

Evaluating the effects of positioning errors on the accuracy of species distribution models using synthetic data



1Crustal Geophysics and Geochemistry Science Center, 2Image Processing Division, National Institute for Space Research—INPE 1{fiwashita, mfriedel}@usgs.gov; 2{miguel, silvana}@dpi.inpe.br

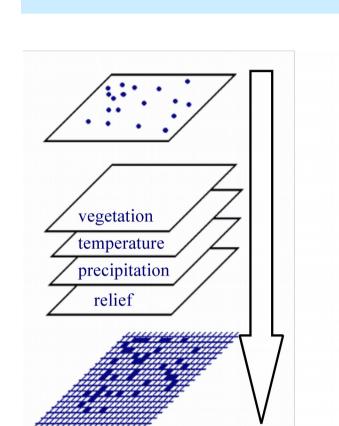


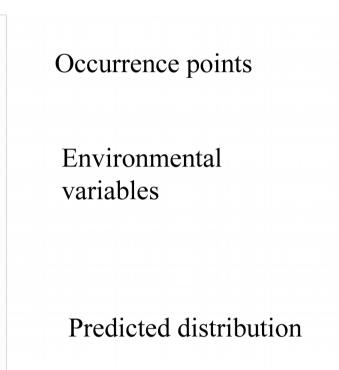


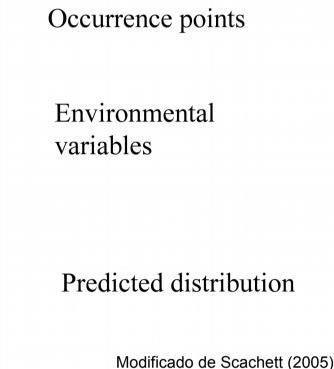
Introduction

Biogeography is the science that attempts to document and understand the spatial patterns of biodiversity, studying the present and the past distributions of organisms.

Species distribution models (SDM) estimates fundamental niche for a certain specie identifying and quantifying relationships between occurrence field data and environmental variables, through statistical, mathematical and data mining methods.





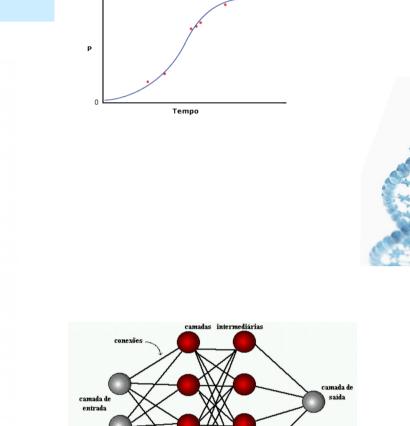


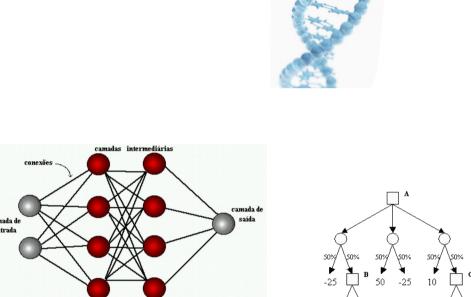
The aim of this study is to evaluate the sensibilities

assessed, bioclimatic envelope, genetic algorithm

of SDM to positioning error using synthetic data.

