

Ecological Niche-Based Species Distribution Modelling

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Approaches to Habitat or Environmental Suitability Assessment

❑ **Habitat Suitability Index (HSI):**

- An HSI is a numerical index that represents the capacity of a given habitat to support a selected species;
- These models are based on hypothesized species-habitat relationships rather than statements of proven cause and effect relationships.

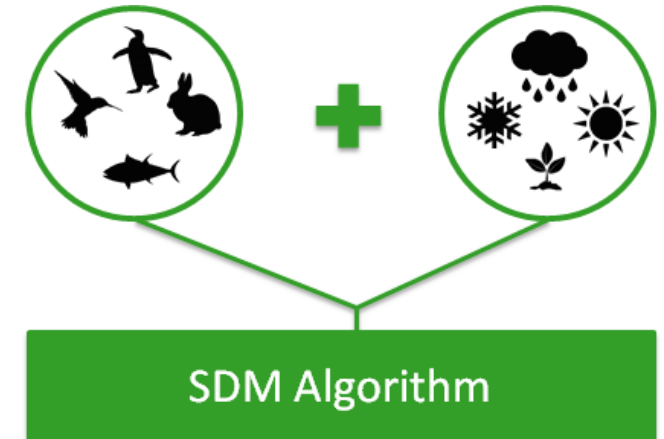


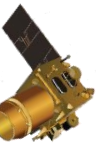
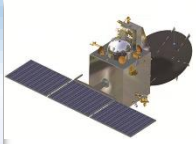
❑ **Mechanistic SDMs**

- Predict suitable environments using a species' physiological limitations in their tolerance to environmental conditions;
- Do not incorporate known occurrence records;
- Difficult to use due to required knowledge of species.

❑ **Correlative SDMs**

- Aim to estimate environmental conditions suitable for a species using known distributions;
- Correlate known occurrence records with various abiotic and biotic factors.

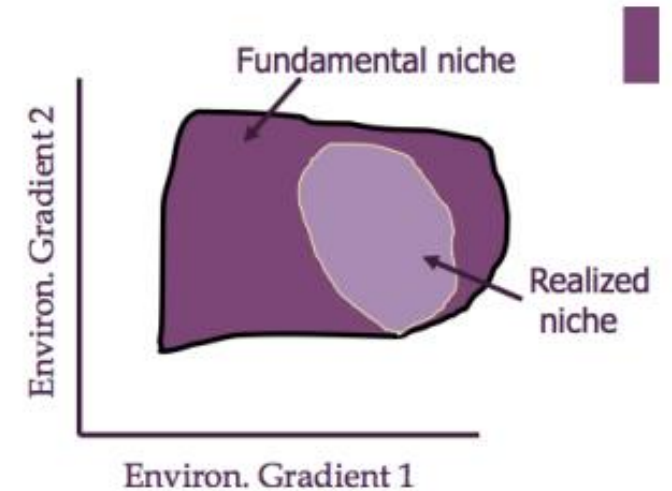




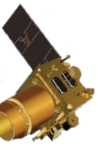
Definition & Type of Ecological Niches

Hutchinson defined ecological niche as:

- n-dimensional hyper-volume. n-equates the number of environmental factors;
- Full range of conditions and resources where an organism can live without interference is its fundamental niche.
- Realized niche includes interactions such as competition that may restrict environments where a species may live.



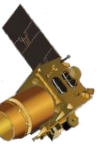
The relation between environmental gradients under ecological niche



Why a Ecological Niche based Correlative SDM)?

- Traditionally, only species locations are available. Available data is generally sparse. Do not tell where else species may occur.
- SDM predict likely areas of occurrence of species.
- Can also predict likely impacts of natural or anthropogenic disturbance on species.
- Species survey planning of RET species
- Reserve design and configuration
- Sites suitability for re-introduction

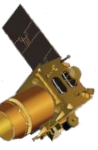
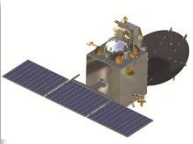




Assumptions in Ecological Niche Modelling

“Ecological niche models are based on the relationship between the species locations and their environmental covariates or predictors.”

- ✓ Over evolutionary time, species would be selected to adapt to a set of environmental variables, characteristic of its ecological niche.
- ✓ Species ecological niche should not change with changing environmental conditions (“**Niche conservatism**”)

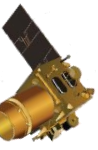


Ecological Niche Modelling Algorithms

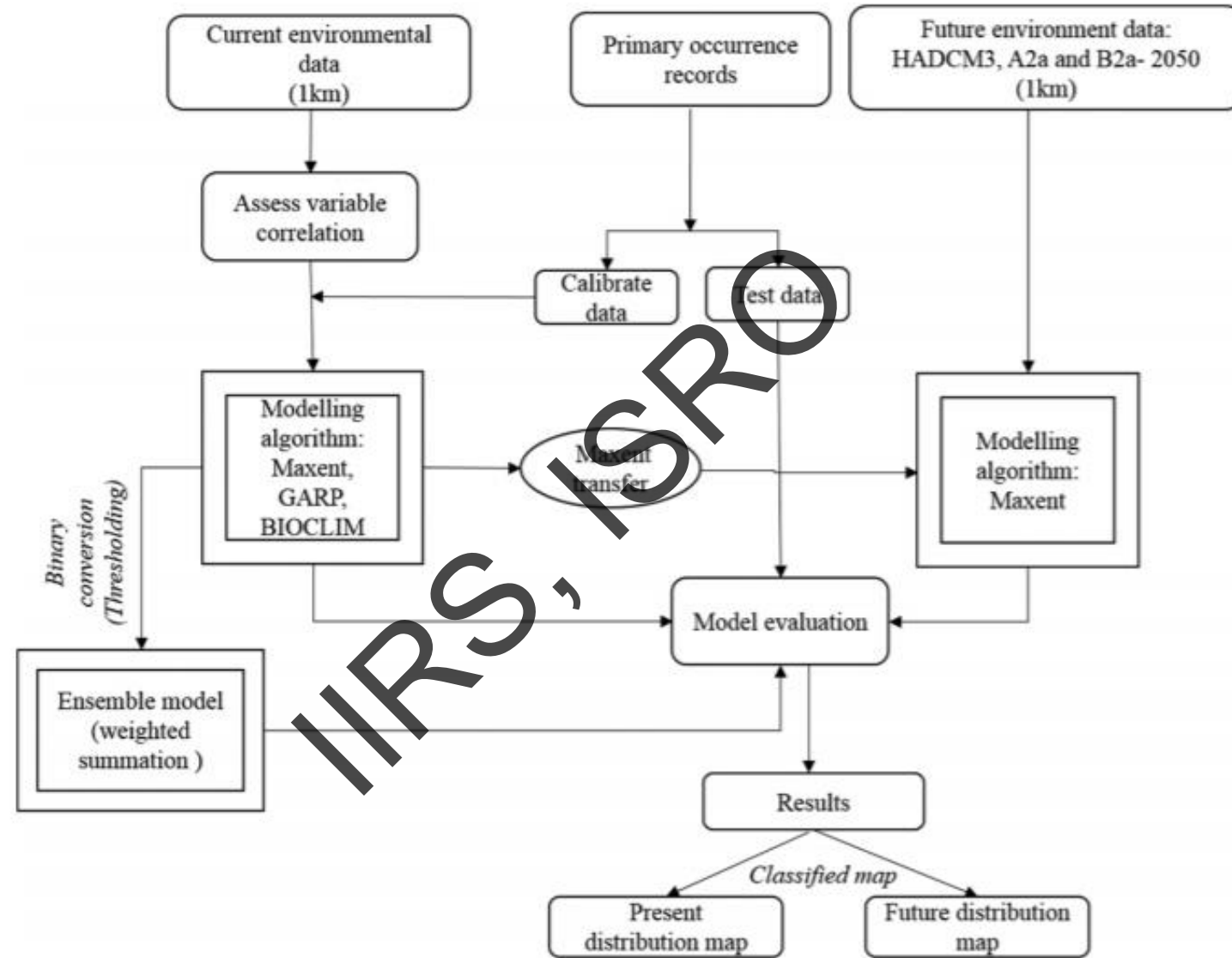
- Logistic Regression (Presence/ Absence data)
- Genetic Algorithm for Rule Set Production (GARP) (Presence data)
- MaxEnt (Maximum Entropy) (Presence data)

