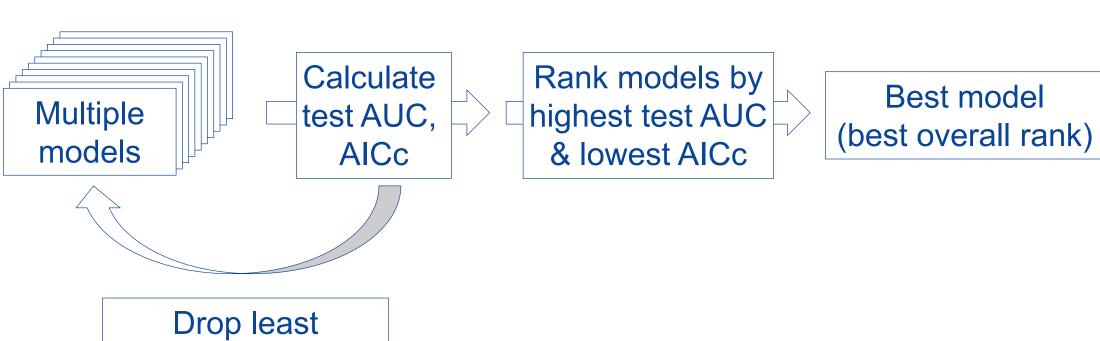


BLOCKCV: An R package for generating spatially or environmentally separated folds for k-fold cross-validation of species distribution models





Step 3: Model selection



Drop least important predictor, repeat model fitting

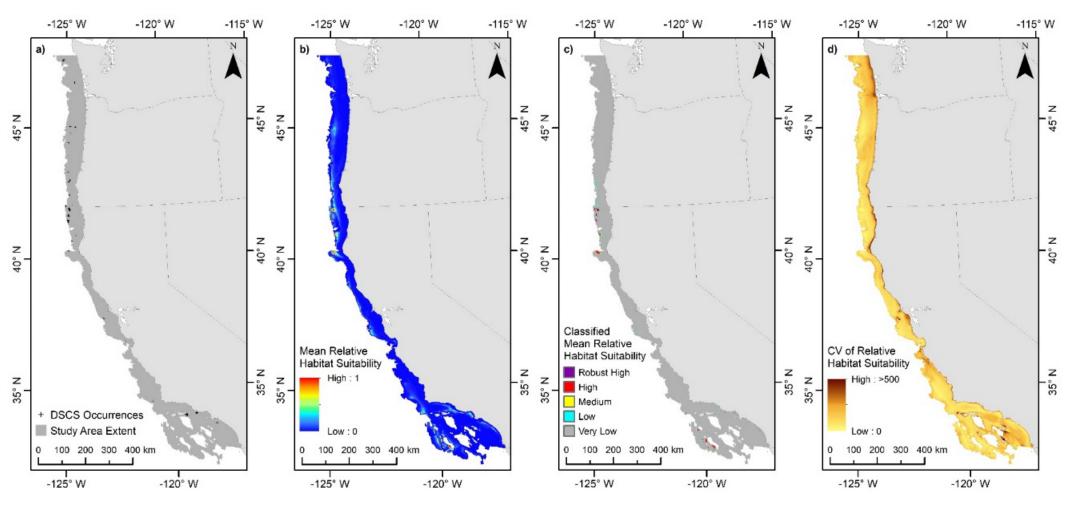


Step 4: Spatial prediction

- Create bootstrap samples
- Fit model for each bootstrap sample, using the predictors from the selected "best" model
- Make predictions at all model grid cells