Three models from different categories were



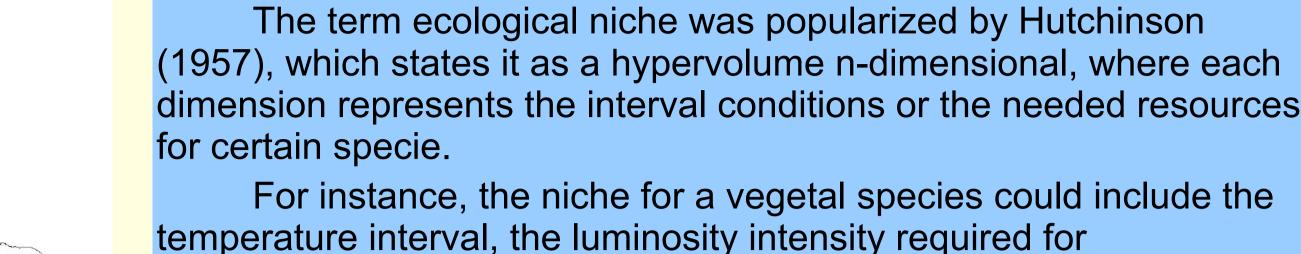


Remote sensing and

the most common

GIS modeling provides

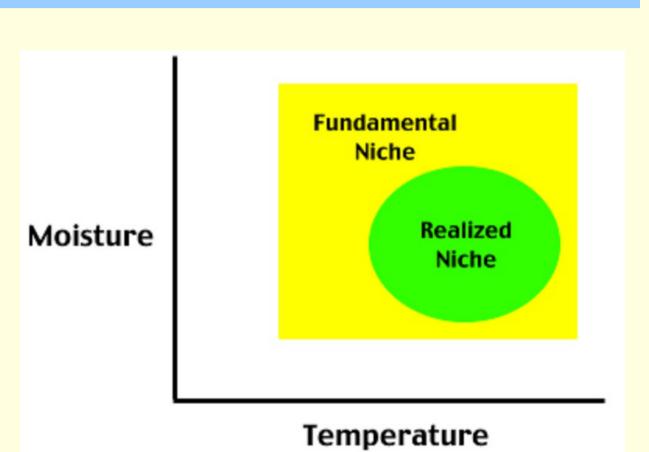
independent variables.



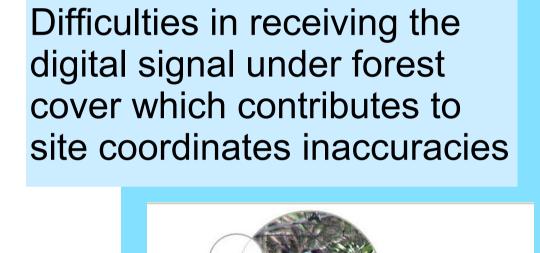
Niche concept

For instance, the niche for a vegetal species could include the temperature interval, the luminosity intensity required for photosynthesis, moisture regime and minimum amount of nutrients in

Fundamental niche comprises the environmental interval conditions without considering the interspecific competition or predation. The **realized niche** consist the part of the fundamental niche where the species really occupies

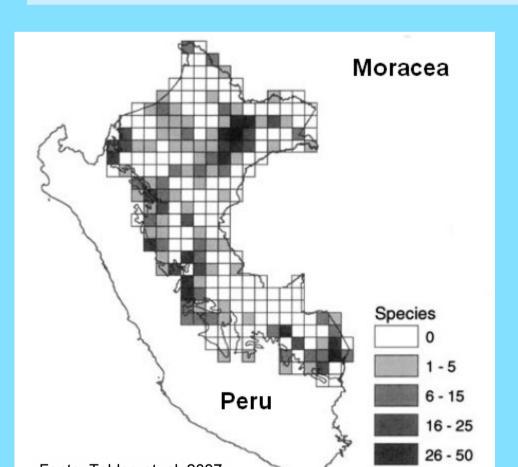


Common problems with data acquisition

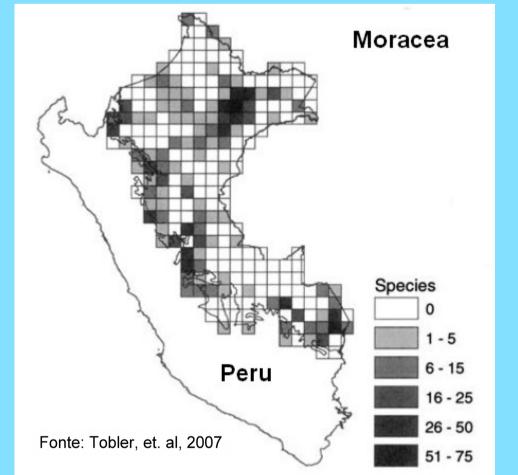




The majority of presence data, especially in tropical countries, are from herbariums and museum collections, which usually contain geographical positioning errors.



