

Marine Geospatial Ecology Tools

Open Source Geoprocessing for Marine Ecology

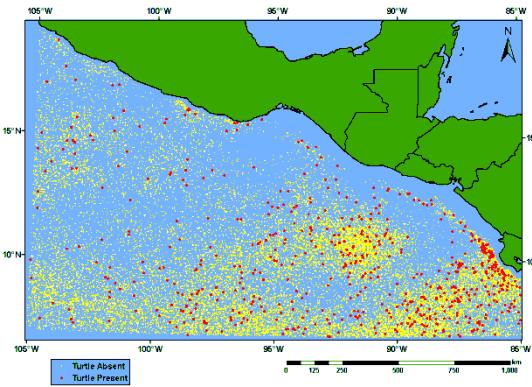
Elliott Hazen, Jason Roberts, Ben Best, Dan Dunn, Pat Halpin
Duke University Marine Geospatial Ecology Lab
Aug 8th, 2010





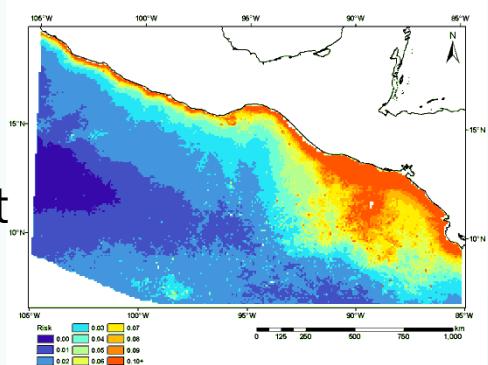
Modeling habitat (overview)

Distribution data, e.g. presence/absence

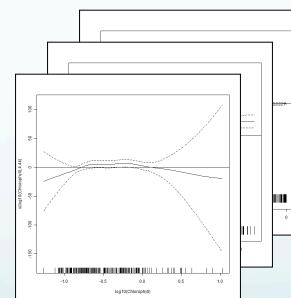
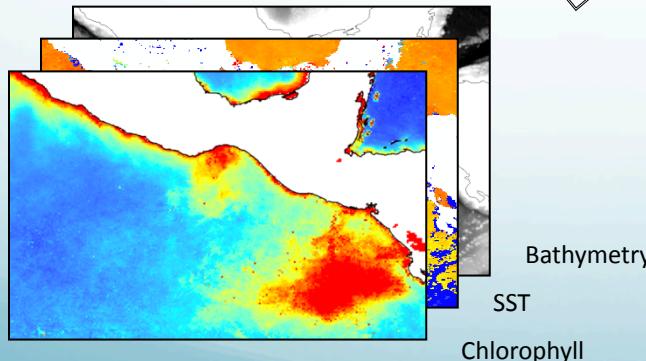


Statistical models

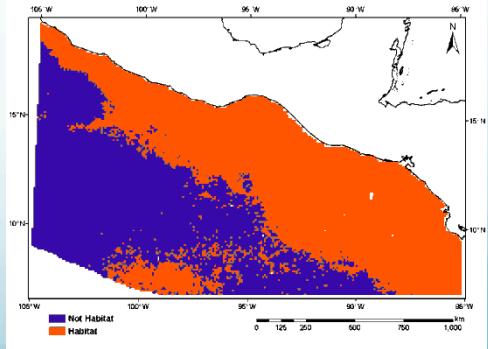
$$g(\mu) = \beta_0 + \beta_1 x_1 + \cdots + \beta_m x_m$$



Sampled predictive data



Binary classification



Warning: Habitat modeling is complicated! This simplified example is meant to briefly illustrate tools. Consult the literature for best practices!

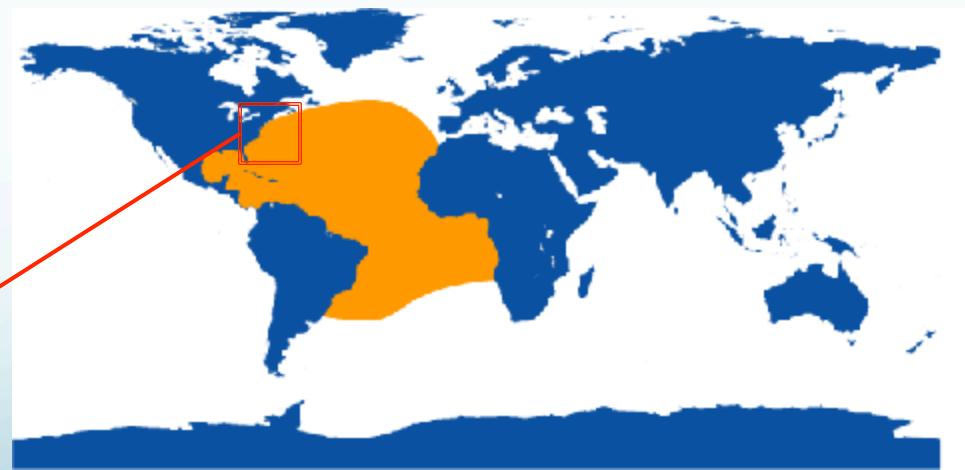
Focal species: *Stenella frontalis*

Common name:
Atlantic Spotted
Dolphin



Photo: Garth Mix

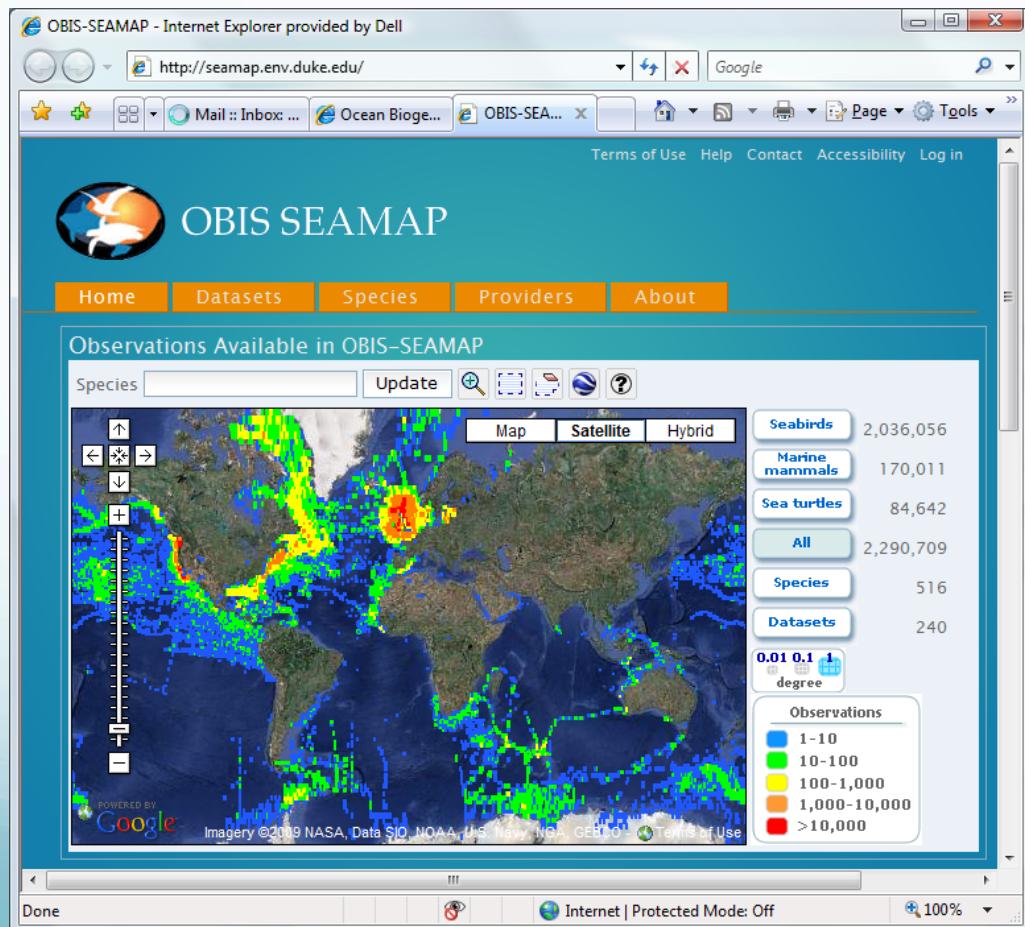
Distribution:
Tropical and warm
temperate Atlantic