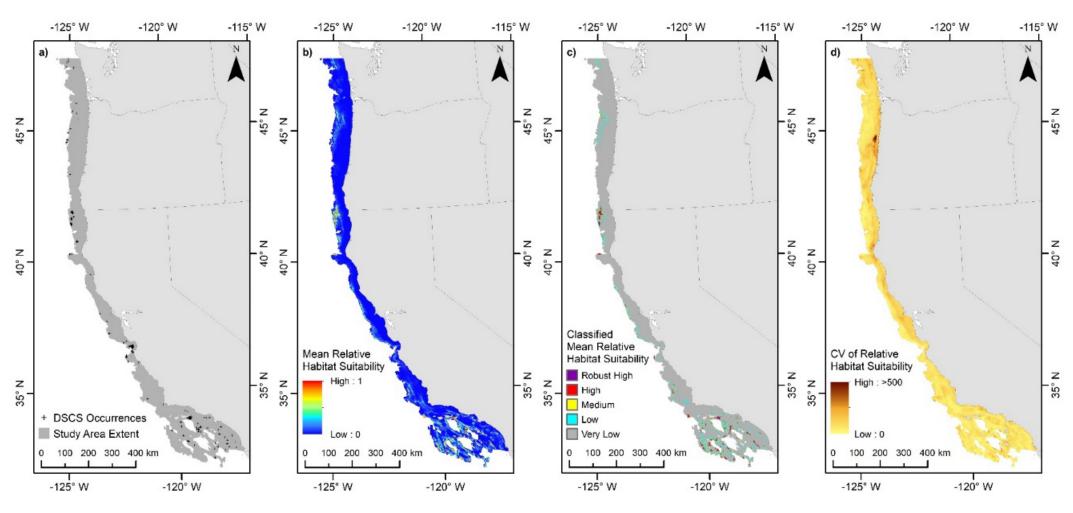
Swiftia pacifica



Cross-Val Mean AUC: 0.83; Model Fit: 85%; Model Stability: 16%



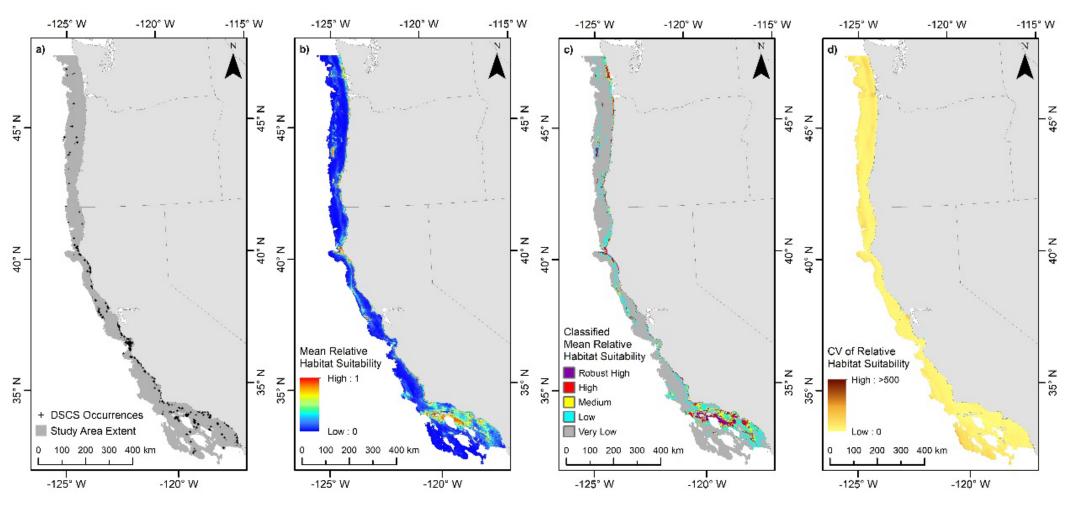
Paragorgia spp.