Cells with 0.5° of spatial resolution 50% of the collection are within 9 cells



MARS-INT

Source: Elith et al. (2006)

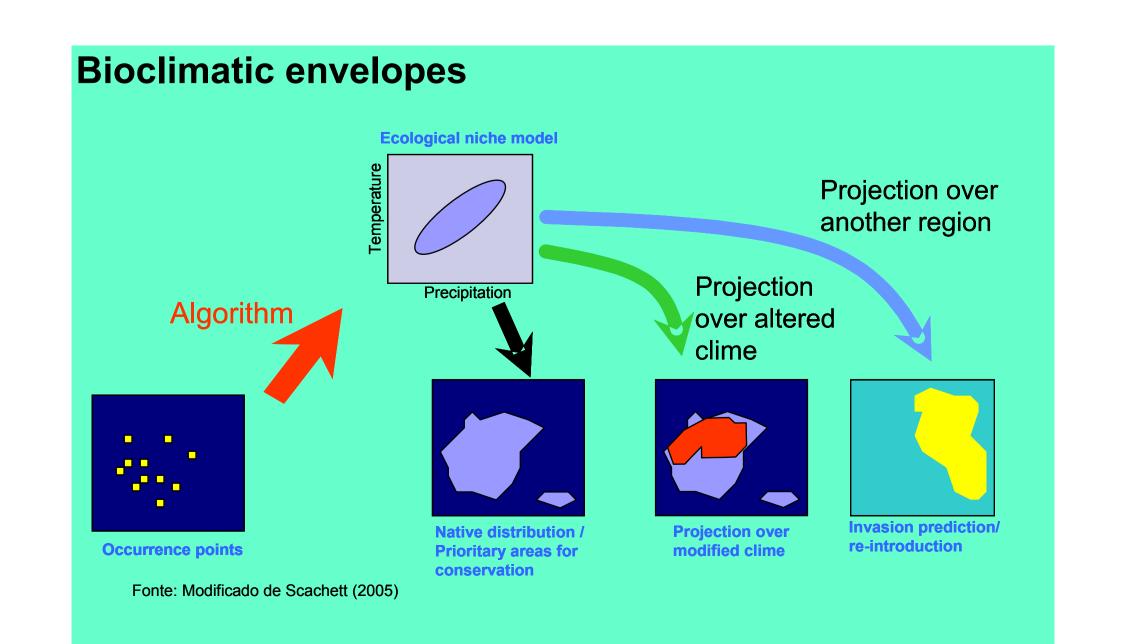
Modeling fundamental niche

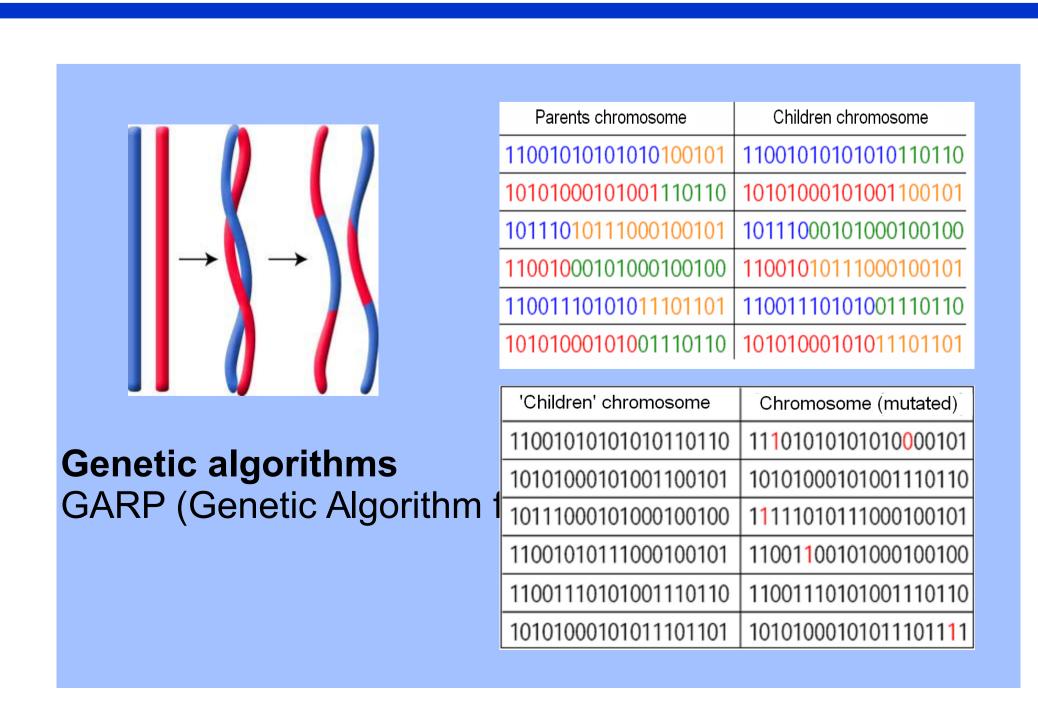