Map: OBIS-SEAMAP

Study area:
Eastern
U.S.

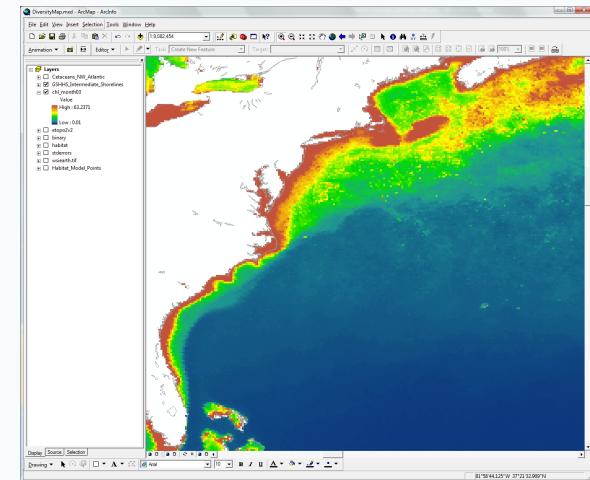
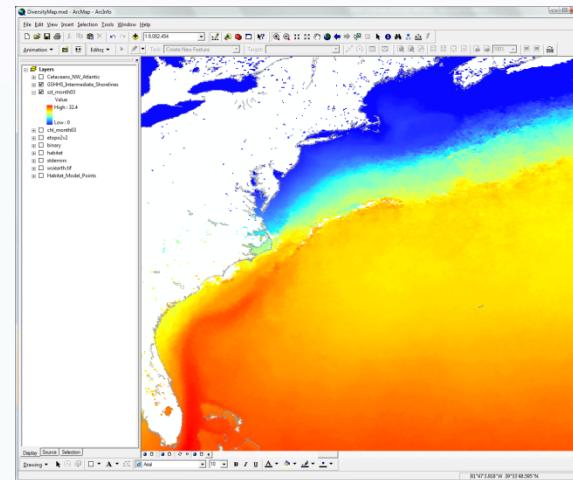
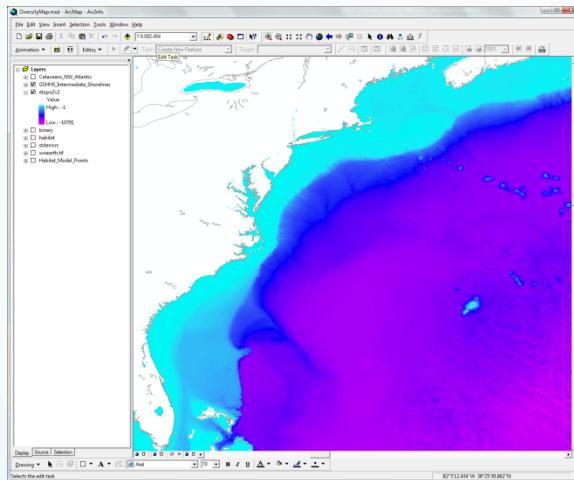
Species observation data



The Ocean Biogeographic Information System (OBIS) is a global database of marine species observations.

The OBIS-SEAMAP system at Duke University holds the records for seabirds, marine mammals, and sea turtles, including records gathered during NOAA cruises.