Cross-Val Mean AUC: 0.87; Model Fit: 91%; Model Stability: 23%



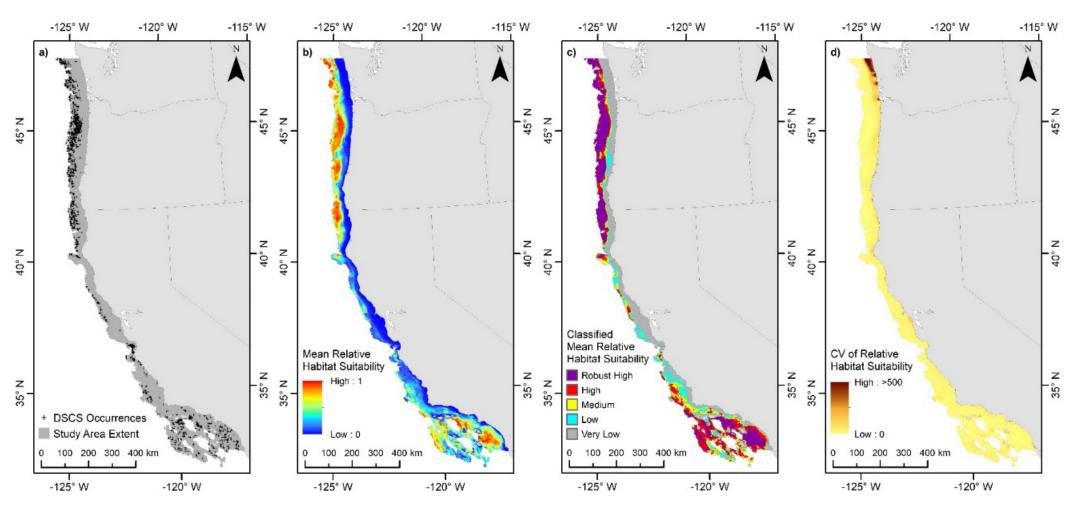
Demospongiae



Cross-Val Mean AUC: 0.81; Model Fit: 90%; Model Stability: 43%



Hexactinellida



Cross-Val Mean AUC: 0.82; Model Fit: 86%; Model Stability: 61%





- Presence-only data
- Spatial and taxonomic precision of DSC&S records
- Scale/resolution of environmental predictors
- Missing environmental predictors



Objectives

- Compile observations of deep-sea corals and sponges (DSC&S)
- Identify potential environmental covariates
- Predict and map spatial patterns of habitat suitability
- Evaluate model performance
- Support management and exploration priorities

'Opportunistic' field validation using data from EXPRESS

Model Validation



Ideal – Independent Field Validation

Ocean & Coastal Management 120 (2016) 110-126



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journal homepage: www.elsevier.com/locate/ocecoaman



Field validation of habitat suitability model ecosystems in the South Pacific Ocean: Imp broad-scale models in fisheries managemen

Owen F. Anderson ^{a, *}, John M. Guinotte ^b, Ashley A. Ro Sophie Mormede ^a, Andrew J. Davies ^c, David A. Bowde

ICES Journal of Marine Science (2018), 75(1), 199-209. doi:10.1093/icesjms/fsx087

Original Article

Validation of deep-sea coral and sponge distribution models in the Aleutian Islands, Alaska

Christopher N. Rooper, ** Rachel Wilborn, ** Pamela Goddard, ** Kresimir Williams, ** Richard Towler, ** and Gerald R. Hoff**

a NIWA, Wellington, New Zealand

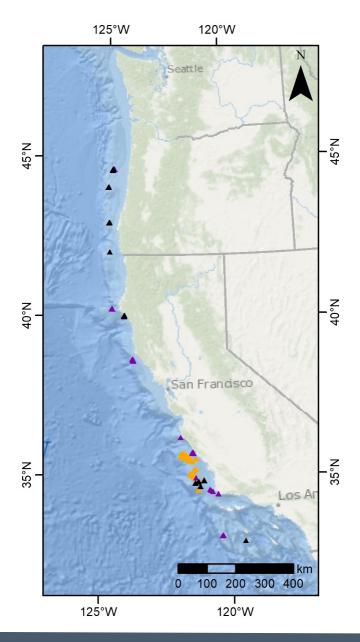
Marine Conservation Institute, Seattle, USA

School of Ocean Sciences, Bangor University, Menai Bridge, Wales, LL59 5AB, UK

¹Resource Assessment and Conservation Engineering Division, Alaska Fisheries Science Center, National Marine Fisheries Service, NOAA, 7600 Sand Point Way NE, Seattle, WA 98115, USA







- 2018 NOAA Ship Bell Shimada
 - ▲ AUV dives
 - ▲ ROV dives
- 2019 MBARI
 - ROV dives



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Backside Heceta 2												X	X						X					X							Х								low relief area, but not completely mapped with multibeam
Brandon High Spot/Coquille 1										X							Χ		X		X	(Χ								
Brandon High Spot/Coquille 2							X			X				X			Х		X	(X	(X		X								transect moves up and then along high slope (>30° feature
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N. Daisy Bank 2				Χ	X					X	Χ		X						X	X	X X	(X				Χ								
N. Daisy Bank 3																					X	(1	X												transect moves up and then along low-moderate slope (>15°) feature
Brush Patch										X	>	< x					Х			X	x x	(x >	(X							Х				only small area shallower than 600m; transect moves along edge of high slope (15°-45°) feature
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Delgada Canyon 2	X						X													X	x x	(Х	X				X									transect moves up and along consecutive high slop (>45°) features
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Approaches

- Assess correlation between observations and model predictions
- Calculate accuracy measure using confusion matrix
- Fit GLMs to see if model predictions explain variation in observations

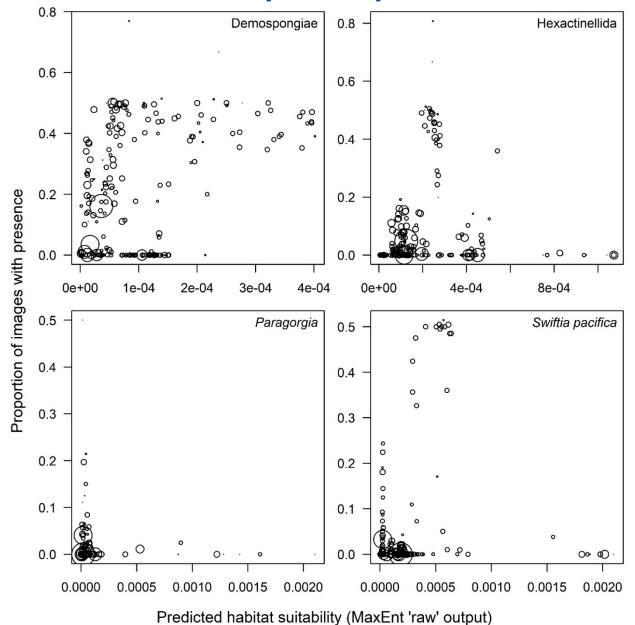




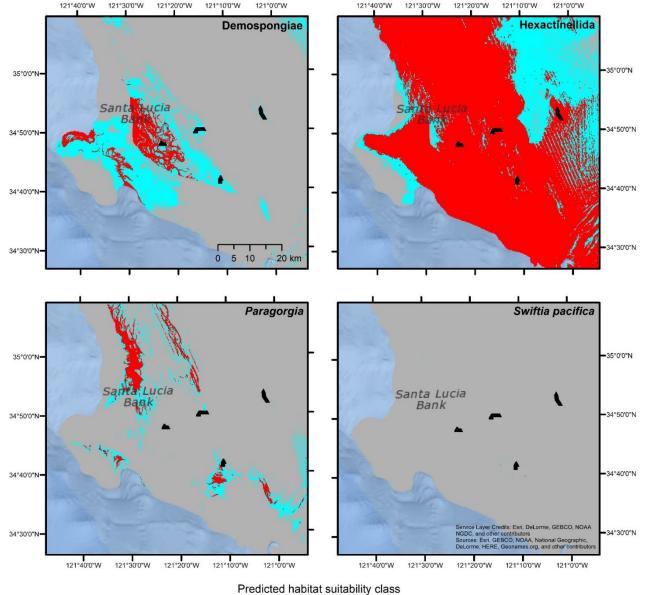


Table 1. Sample size (number of study grid cells with ≥1 image) by predicted habitat suitability class for each taxon. There were a total of 220 cells with images.

					Predicted hab	itat suitability class						
Taxon		Very	low		Low	Medium	High					
1.5	1	2	3	4	5	6	7	8	9	10		
Demospongiae	0	0	12	20	38	81	44	25	0	0		
Hexactinellida	0	0	8	9	20	6	56	79	33	9		
Paragorgia	89	15	69	14	23	1	2	5	2	0		
Swiftia pacifica	24	49	52	69	17	0	9	0	0	0		

Very Low





Medium





Table 2. Results of statistical analyses of the relationship between taxa occurrence and predicted habitat suitability (MaxEnt 'raw' predictions): 1) Spearman rank correlation coefficient (*r*) between proportion of images where taxa were present and predicted habitat suitability; and 2) quasi-Poisson generalized linear model (GLM) of number of images where taxa was present as function of predicted habitat suitability. For the GLM, the percentage of deviance explained by the model and the *p*-value of the positive effect of predicted habitat suitability are presented.

Taxon	R	GLM							
		% deviance explained	р						
Demospongiae	0.24	9	<1e-7						
Hexactinellida	0.27	1	0.135						
Paragorgia	0.44	0	0.894						
Swiftia pacifica	0.14	5	< 0.01						



Conclusions

- Challenging to collect samples across range of model predictions when using 'opportunistic' samples
- When incorporating different sources of data (e.g., AUV + ROV), need to consider sampling effort
- Important to link modeling efforts with exploration



Next steps

- Continue opportunistic collection of data for model validation
- Additional models using absence, abundance data
- Explore additional environmental predictor variables

Acknowledgments

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- Robert McGuinn (NOAA NCEI)



Questions?

contact: Matthew Poti, matthew.poti@noaa.gov