Surface temperature, NDVI,

features, categorical maps, etc.

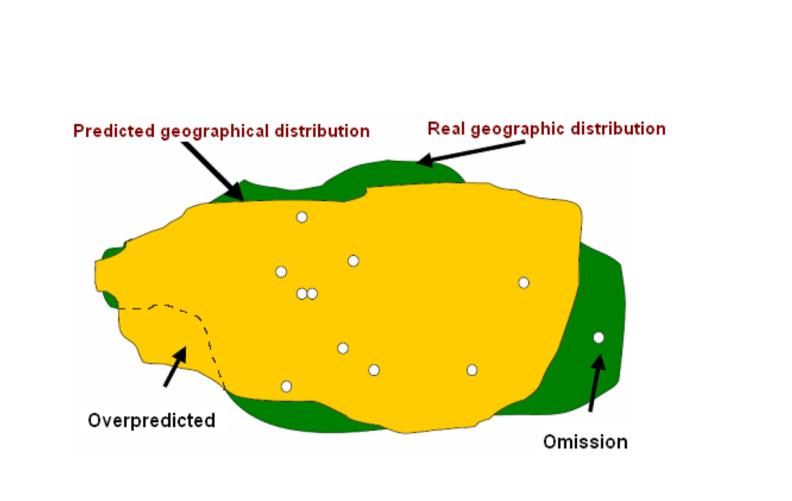
precipitation, topographic

and maximum entropy.





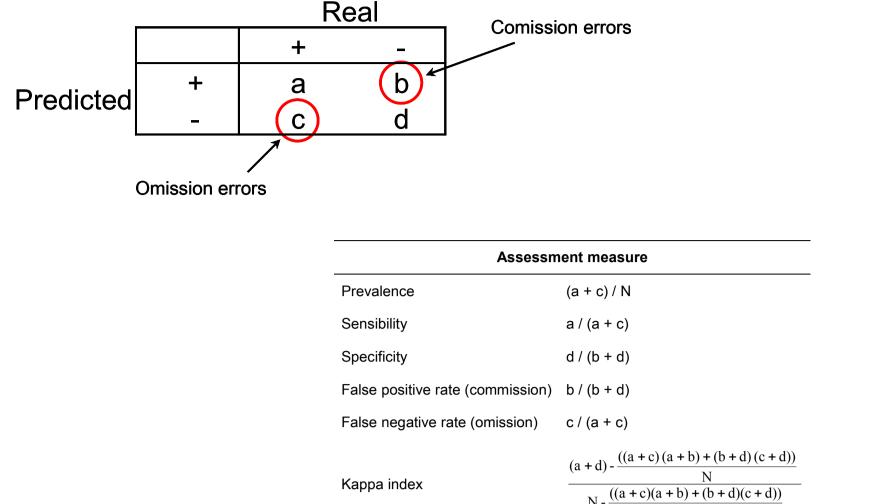
Model errors and performance measurements