Ecological Niche Modelling Software

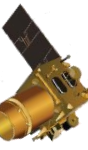
- [DIVA-GIS](#) is useful in preparation of bioclimatic variables
- [MaxEnt](#) uses presence only data and performs well when there are few presence records available.
- [ModEco](#) implements various SDM algorithms.
- [SPACES](#) is an online Environmental niche modeling platform



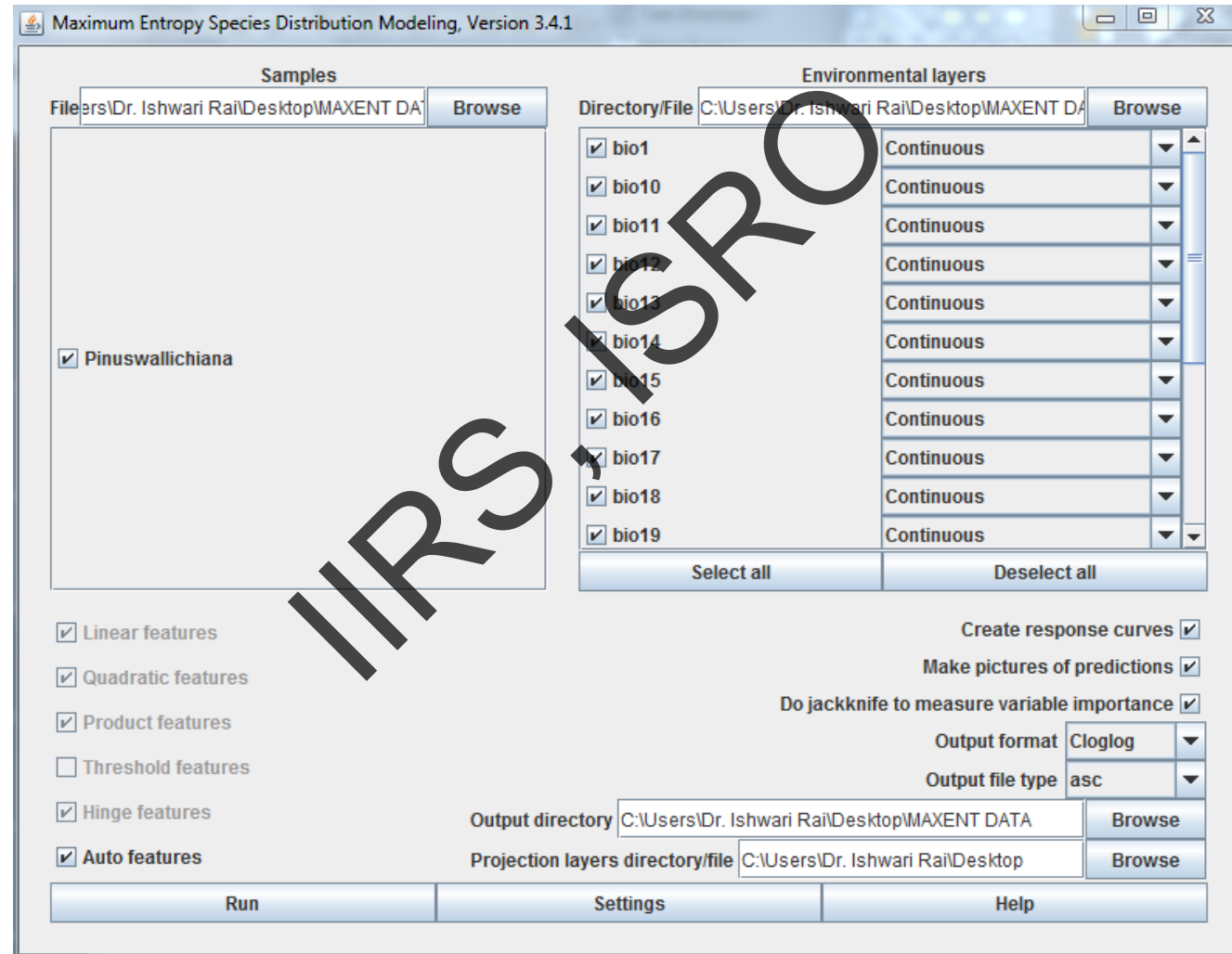
General Methodology



[Mapping invasion potential using ensemble modelling. A case study on Yushania maling in the Darjeeling Himalayas](#), V Srivastava, VC Griess, H Padalia - Ecological Modelling, 2018



The MaxEnt approach is used to estimate probability of distribution of target species by analyzing the probability distribution of maximum entropy ([Phillips et al., 2006](http://www.cs.princeton.edu)).