Environmental predictor variables



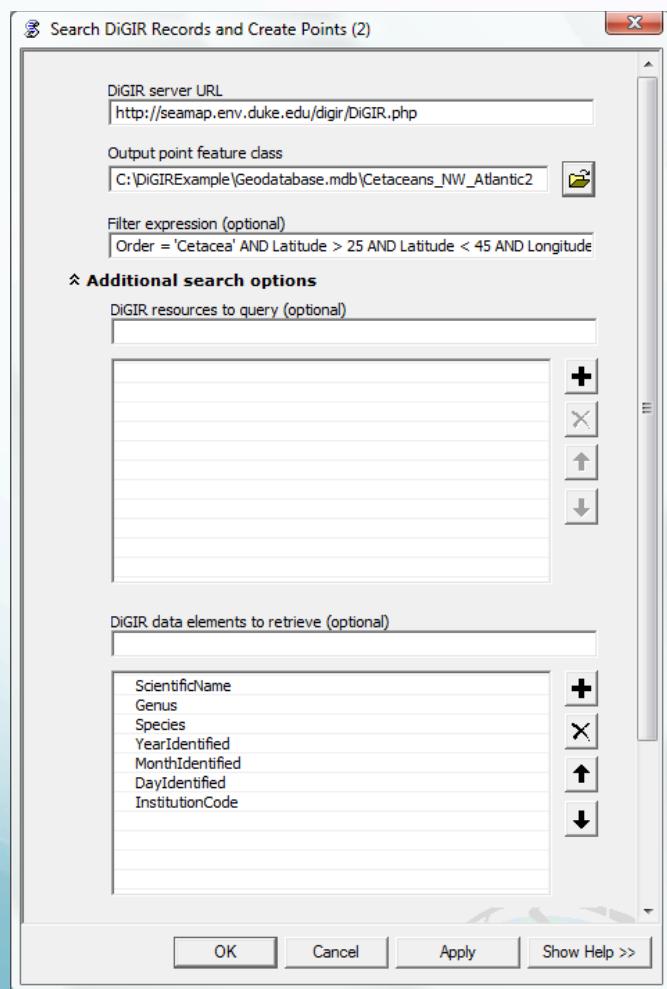
Bathymetry:
ETOPO2V2 from
NOAA NGDC

SST:
Monthly
climatological 4km
AVHRR Pathfinder
from NOAA NODC

Chlorophyll:
Monthly
climatological
SeaWiFS chlorophyll-
a from NASA GSFC

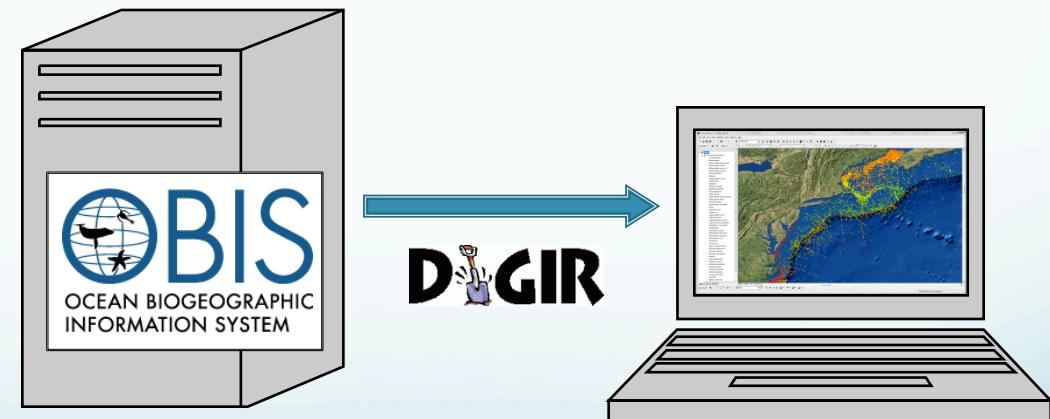
Images shown above are for month of March

Step 1: Download species points

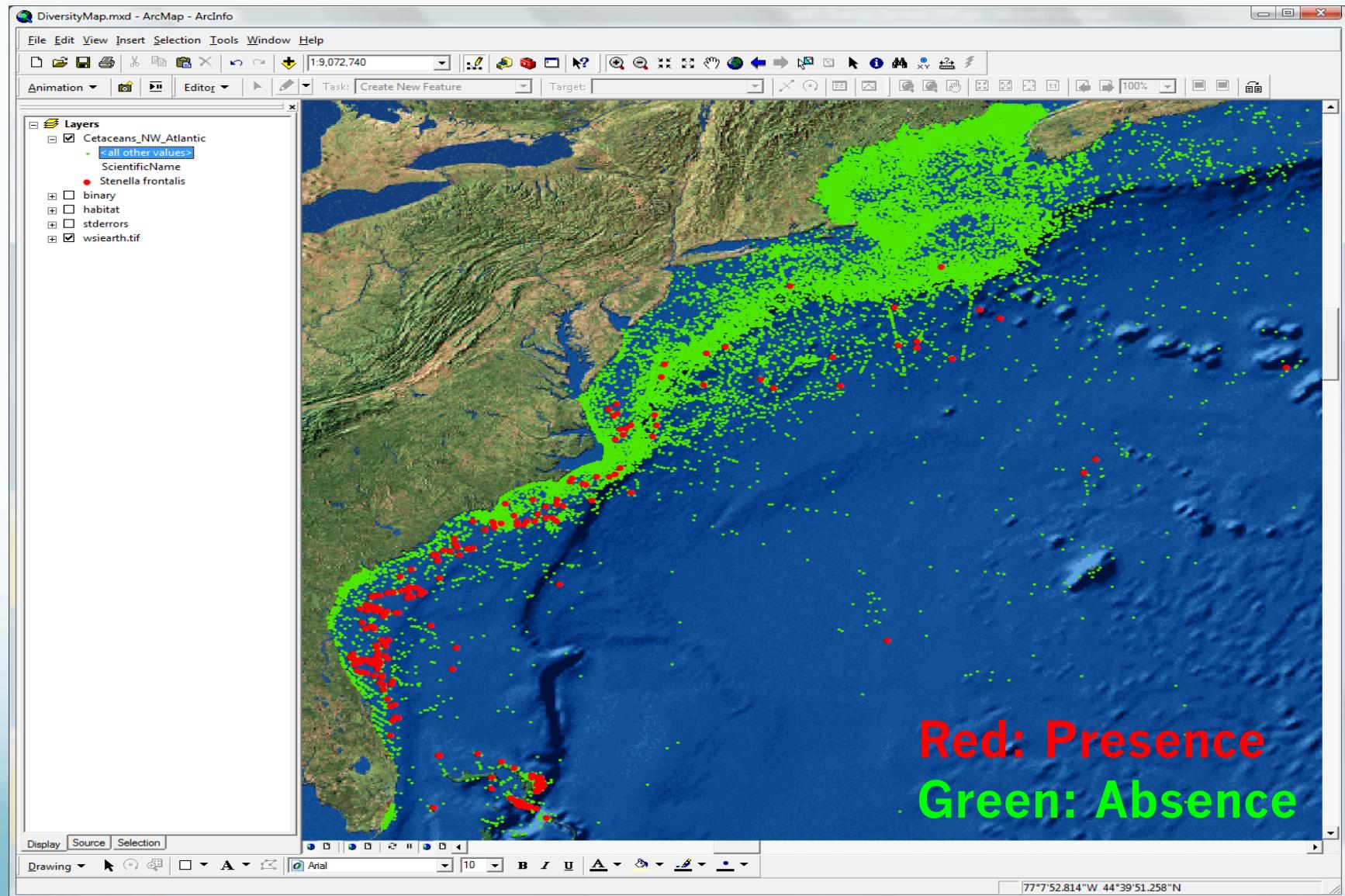


Download points using MGET tool:

- Presence: Records of *Stenella frontalis*
- Absence: Records of other cetaceans