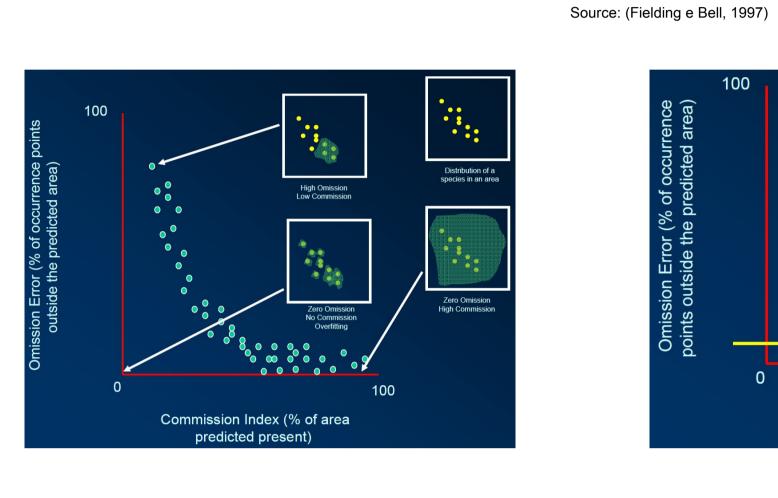


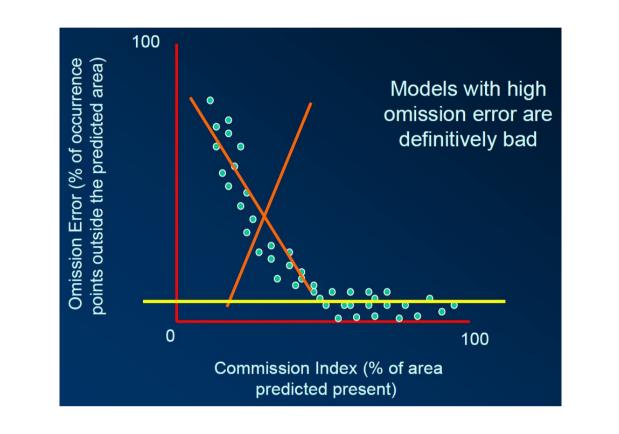
Region of the best

models

Commission Index (% of area

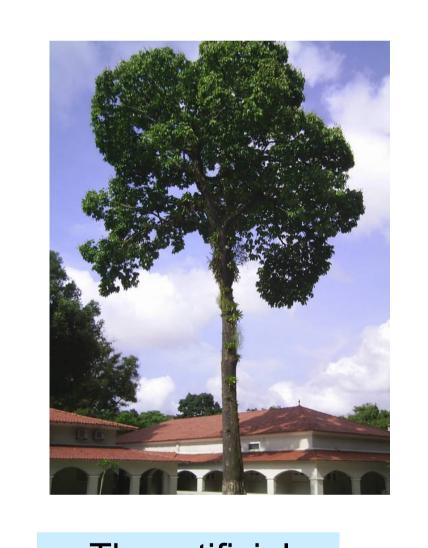


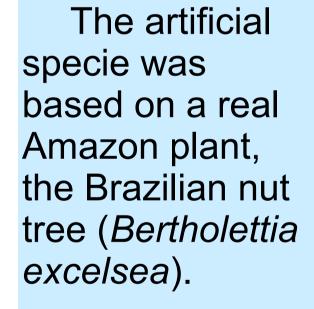


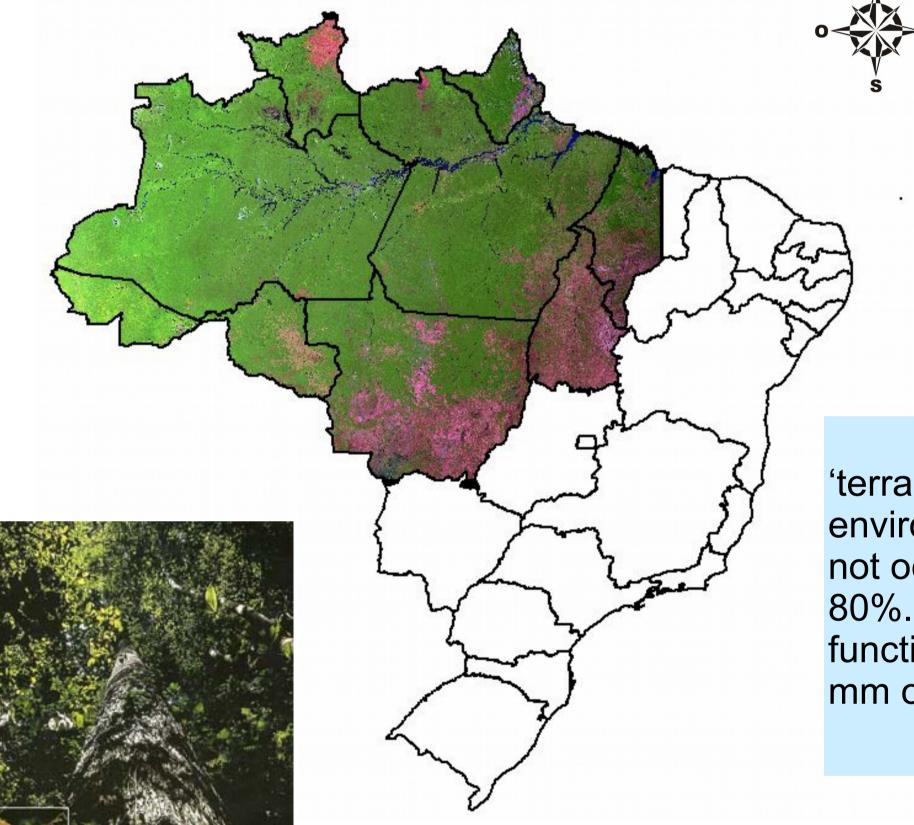


0 250 500 750 1000 Resolution (m) Some species are naturally rare. The number of samples may affect the models performance

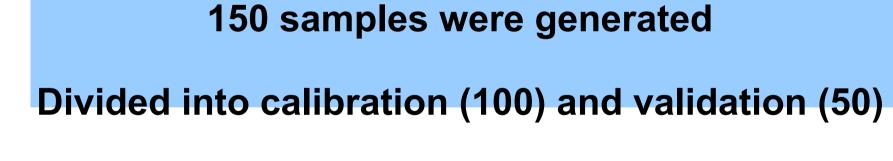
Study area and dataset

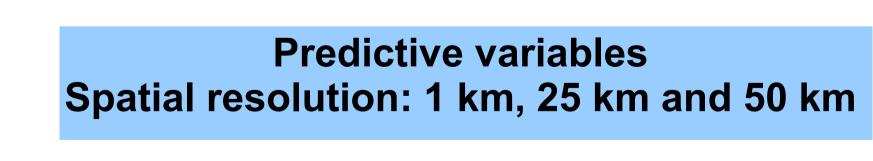