Maximum Entropy Species Distribution Modeling, Version 3.4.1

Samples

File: C:\Users\Dr. Ishwari Rai\Desktop\MAXENT DATA Browse

☒ Pinuswallichiana

Environmental layers

Directory/File: C:\Users\Dr. Ishwari Rai\Desktop\MAXENT DATA Browse

Variable	Data Type
<input checked="" type="checkbox"/> bio1	Continuous
<input checked="" type="checkbox"/> bio10	Continuous
<input checked="" type="checkbox"/> bio11	Continuous
<input checked="" type="checkbox"/> bio12	Continuous
<input checked="" type="checkbox"/> bio13	Continuous
<input checked="" type="checkbox"/> bio14	Continuous
<input checked="" type="checkbox"/> bio15	Continuous
<input checked="" type="checkbox"/> bio16	Continuous
<input checked="" type="checkbox"/> bio17	Continuous
<input checked="" type="checkbox"/> bio18	Continuous
<input checked="" type="checkbox"/> bio19	Continuous

Select all Deselect all

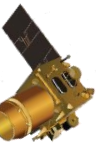
☒ Linear features
☒ Quadratic features
☒ Product features
☐ Threshold features
☒ Hinge features
☒ Auto features

Create response curves ☒
 Make pictures of predictions ☒
 Do jackknife to measure variable importance ☒

Output format: Cloglog
 Output file type: asc

Output directory: C:\Users\Dr. Ishwari Rai\Desktop\MAXENT DATA Browse
 Projection layers directory/file: C:\Users\Dr. Ishwari Rai\Desktop Browse

Run Settings Help



Maximum Entropy Parameters

Basic Advanced Experimental

☐ Random seed
☒ Give visual warnings
☒ Show tooltips
☒ Ask before overwriting
☐ Skip if output exists
☒ Remove duplicate presence records
☒ Write clamp grid when projecting
☒ Do MESS analysis when projecting

Random test percentage: 0
 Regularization multiplier: 1
 Max number of background points: 10000
 Replicates: 1
 Replicated run type: Crossvalidate
 Test sample file: Browse

Maximum Entropy Parameters

Basic Advanced Experimental

☒ Add samples to background
☐ Add all samples to background
☐ Write plot data
☒ Extrapolate
☒ Do clamping
☒ Write output grids
☒ Write plots
☐ Append summary results to maxentResults.csv file
☒ Cache ascii files

Maximum iterations: 500
 Convergence threshold: 0.00001
 Adjust sample radius: 0
 Log file: maxent.log
 Default prevalence: 0.5
 Apply threshold rule:
 Bias file: Browse

Case study: Predicting Bushmint's Invasion Range in India

Multicollinearity

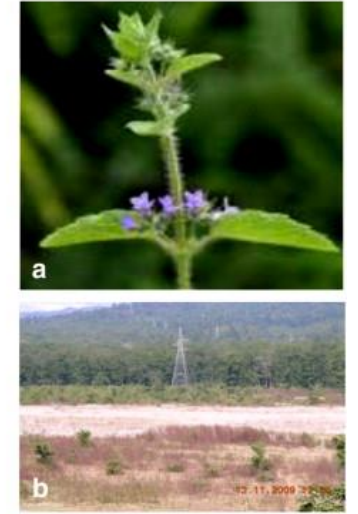
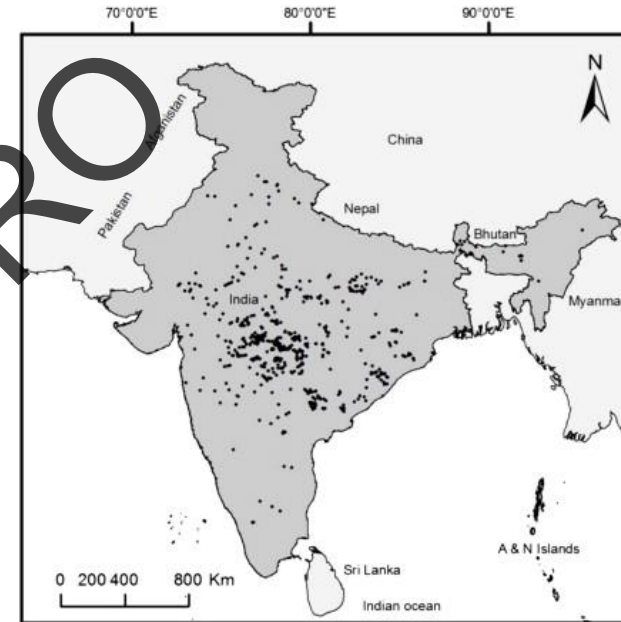
Table A2
Multi-collinearity test by using cross-correlations (Pearson correlation coefficients, r) among environmental variables.

Variables	Bio1	Bio2	Bio3	Bio4	Bio5	Bio6	Bio7	Bio8	Bio9	Bio10	Bio11	Bio12	Bio13	Bio14	Bio15	Bio16	Bio17	Bio18	Bio19	LULC	ELE	SLO	ASP
Bio1	1																						
Bio2	-.01	1																					
Bio3	.94	.94	1																				
Bio4	.98	.98	.90	1																			
Bio5	-.01	-.01	.94	.98	1																		
Bio6	-.01	.99	.94	.97	-.01	1																	
Bio7	-.01	-.01	.93	.99	-.01	.99	1																
Bio8	-.01	-.01	.93	.99	-.01	-.01	-.01	1															
Bio9	.82	.83	.80	.84	.83	.82	.83	.83	1														
Bio10	-.01	-.01	.93	.99	-.01	.99	-.01	.83	.83	1													
Bio11	-.01	-.01	.94	.97	-.01	.99	-.01	.81	-.01	.81	1												
Bio12	-.69	-.69	-.62	-.73	-.70	-.69	-.70	-.70	-.64	-.70	-.70	1											
Bio13	-.51	-.51	-.44	-.56	-.52	-.51	-.52	-.52	-.44	-.51	-.50	.85	1										
Bio14	-.53	-.54	-.50	-.62	-.54	-.53	-.55	-.55	-.70	-.55	-.50	.78	.84	1									
Bio15	.45	.45	.51	.42	.45	.48	.44	.45	.58	.44	.45	-.14	.11	-.46	1								
Bio16	-.63	-.63	-.55	-.68	-.64	-.63	-.64	-.64	-.57	-.64	-.61	.99	.97	.73	.98	1							
Bio17	-.74	-.75	-.73	-.78	-.75	-.75	-.75	-.75	-.77	-.74	-.72	.88	.74	.88	-.53	.81	1						
Bio18	-.74	-.74	-.68	-.78	-.75	-.74	-.75	-.75	-.70	-.74	-.72	.99	.90	.78	-.18	.97	.88	1					
Bio19	-.72	-.74	-.72	-.67	-.73	-.73	-.73	-.71	-.63	-.71	-.72	.78	.67	.59	-.49	.70	.85	.84	1				
LULC	-.66	-.66	-.66	-.66	-.66	-.66	-.66	-.66	-.63	-.66	-.66	.65	.65	.64	-.66	.65	.65	.64	.65	1			
ELE	-.01	-.01	-.01	-.01	-.01	-.01	-.01	-.01	-.99	-.01	-.01	.97	.97	.94	-.99	.97	.94	.96	.96	.66	1		
SLO	-.46	-.46	-.52	-.49	-.47	-.40	-.48	-.46	-.41	-.47	-.44	.61	.59	.62	-.53	.60	.64	.63	.62	.28	.46	1	
ASP	-.84	-.84	-.85	-.85	-.84	-.83	-.84	-.84	-.82	-.84	-.84	.85	.85	.84	-.85	.86	.84	.85	.85	.60	.84	.47	1

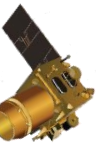
Note: If two variables had >0.8 , only one of them was selected in the same model. Correlations were significant at $\alpha=0.05$ (calculated by statistic software SPSS 16.0).