The tool uses the DiGIR protocol to retrieve data from OBIS servers

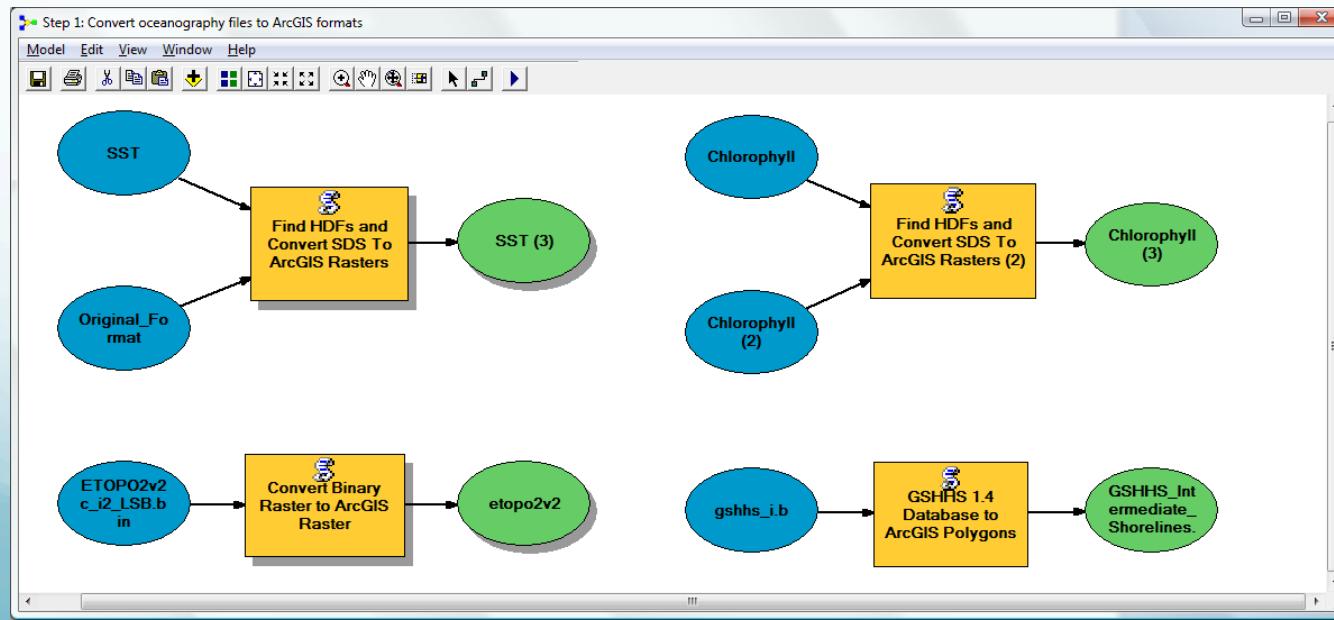


Step 2: Convert oceanography to Arc rasters

1. Download with FTP from NOAA and NASA:

- ETOPO2 bathymetry – 1 binary file
- AVHRR Pathfinder monthly climatological SST – 12 HDF files
- SeaWiFS monthly climatological chlorophyll – 12 HDF files

2. Convert to ArcGIS rasters using MGET tools:



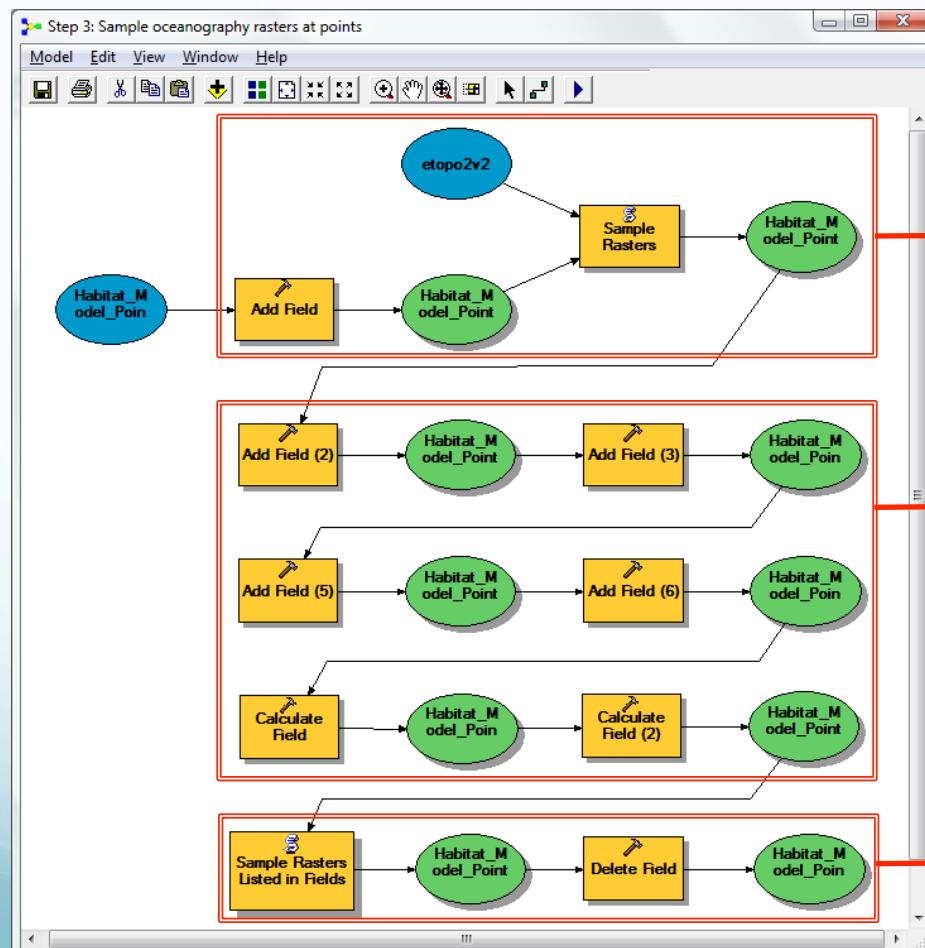
Step 3: Sample oceanography at points

Attributes of Habitat_Model_Points										
OBJECTID *	Shape *	ScientificName	YearIdentified	MonthIdentified	DayIdentified	Presence	Depth	SST	Chlorophyll	
1	Point	Tursiops truncatus	1998	8	18	0				
2	Point	Tursiops truncatus	1998	8	19	0				
3	Point	Physeter catodon	1998	7	10	0				
4	Point	Physeter catodon	1998	7	10	0				
5	Point	Physeter catodon	1998	7	10	0				
6	Point	Balaenoptera acutorostrata	1998	7	10	0				

- Need to sample rasters and populate fields
- Must sample SST and chlorophyll by date

Attributes of Habitat_Model_Points										
OBJECTID *	Shape *	ScientificName	YearIdentified	MonthIdentified	DayIdentified	Presence	Depth	SST	Chlorophyll	
1	Point	Tursiops truncatus	1998	8	18	0	-567	29.700001	0.094755	
2	Point	Tursiops truncatus	1998	8	19	0	-541	29.85	0.096544	
3	Point	Physeter catodon	1998	7	10	0	-3187	26.25	0.209108	
4	Point	Physeter catodon	1998	7	10	0	-3283	26.775	0.187317	
5	Point	Physeter catodon	1998	7	10	0	-3477	27	0.183823	
6	Point	Balaenoptera acutorostrata	1998	7	10	0	-3633	26.924999	0.173819	