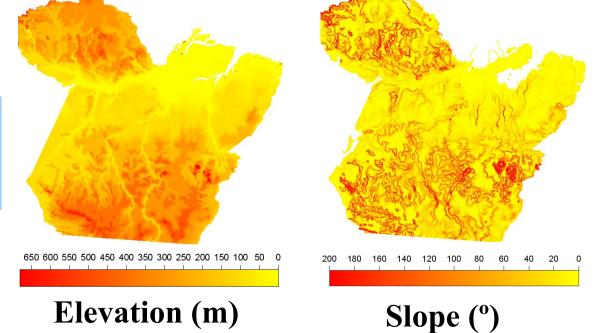


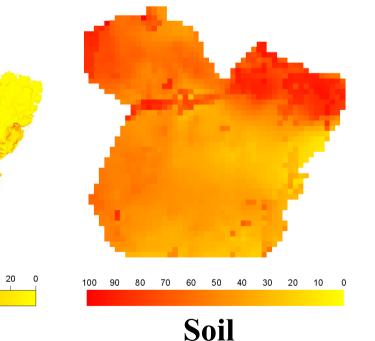


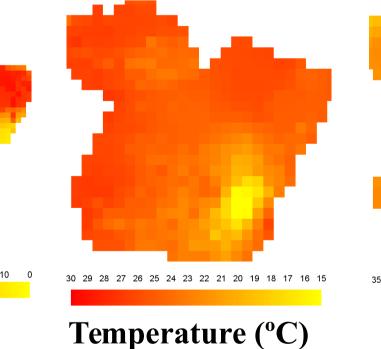
This virtual specie is found on cluster patterns, on 'terra firme' terrain, because does not tolerate acidic environments and saturated soils from floodplains. It does not occur on heights over 1600 m and slopes steeper than 80%. The temperature optimal range for physiological functioning is between 22°C and 32°C. It requires 1700 mm of annual precipitation and 10% to 65% soil moisture.

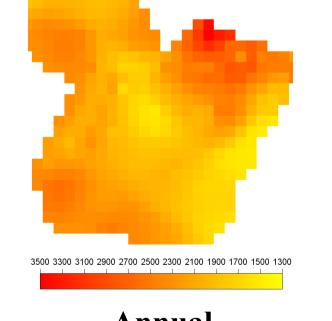






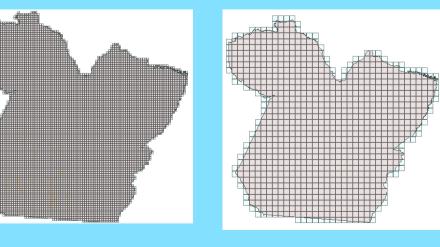




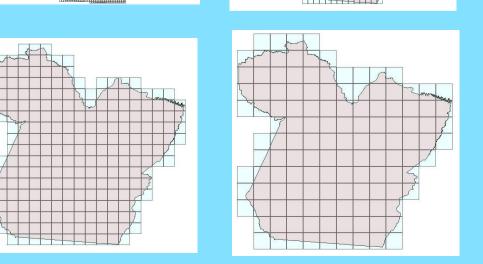


Precipitation (mm)

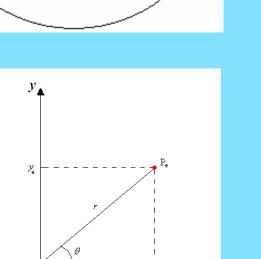
Positioning errors simulation

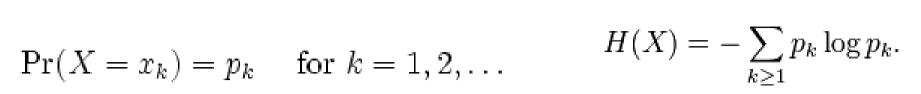


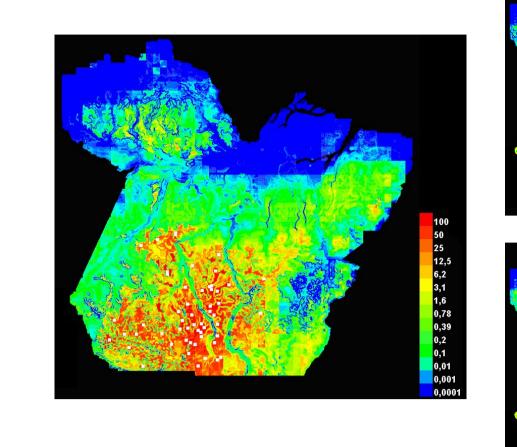
Coordinate displacements at four scales (10 km, 0.25°, 0.5° and 1°) were introduced using two methods: sample coordinates projection towards cellular center point, and errors with normal distribution in polar coordinates parameters.

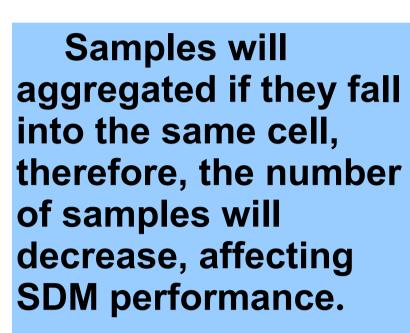


The first approach reflected changes in grain size and is commonly adopted when dealing with herbarium and museum database, whereas the second approach simulated the problem in obtaining a GPS signal under the forest canopy

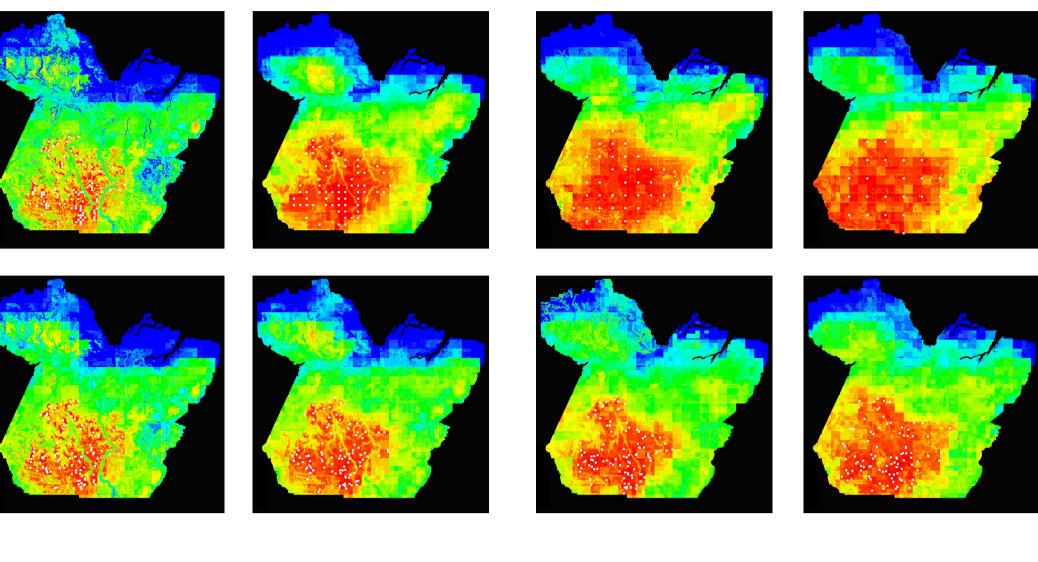








Din periormanee.								
	Nº Samples							
Training	100							
10 km	100							
0,25°	66							
0,25° 0,5° 1°	39							
1°	24							



	Bioc		Omissão GAI		Maxent		
	Centroi	Polar	Centroi	Polar	Centroi	Polar	
	d		d		d		
10 km	0	0	14,67	10,67	4,14	3,4	
0,25°	0,67	0,61	4	3,33	5,52	4,76	
$0,5^{\circ}$	4	1,33	10,67	2	4,83	4,76	
1°	6	2,11	6	4	2,07	2,04	
		Áro	a mínim:	a			

	_		_, _			_, _	_,	
Área mínima								
		Bioclim		GARP		Maxent		
		Centroid	Polar	Centroid	Polar	Centroid	Polar	
	10 km	7,45	8,24	20,2	21,08	3,23	2,68	
	$0,25^{\circ}$	11	12,17	19,9	21,56	4,55	6,57	
	0,5°	10,05	12,57	19,61	22,47	5,31	8,38	
	1°	10,01	16,73	21,62	23,57	6,6	9,16	

Kappa						
	Bioclim		GARP		Maxent	
	Centroid	Polar	Centroid	Polar	Centroid	Polar
10 km	0,5036	0,4456	0,4121	0,4462	0,3994	0,4045
$0,25^{\circ}$	0,3502	0,3474	0,3909	0,4053	0,3384	0,3442
0,5°	0,3051	0,3229	0,3841	0,3912	0,3381	0,3352
1°	0,2916	0,3004	0,3893	0,395	0,3304	0,3276

	AUC						
	Bioclim		GARP		Maxent		
	Centroid	Polar	Centroid	Polar	Centroid	Pola	
10 km	0,993	0,980	0,923	0,959	0,978	0,97	
0,25°	0,975	0,979	0,912	0,941	0,933	0,96	
0,5°	0,978	0,973	0,907	0,933	0,923	0,96	
1°	0,925	0,955	0,893	0,925	0,936	0,94	

Conclusions

1. More than one measure is needed to evaluate the sensibility, each model performance varied according to the assessment metric and type of error (cellular center projection/polar parameters).

2. If the error are up to 10 km, it is possible to apply any of the models without bigger performance losses;

3. When the errors have unknown spatial scale and under the premise of errors with Gaussian distributions, it is advisable to use GARP-BS which can cope more noisy data