Presence data:

463 presence points in a total of 16,518 biodiversity plots (2002 to 2007). 67 records of bushmint were compiled from the independent surveys (2007 to 2011).



Bushmint in India (black dots depict presence locations, field photographs: (a) Bushmint in flowering during September and (b) monothickets of bushmint (in light violet color) along the river course during December in Dun valley).



Environmental data layers

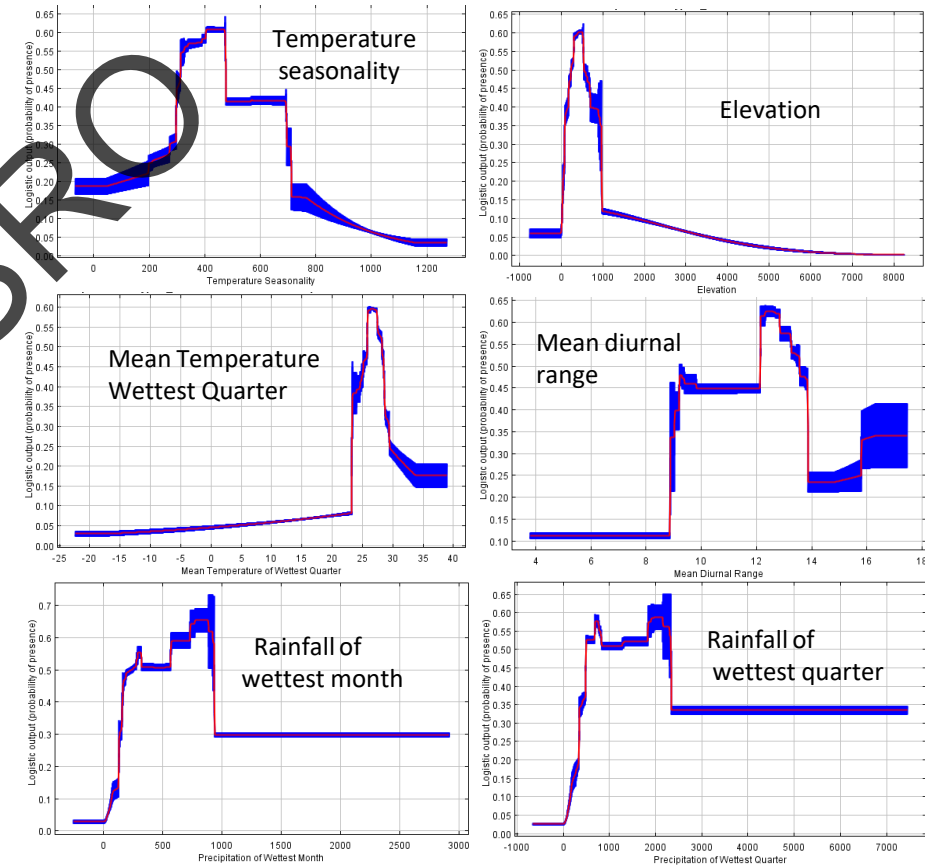
Table A.1

Environmental variables used in the study and their percentage contribution.

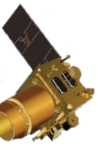
Code	Environmental variables	Unit	% Contribution
Bio1	Annual mean temperature	°C	
Bio2	Mean diurnal range (mean of monthly max. and min. temp.)	°C	
Bio3	Isothermality $((\text{Bio2}/\text{Bio7}) \times 100)$	–	72.2
Bio4	Temperature seasonality (standard deviation $\times 100$)	C of V	
Bio5	Maximum temperature of warmest month	°C	
Bio6	Minimum temperature of coldest month	°C	
Bio7	Temperature annual range (Bio5–Bio6)	°C	
Bio8	Mean temperature of wettest quarter	°C	
Bio9	Mean temperature of driest quarter	°C	0.7
Bio10	Mean temperature of warmest quarter	°C	
Bio11	Mean temperature of coldest quarter	°C	
Bio12	Annual precipitation	mm	1.3
Bio13	Precipitation of wettest period	mm	
Bio14	Precipitation of driest period	mm	1.3
Bio15	Precipitation seasonality (CV)	C of V	1.0
Bio16	Precipitation of wettest quarter	mm	
Bio17	Precipitation of driest quarter	mm	
Bio18	Precipitation of warmest quarter	mm	
Bio19	Precipitation of coldest quarter	mm	0.8
LULC	Land use and land cover	15 types	21.9
ELE	Elevation	m	
SLO	Slope	°	0.8
ASP	Aspect	°	

Note: The highlighted variables, selected through multi-collinearity test, were used in modeling.