Step 3: Sample oceanography at points

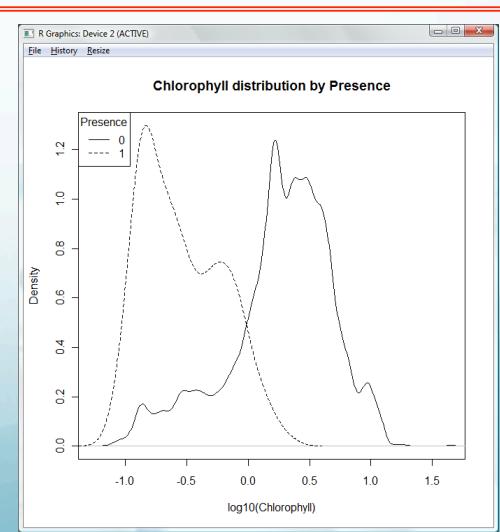
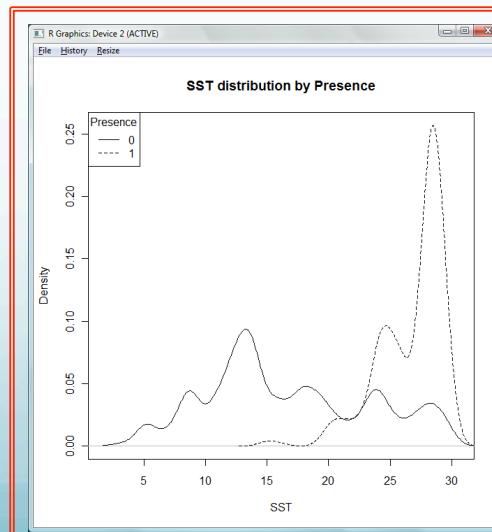
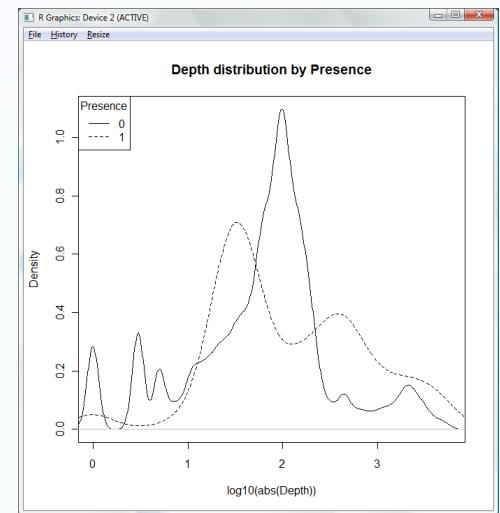
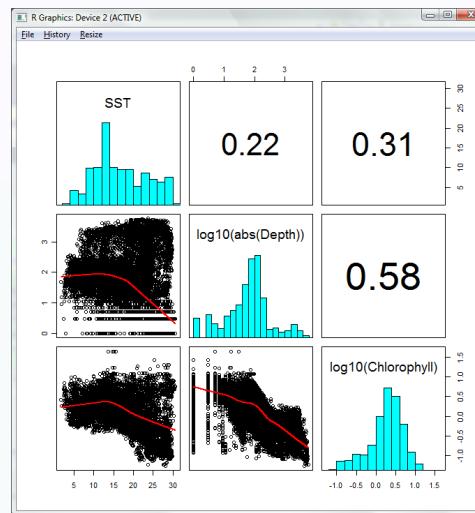
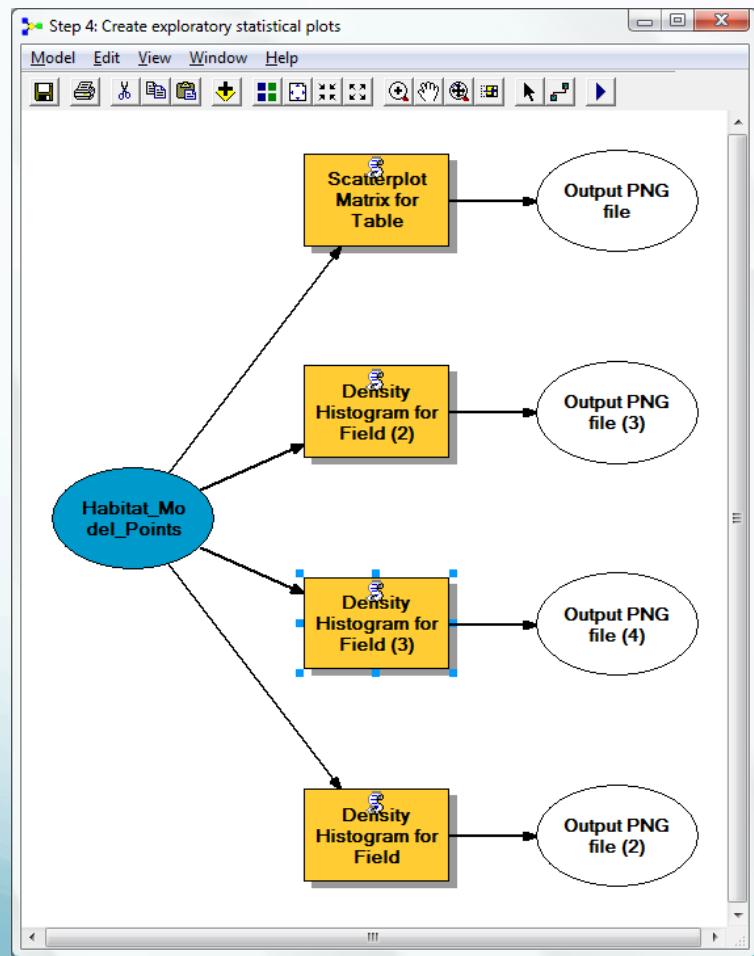


Sampling bathymetry is easy because it is static

To sample dynamic data such as SST and chlorophyll, you must first calculate the paths to rasters to sample from the points' dates

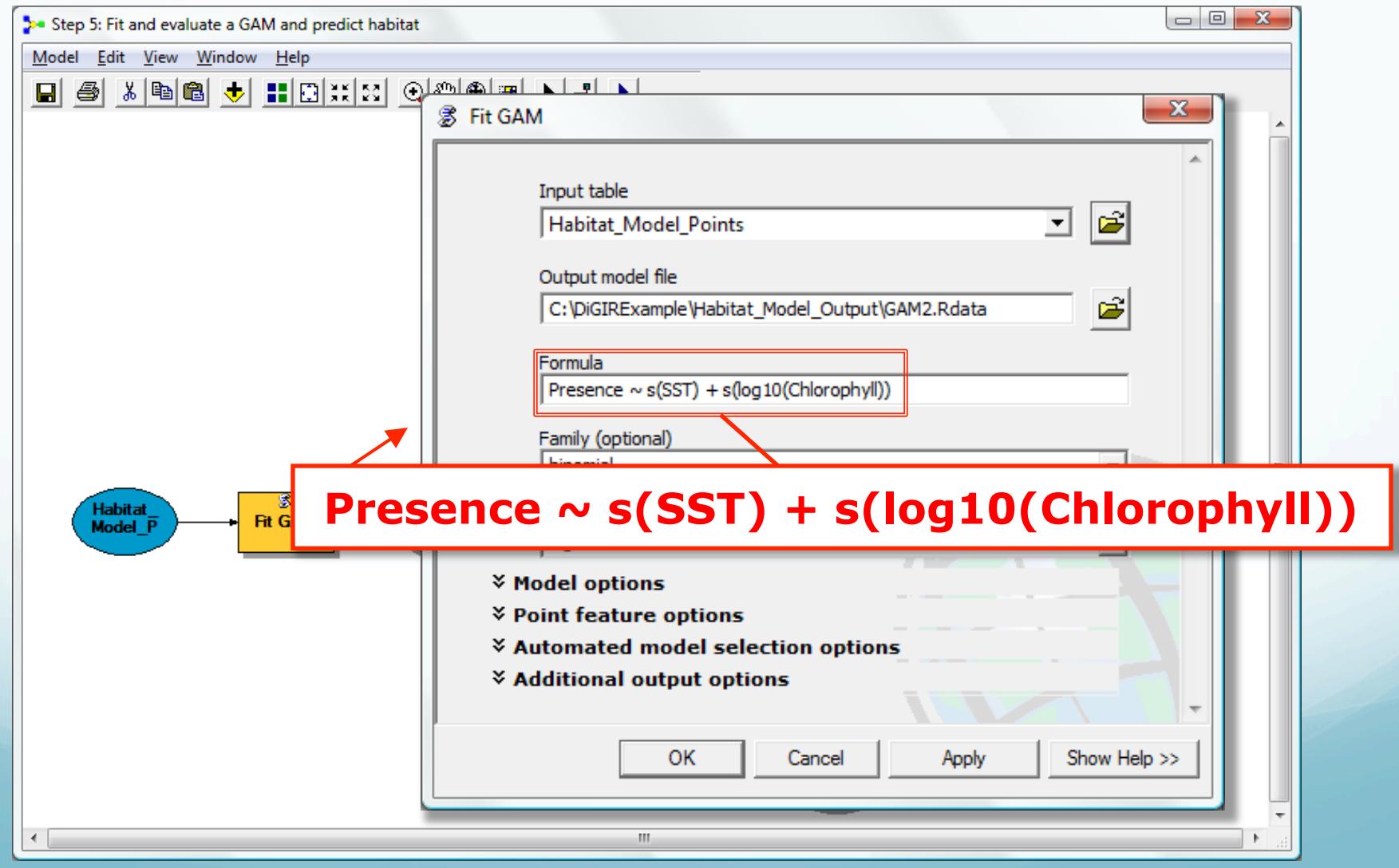
Then use an MGET batch sampling tool

Step 4: Create exploratory plots

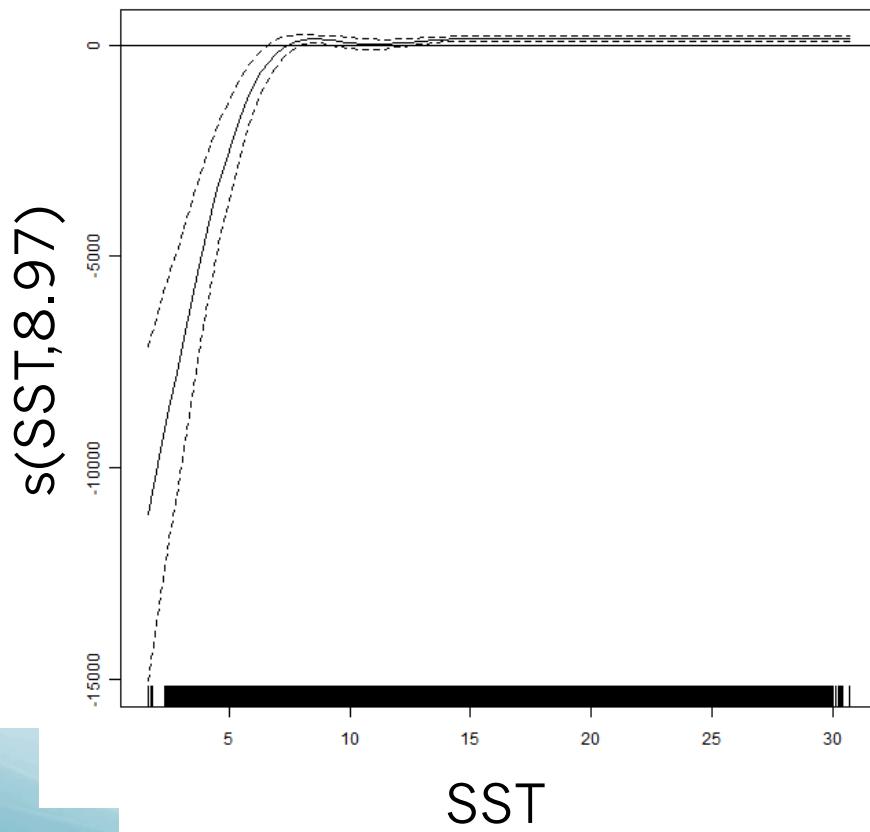


Best predictors: SST and Chl

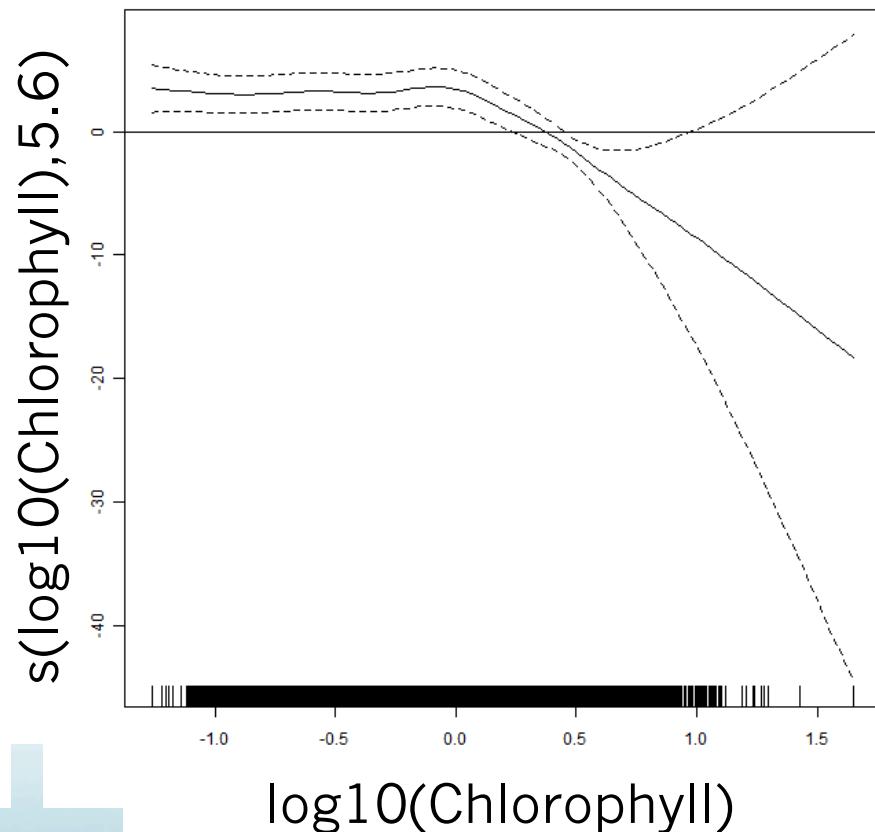
Step 5: Fit, evaluate, and predict model