Response of Bushmint towards key environmental variables

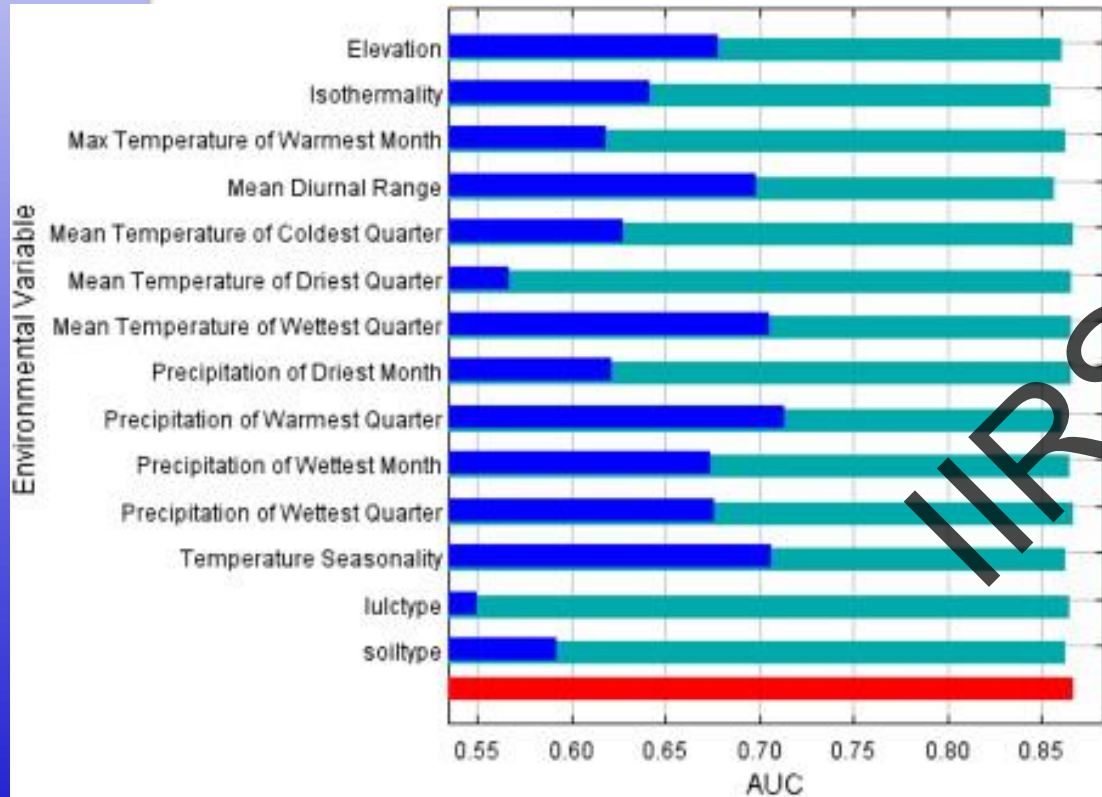


Mean response and ± 1 SD calculated over 10 replicates.

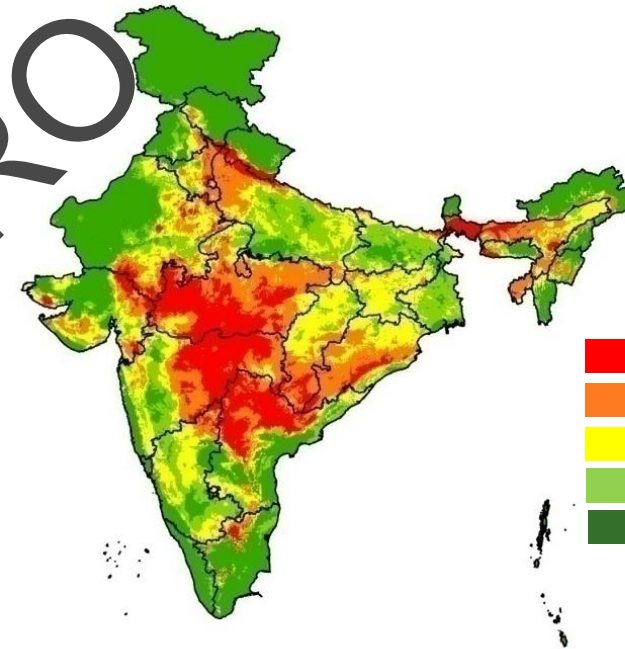


Case study: Predicting Bushmint's Invasion Range in India

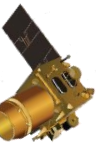
AUC Test Gain for different variables



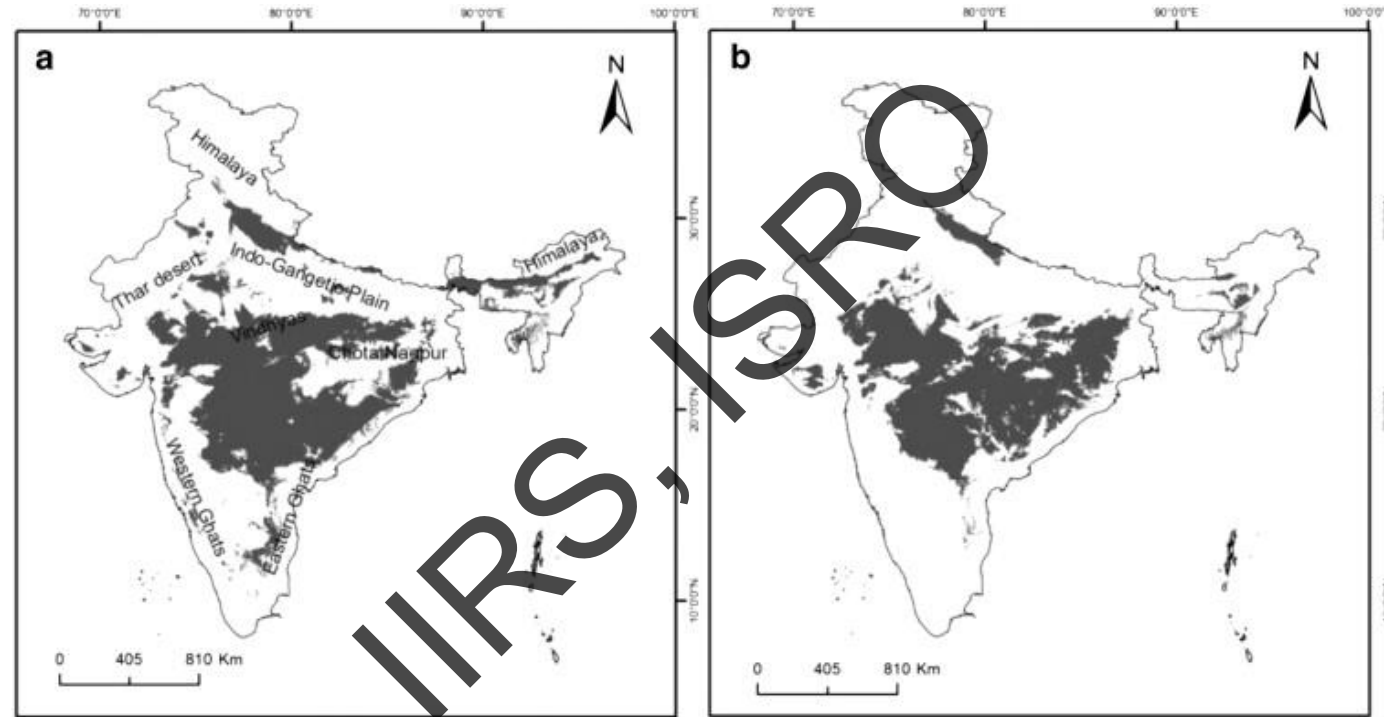
Jackknife test for AUC



MaxEnt based probability distribution map of bushmint (*Hyptis suaveolens*) over India.

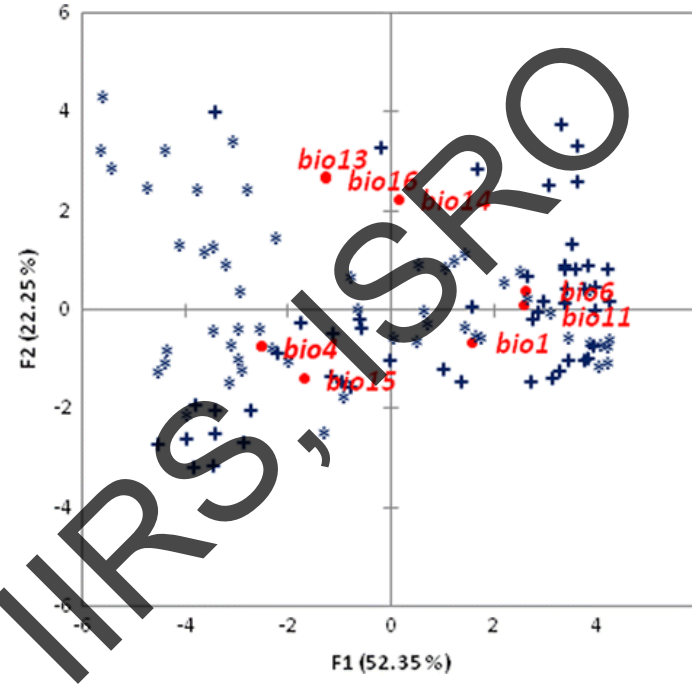
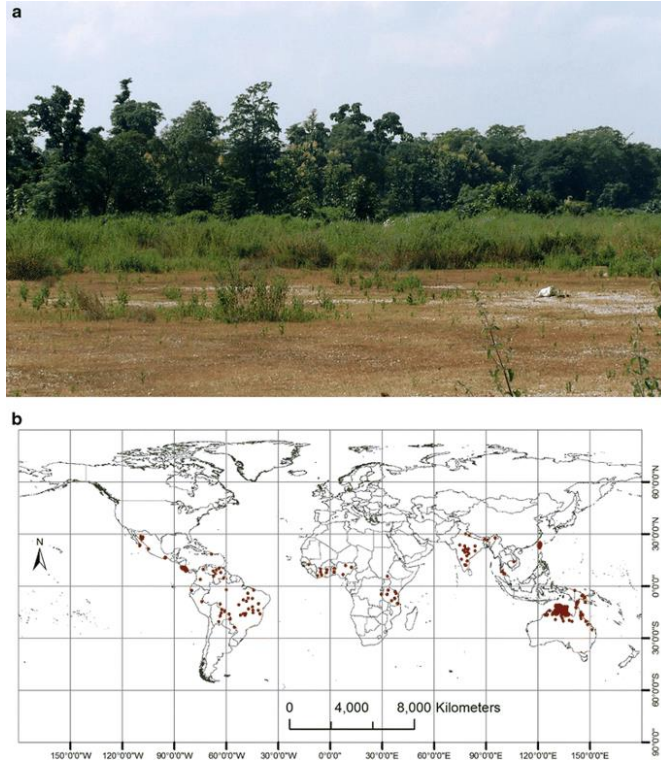
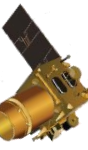


Comparison of MaxEnt and GARP Outputs

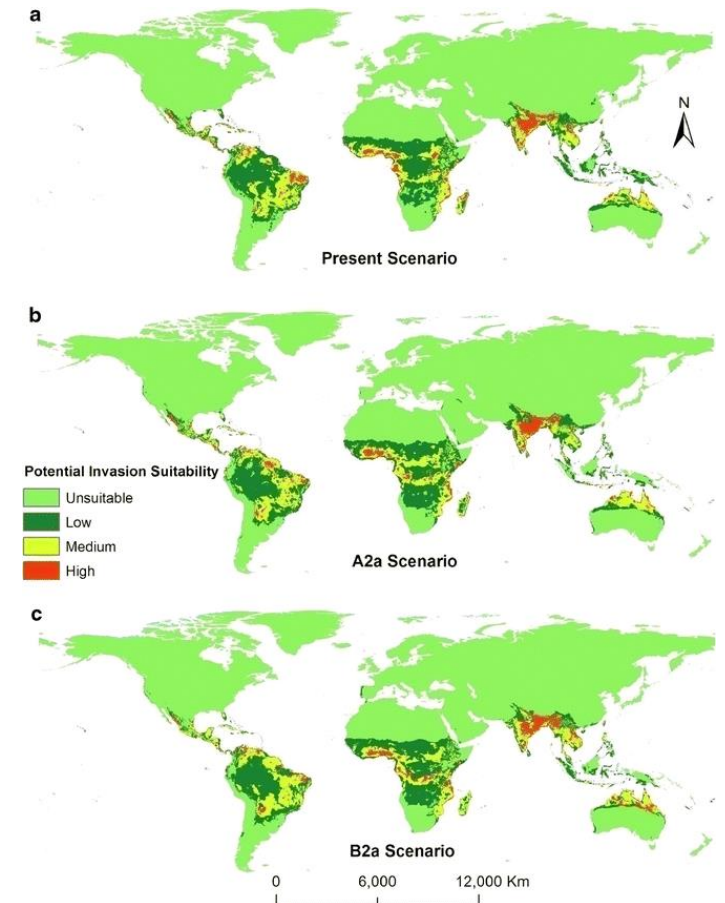


GARP had a relatively lower area under curve (AUC) score (AUC: 0.75), suggesting its lower ability in discriminating the suitable/unsuitable sites. MaxEnt performed better with an AUC value of 0.86.

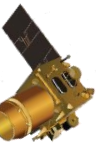
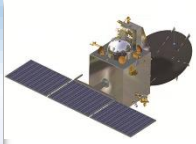
Modeling potential invasion range of alien invasive species, *Hyptis suaveolens* (L.) Poit. in India:
 Comparison of MaxEnt and GARP (2014) [Ecological Informatics](#), 36-43
[Hitendra Padalia, Vivek Srivastava, S.P.S. Kushwaha Volume 22,](#)



Plot of PC-1 and PC-2 scores of climatic profile of native and invaded sites of bushmint. The *plus* and *star* symbol in the plot represents bushmint's occurrence records from native and invaded populations respectively.

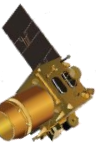
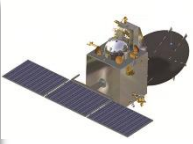


How climate change might influence the potential distribution of weed, bushmint (*Hyptis suaveolens*)?
 Hitendra Padalia, Vivek Srivastava & S. P. S. Kushwaha (2015) Environmental Monitoring and Assessment, 187, 2010



Considerations in Ecological Niche Modelling

- 1) All sampling data are incomplete and potentially biased;
 - Location data errors
 - Sampling prevalence and sample size
 - Spatial extent and background selection
- 2) No single model works best for all species, in all areas, at all spatial scales, and over time; and
 - No two models are the same
 - Species characteristics matter
 - Evaluation and calibration



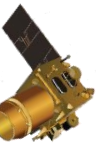
Considerations in Ecological Niche Modelling

3) Predictor variables must capture distribution constraints

- Multi-collinearity
- Missing a key environmental layer

4) The results of species distribution models should be treated like a hypothesis to be tested and validated with additional sampling and modeling in an iterative process.

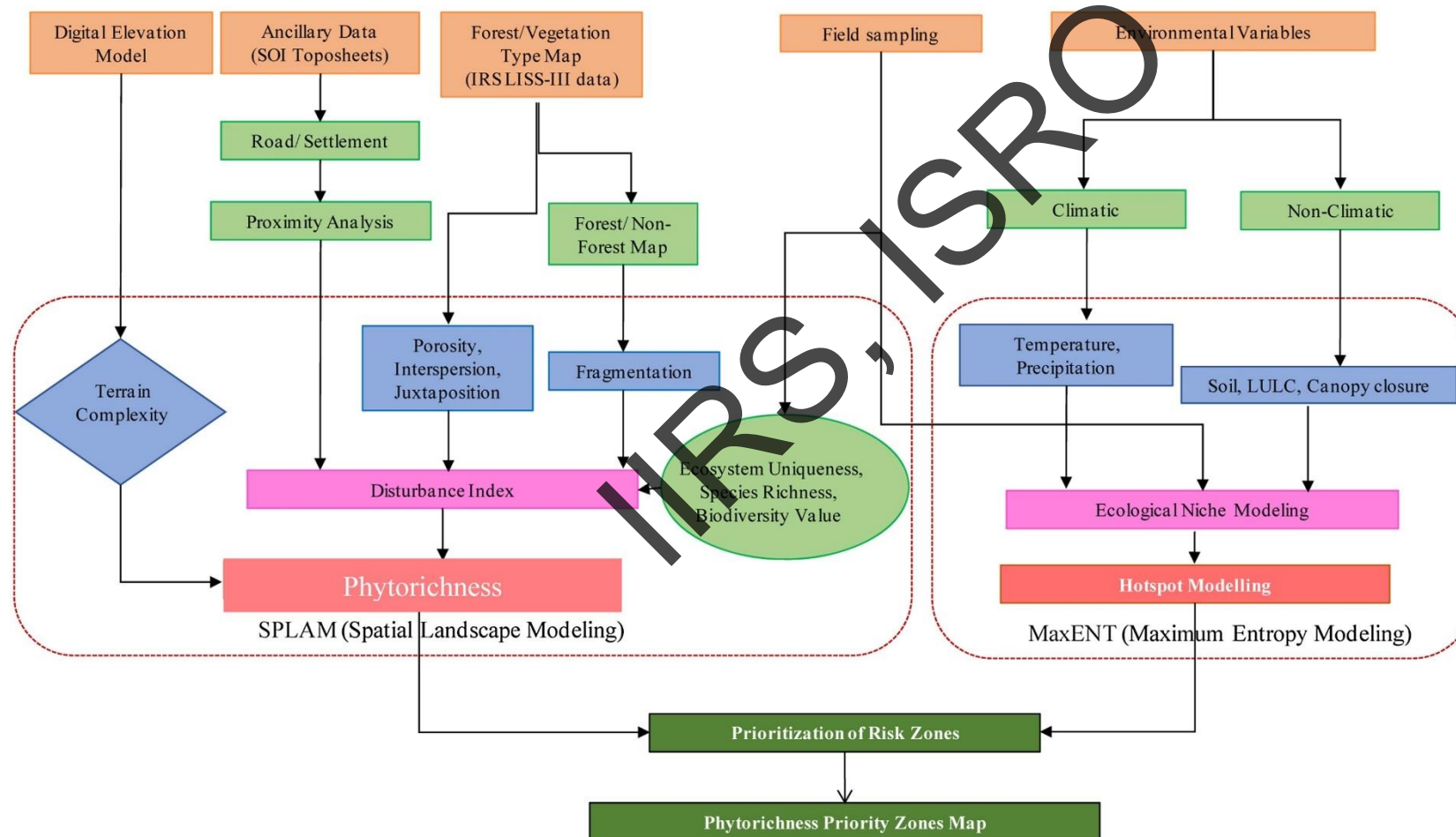
- Model interpretation
- Creating “yes/no” maps with thresholds