Partial plots produced by the Fit GAM tool

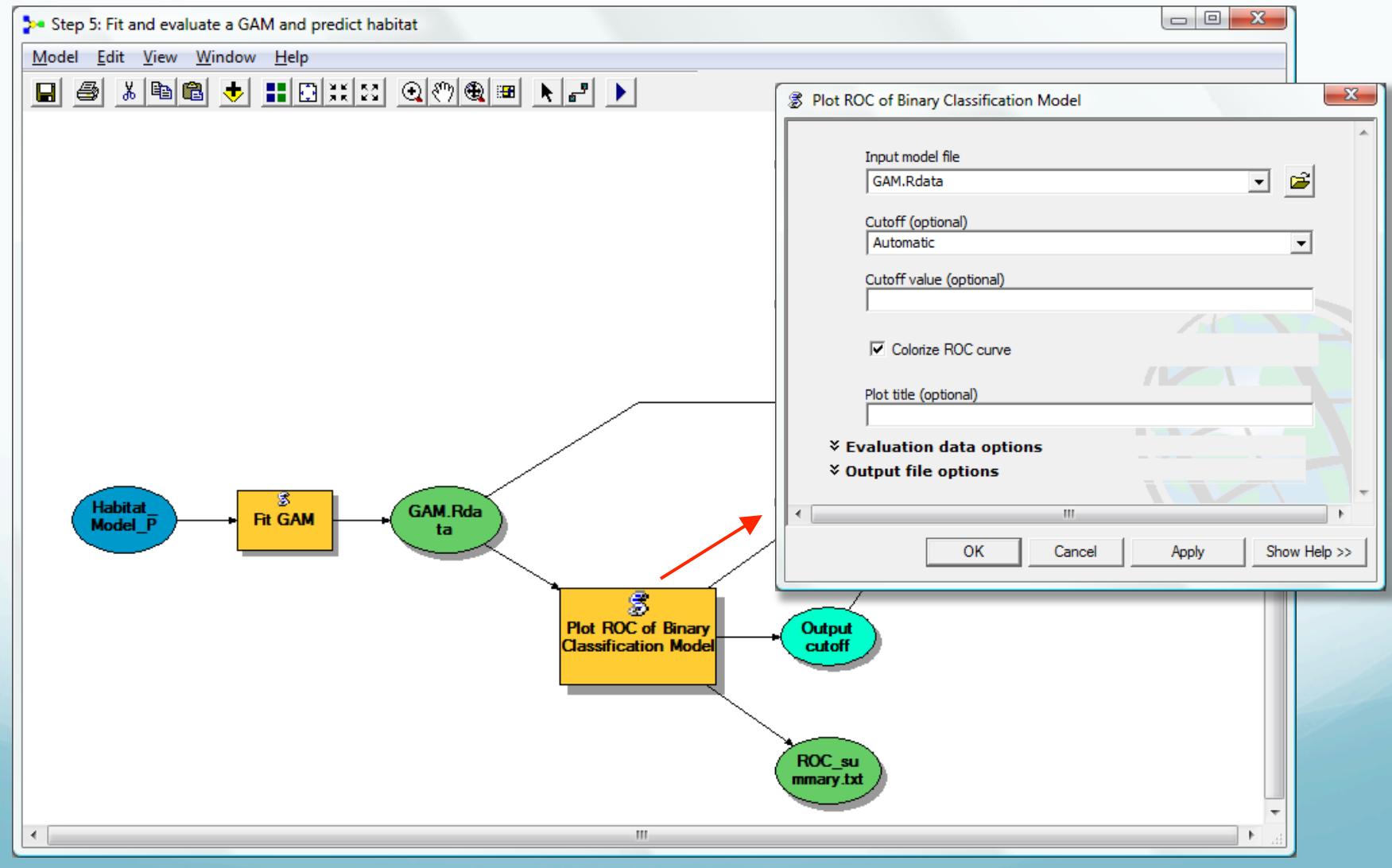


Presence more likely at higher SST

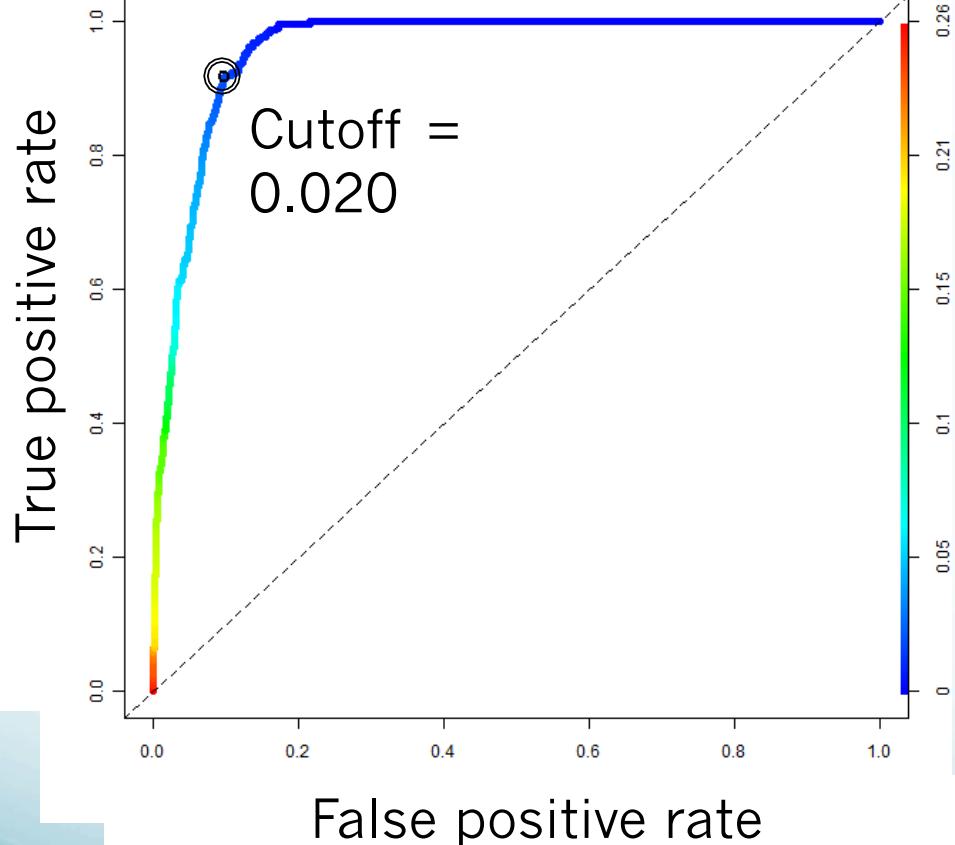


Presence more likely at lower Chl

Plotting a receiver operating characteristic curve



The ROC plot



ROC summary stats for cutoff:

Model summary statistics:

Area under the ROC curve (auc)	= 0.960779
Mean cross-entropy (mxe)	= 0.030566
Precision-recall break-even point (prbe)	= 0.001866
Root-mean square error (rmse)	= 0.087781