New trends in Ecological Niche based SDM

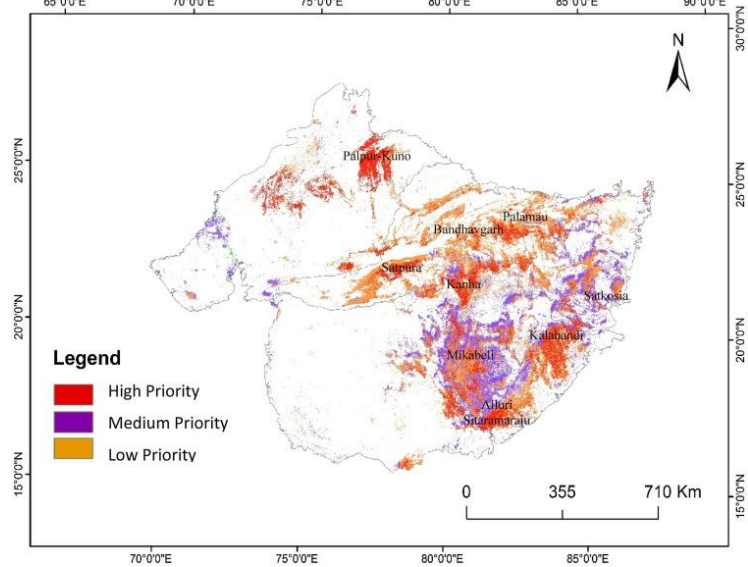
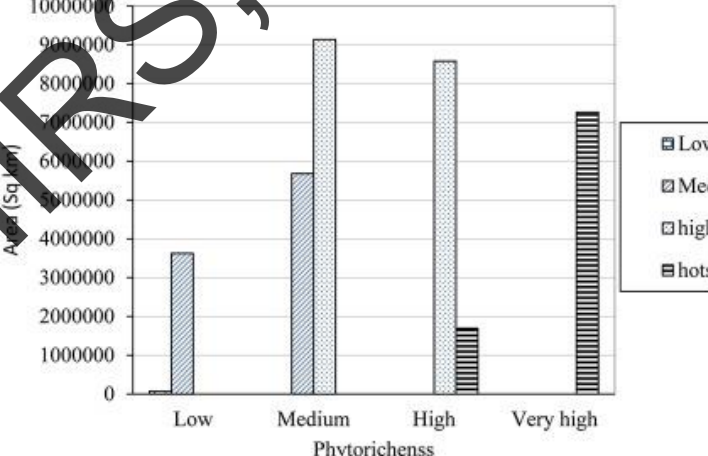
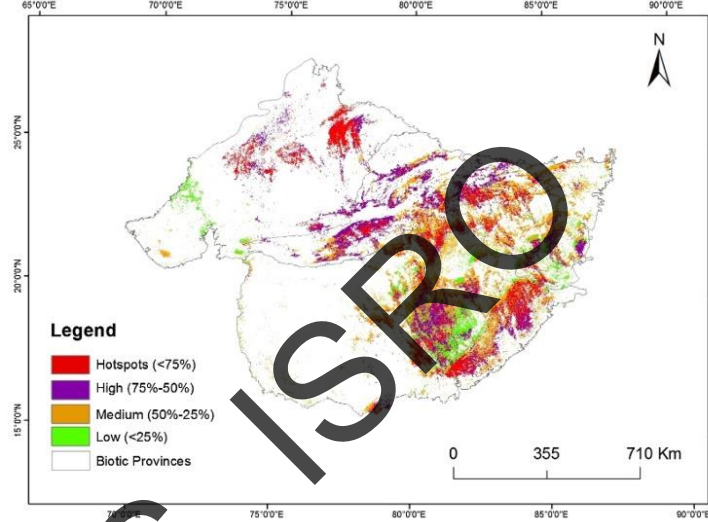
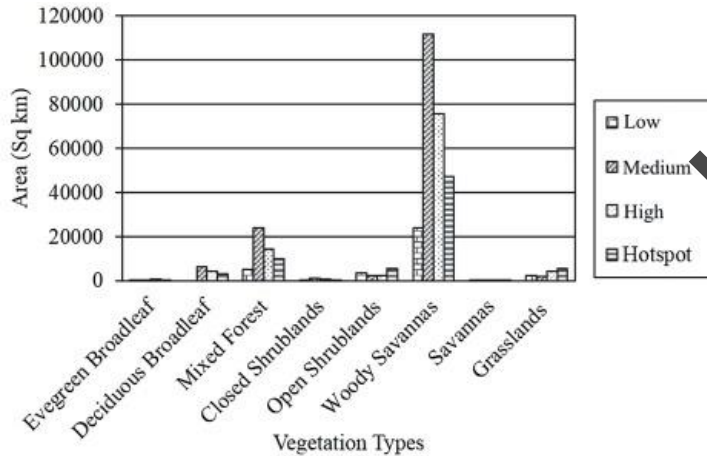
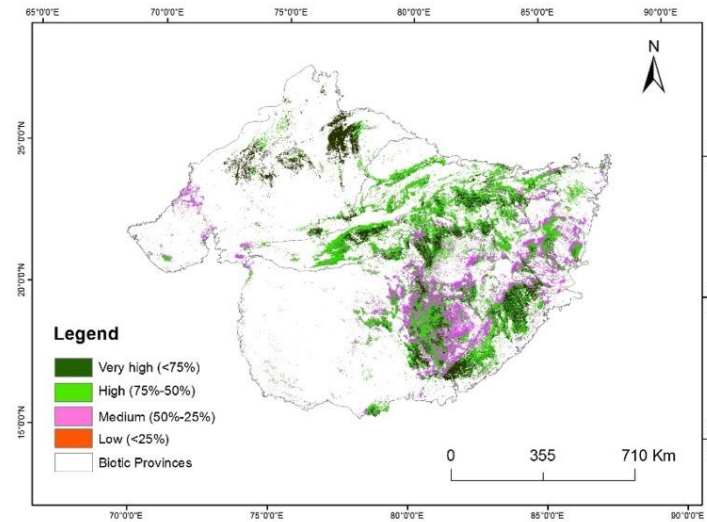
- ❑ Phylogeography and population dynamics of an endemic oak (*Quercus fabri* Hance) in subtropical China revealed by **molecular data and ecological niche modeling**;
- ❑ Evidence of **ecological niche shift** in *Rhododendron ponticum* (L.) in Britain: Hybridization as a possible cause of rapid niche expansion;
- ❑ Is there a correlation between **abundance and environmental suitability** derived from ecological niche modelling? A meta-analysis;
- ❑ Spatial modelling of **congruence of native biodiversity and potential hotspots of forest invasive species** (FIS) in central Indian landscape

Spatial modelling of congruence of native biodiversity and potential hotspots of forest invasive species (FIS) in central Indian landscape



111 Forest Invasive Species (FIS) Invading Central Indian landscape were considered.

Spatial modelling of congruence of native biodiversity and potential hotspots of forest invasive species (FIS) in central Indian landscape

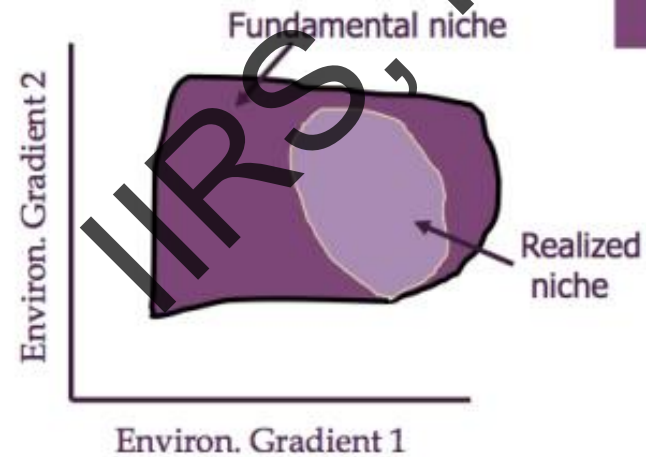


Priority zones for phytoreichens conservation.

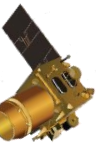
Spatial modelling of congruence of native biodiversity and potential hotspots of forest invasive species (FIS) in central Indian landscape (2017) [Hitendra Padalia & Utsav Bahuguna, J. for Nature Conservation, 36, 29-37](#)

Constraints in Ecological Niche Based SDM

- ENM does not include biotic interactions (competition), dispersal and lag effects(e.g. invasive species).
- Extrapolation violates several statistical and ecological assumptions.



The relation between environmental gradients under ecological niche



Thank You

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