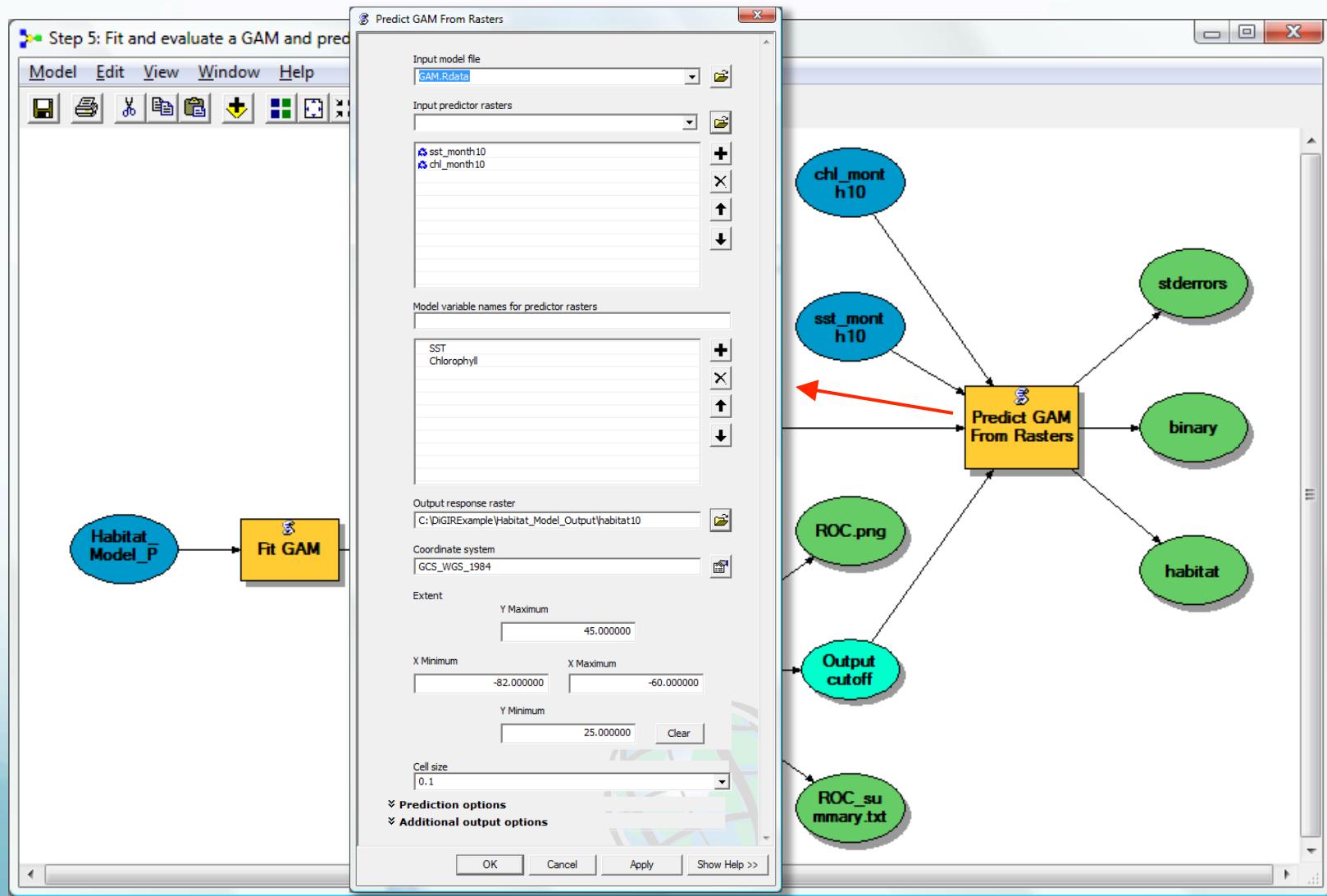
Contingency table for cutoff = 0.019638:

	Actual P	Actual N	Total
Predicted P	287	3541	3828
Predicted N	26	32408	32434
Total	313	35949	36262

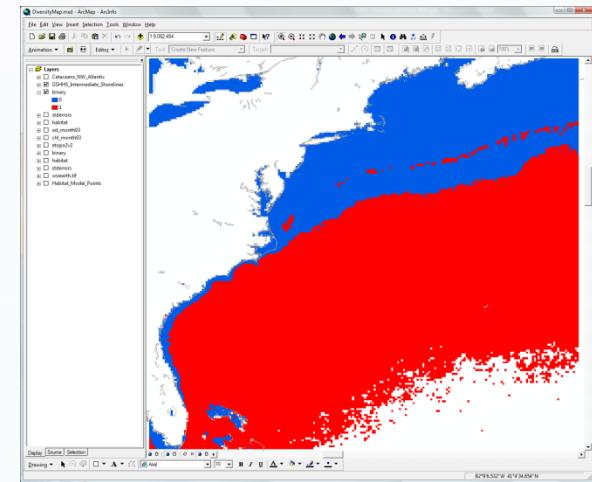
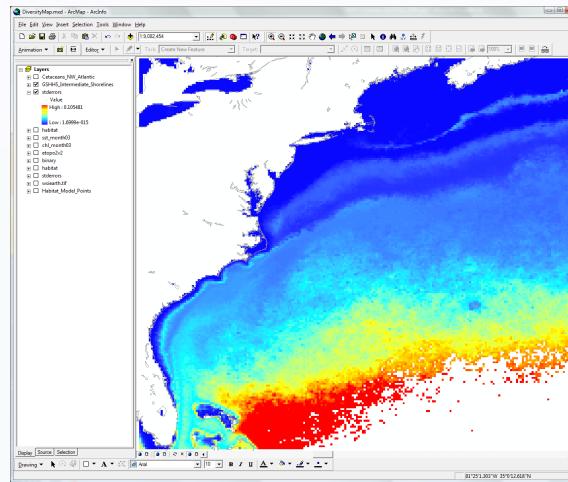
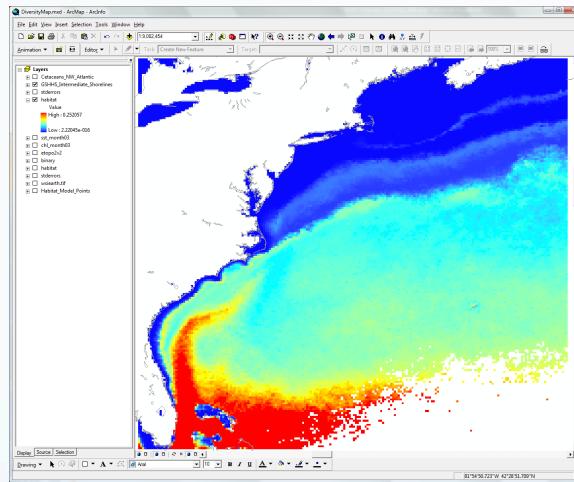
Accuracy (acc)	= 0.901633
Error rate (err)	= 0.098367
Rate of positive predictions (rpp)	= 0.105565
Rate of negative predictions (rnp)	= 0.894435
True positive rate (tpr, or sensitivity)	= 0.916933
False positive rate (fpr, or fallout)	= 0.098501
True negative rate (tnr, or specificity)	= 0.901499
False negative rate (fnr, or miss)	= 0.083067
Positive prediction value (ppv, or precision)	= 0.074974
Negative prediction value (npv)	= 0.999198
Prediction-conditioned fallout (pcfall)	= 0.925026
Prediction-conditioned miss (pcmiss)	= 0.000802
Matthews correlation coefficient (mcc)	= 0.246384
Odds ratio (odds)	= 101.026394
SAR	= 0.650065

By default, tool selects the cutoff closest to the point of perfect classification (0, 1)

Predicting presence for oceanographic rasters



Rasters output by the Predict GAM tool



Predicted presence:

Range: 0 - 0.25

Similar to OBIS-
SEAMAP range map?

Predictions for
October

Standard errors:

Range: 0 - 0.11



**Binary
classification:**

Species range map
produced by
classifying presence
into 0 or 1 according
to ROC cutoff

Acknowledgements

A special thanks to the many developers of the open source software that MGET is built upon! Also, folks that have helped with this talk:

MGET lab, NOAA-SWFSC-ERD, and many, many others



Thanks to our funders:

