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Projected Distribution of 218 Invasive Plant Species in India under climate change: Identification of Hotspots and Management Implications

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Agenda

- Problem Statement
- Materials and Methods
- Results
- Future Direction

Invasive Alien Plants



Lantana Camara



Mikania micrantha



Jatropha Curcas



Catharanthus roseus



Why is this study important?



Alien plants growing together threatening tiger habitats: Study
20 January 2023



Think twice before you eat *Ponnaganti kura*
24 November 2017

- **India has numerous Invasive Alien species pose an unprecedented threat to biodiversity and ecosystems at different spatial scales.**
- **It is crucial to understand their current distribution and also under climate change.**

Species Distribution Modelling



- The Probability of Occurrence of each species in a particular location is modelled as a function of climatic Variables.
- This is typically obtained by fitting logistic regression model, F is the logistic function.

Materials and Methods

- Collection of occurrence records and Climate Data
- Processing of data
- Variable Selection Method
- Model building (Ensemble)
- Method Selection
- Modelling and Projections
- Categorization of species
- Post Modelling Analysis
- Results

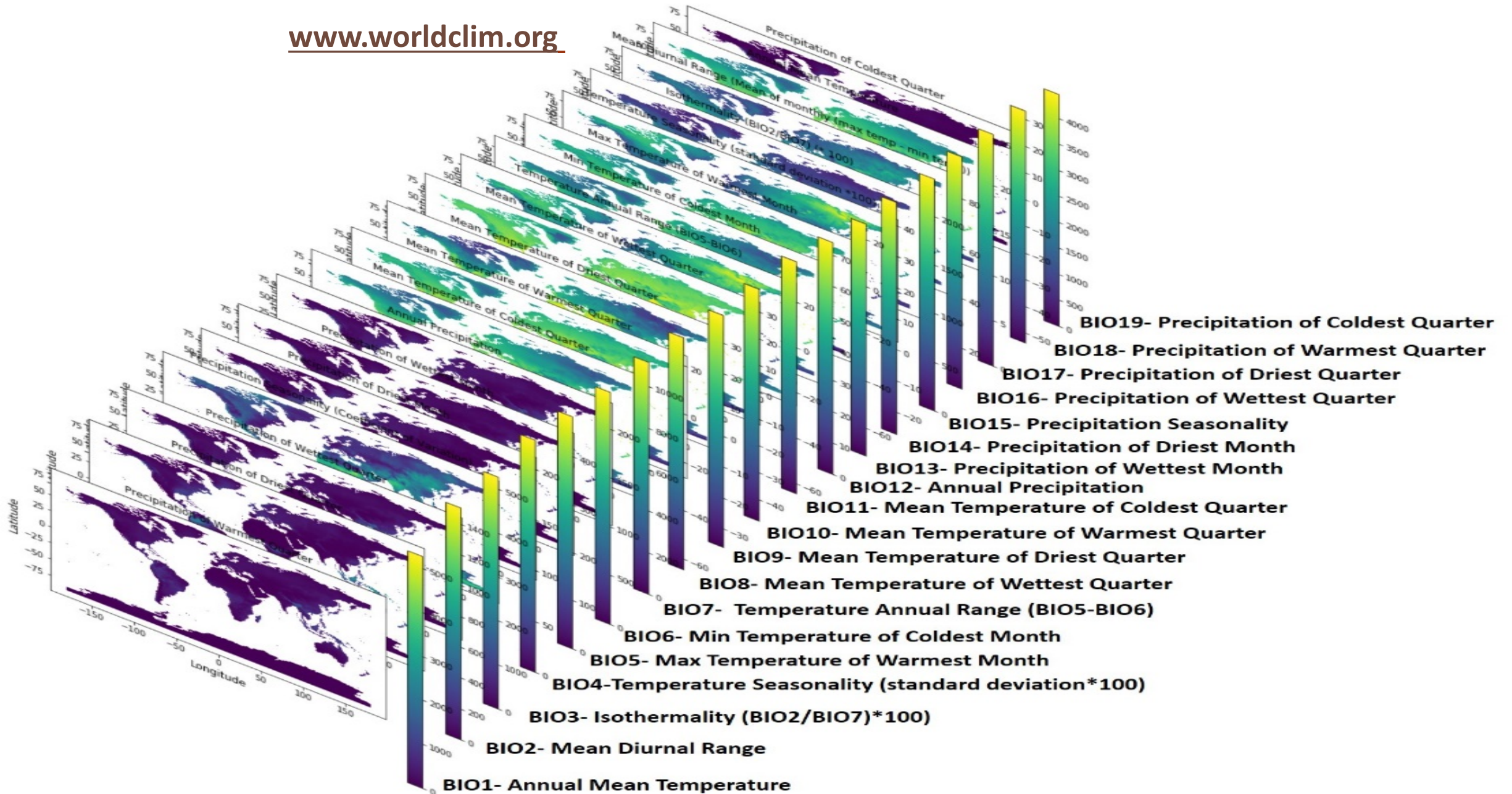
What is GBIF?

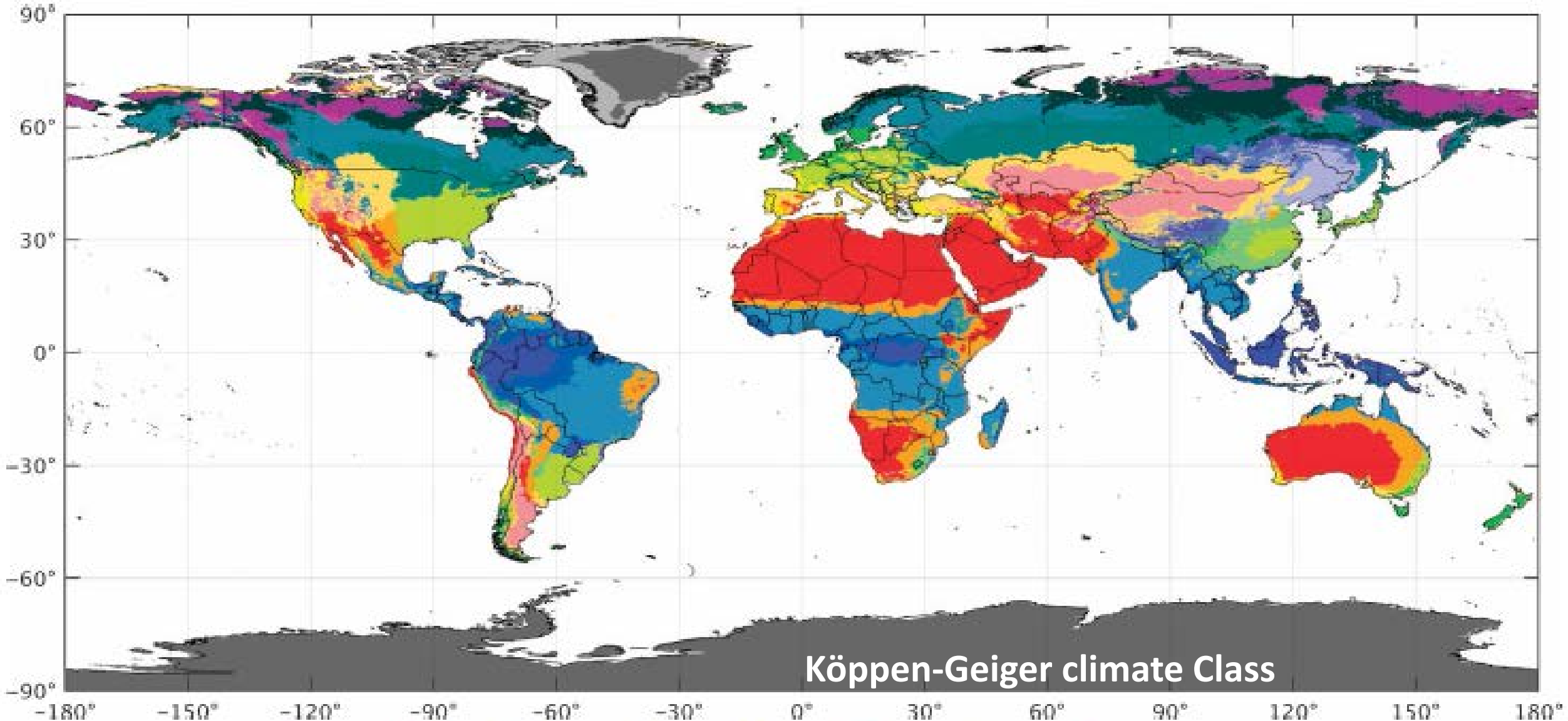
GBIF—the Global Biodiversity Information Facility—is an international network and data infrastructure funded by the world's governments and aimed at providing anyone, anywhere, open access to data about all types of life on Earth.



Bioclimatic Variable

www.worldclim.org





 Af	 BWh	 Csa	 Cwa	 Cfa	 Dsa	 Dwa	 Dfa	 ET
 Am	 BWk	 Csb	 Cwb	 Cfb	 Dsb	 Dwb	 Dfb	 EF
 Aw	 BSh	 Csc	 Cwc	 Cfc	 Dsc	 Dwc	 Dfc	
	 BSk				 Dsd	 Dwd	 Dfd	

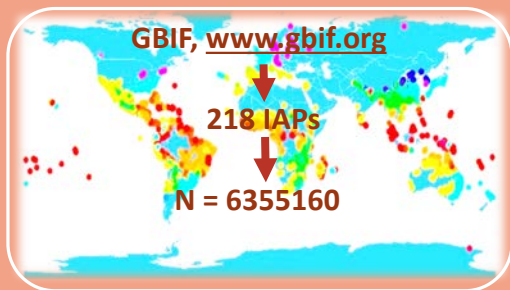
Species Selection

<https://ilora2020.wixsite.com/ilora2020>

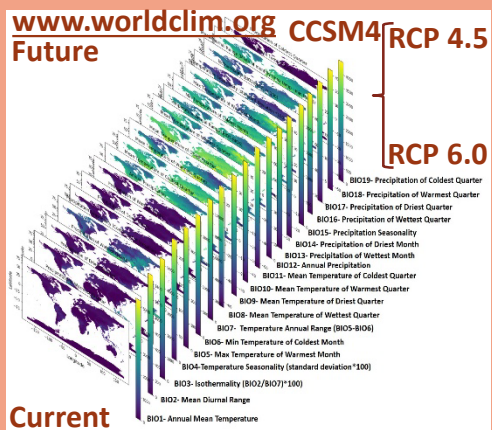


218 IAPs

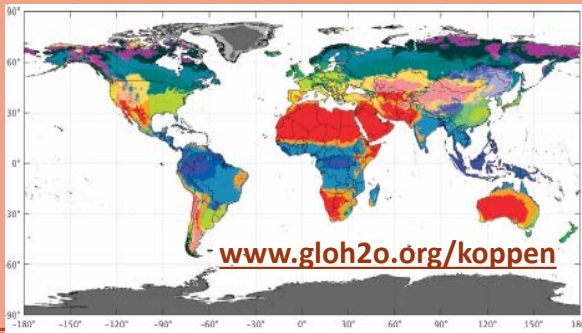
Occurrence collection



Climate information

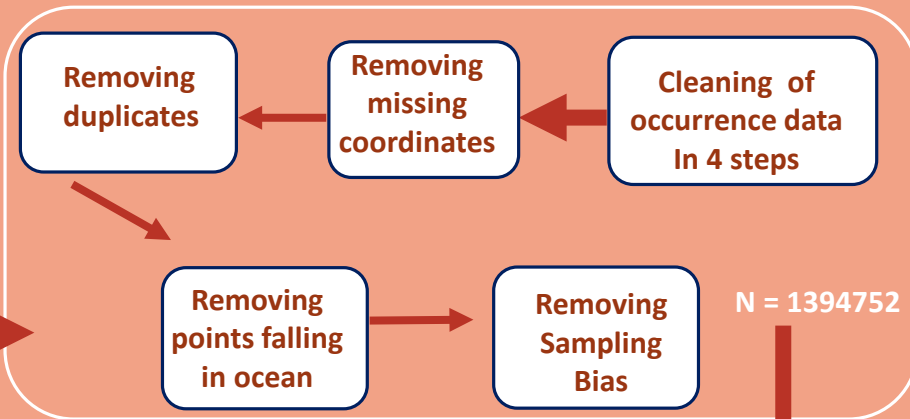


Current

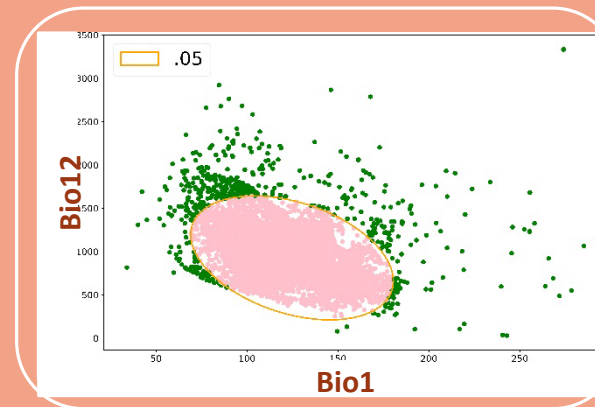


Köppen-Geiger climate Class

Data Cleaning



Delimiting geographic range



Background selection



Data Collection

Data Processing

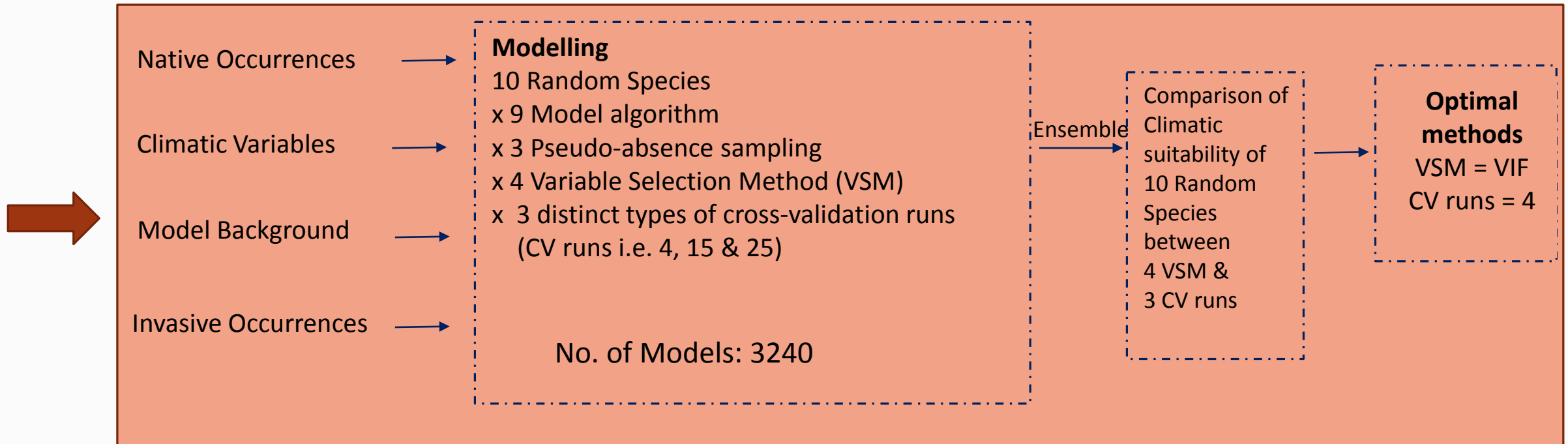
Model Description

Consider, a collection of m independent covariates denoted by $\mathbf{X}_0' = (\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{19})$, where $\mathbf{x}_1 = \text{BIO1}$, $\mathbf{x}_2 = \text{BIO2}, \dots, \mathbf{x}_{19} = \text{BIO19}$. There are total 19 covariates. $m = 19$. $\mathbf{Y} = \mathbf{0}/\mathbf{1}$, where **1** represent presence (occurrences) and **0** represent pseudo-absence

- **Logistic Regression**

- The logistic regression function for estimating probability of occurrence data on given values of environmental factors, $\Pr(\mathbf{Y} = \mathbf{1} | \mathbf{X}) = \pi(\mathbf{X})$ is given by

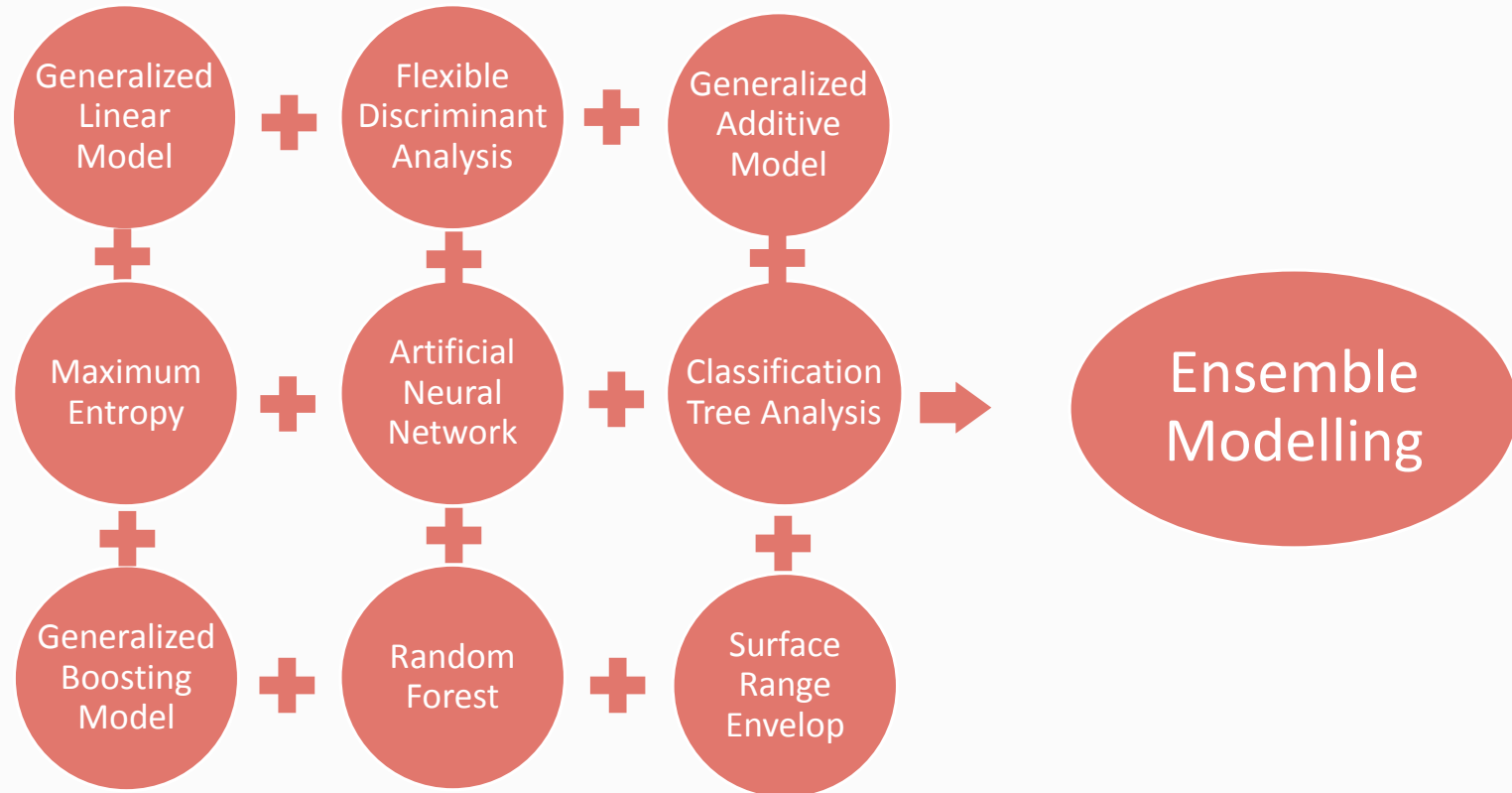
$$\pi(x) = \frac{e^{g(x)}}{1 + e^{g(x)}} \text{ where } g(x) = \beta_0 + \sum_{i=1}^{19} \beta_i X_i, \beta_i' \text{ s are the coefficients (James et al 2013)}$$



Method Selection

Models using Biomod2 package in R

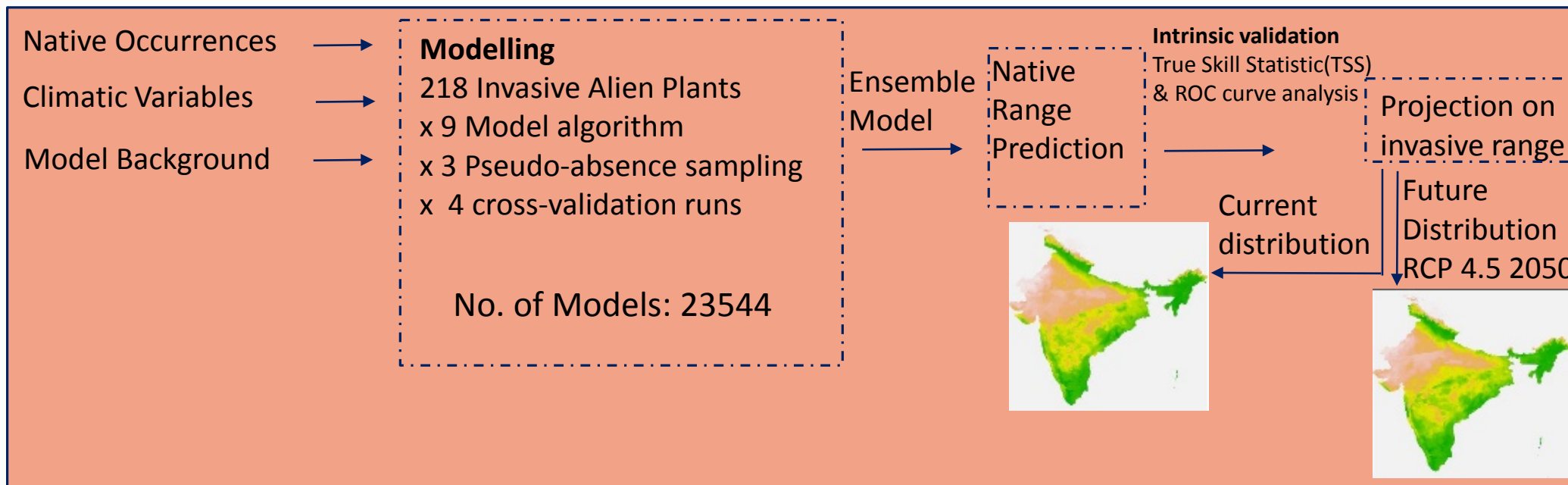
1. Generalized Linear Model (Searle and McCulloch, 2001)
2. Flexible Discriminant Analysis (Hastie et al., 2009)
3. Generalized Additive Model (Guisan et al., 2002)
4. Maximum Entropy (Favretti, 2017)
5. Artificial Neural Network (Zhang, 2010)
6. Generalized Boosting Model (Einzigler et al., 2019)
7. Random Forest (Nordhausen, 2014)
8. Surface Range Envelop (Hannah, 2012)
9. Classification Tree Analysis (Breiman et al., 2017)



Variable Selection method

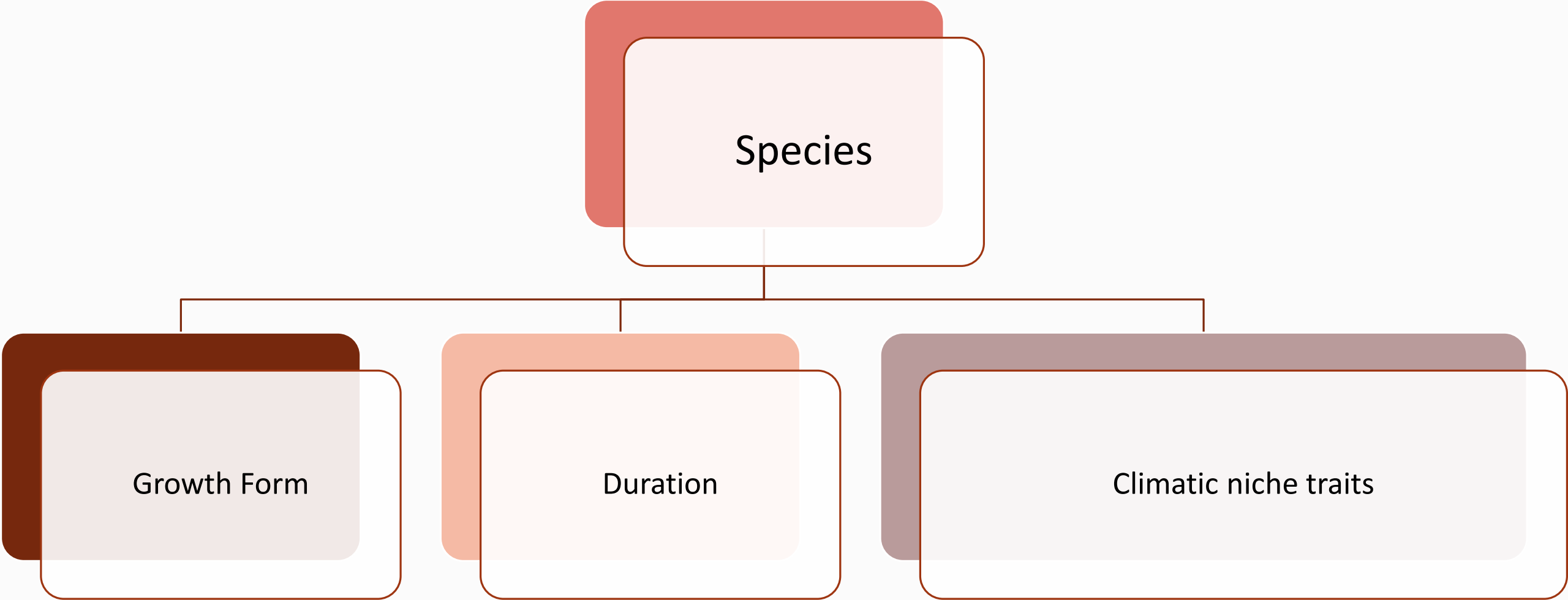
- 1) First four bioclimatic variables from the first two principal with highest loadings in PC1 & PC2 using *PCA package* in R ([Guisan et al., 2017](#)).
- 2) Variables with a correlation value less than 0.7 using *stats package* ([Braunisch et al., 2013](#)).
- 3) Principal components developed globally using bioclimatic variables using the *kuenm package* in R and used first 5 principal components as predictors ([Cobos et al., 2019](#)).
- 4) Variation Inflation Factor with a threshold as 5 using *regclass package* in R ([Mpakairi et al., 2017](#); [Rodríguez-Rey et al., 2019](#)).

VARIATION INFLATION FACTOR
WITH A THRESHOLD AS 5

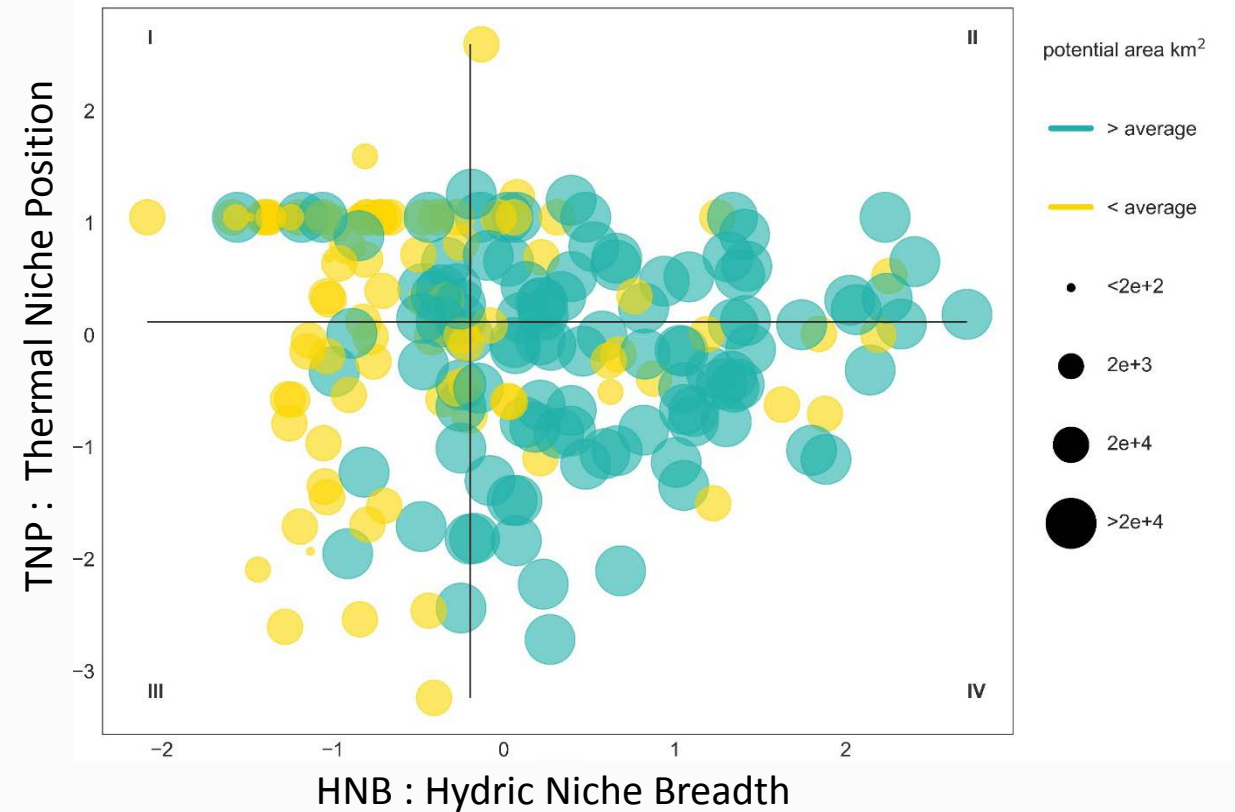


Modelling and Projections

Categorization of Species



Climatic niche traits



$$TNP = \frac{\max(\text{Bio5}) + \min(\text{Bio6})}{2} - \text{India} \left(\frac{\max(\text{Bio5}) + \min(\text{Bio6})}{2} \right)$$

$$HNB = \max(\text{Bio13}) - \min(\text{Bio14})$$

Bio5 : The maximum temperature in the warmest month

Bio6 : the minimum temperature in the coldest month

Bio13 : the maximum precipitation in the wettest month

Bio14 : the minimum precipitation in the driest month

Species

Growth Form

Duration

Quadrants of Climatic niche traits

Annual

Perennial

First

Second

Third

Fourth

Herb

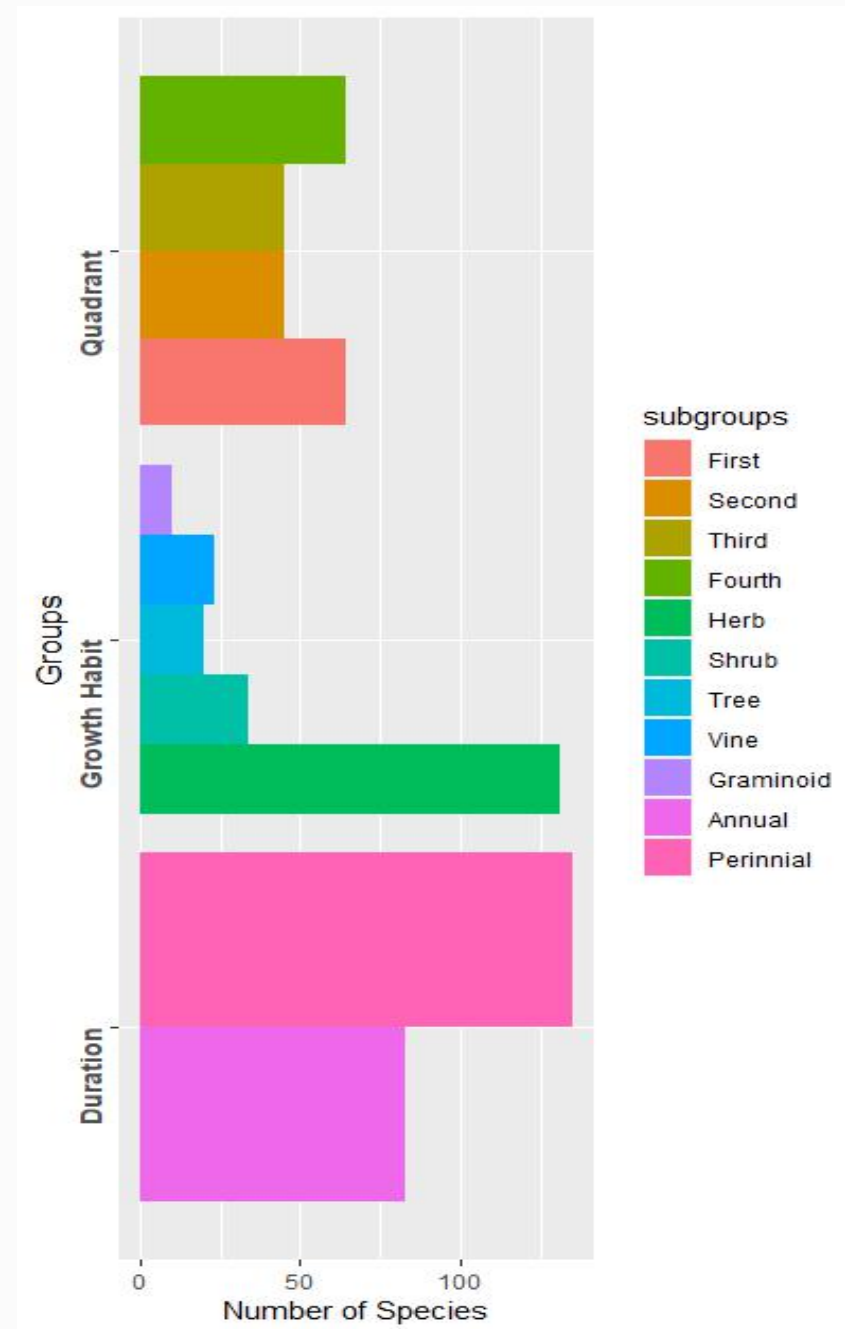
Shrub

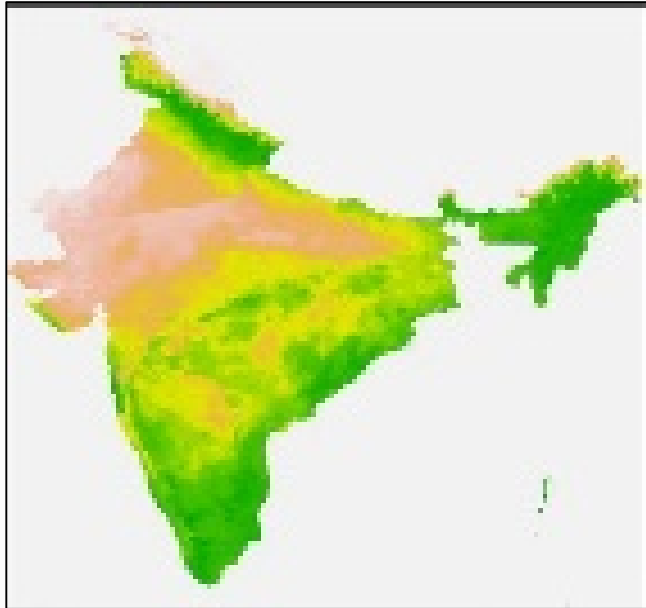
Tree

Vine

Graminoid

Species Distribution in Sub Categories

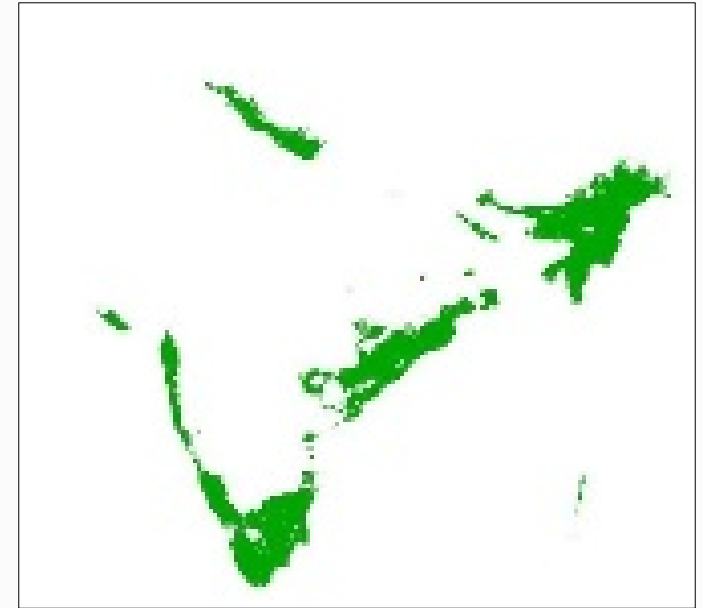


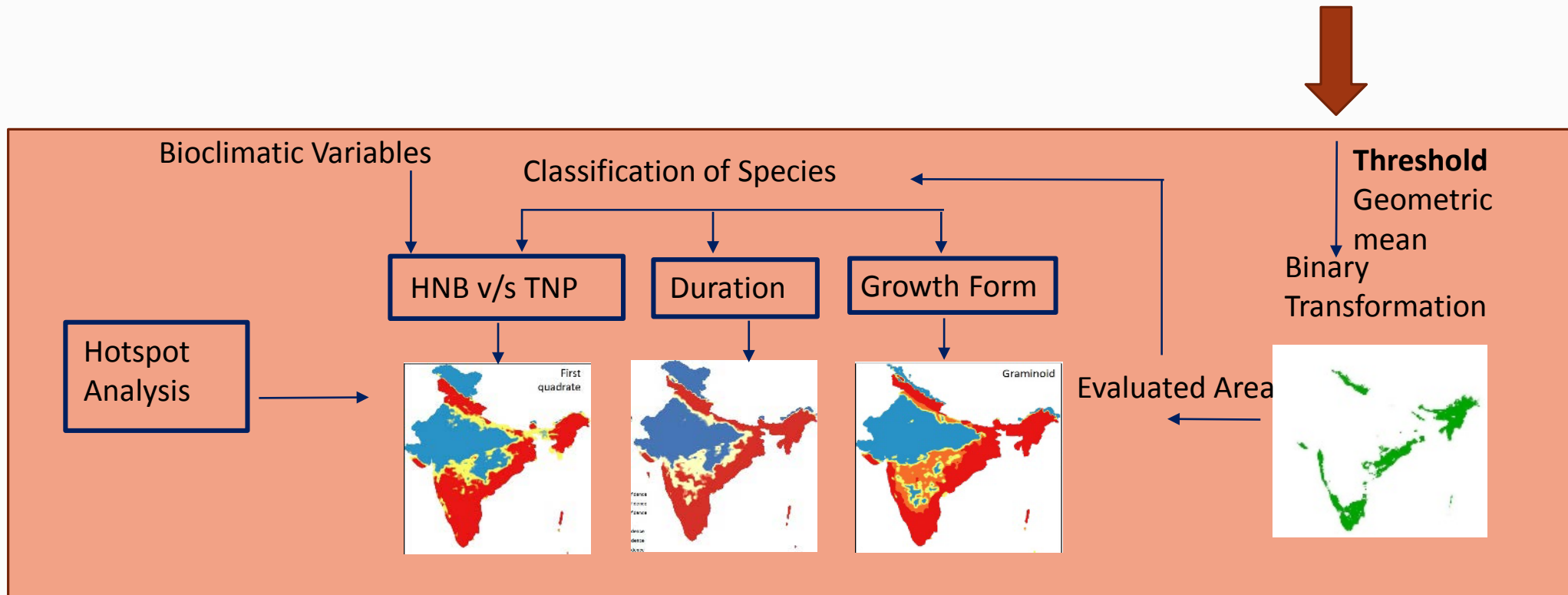


Threshold

$$\text{Geometric Mean} = \sqrt{\text{specificity} * \text{sensitivity}}$$

(Diego et al., 2022; Tharwat, 2020)





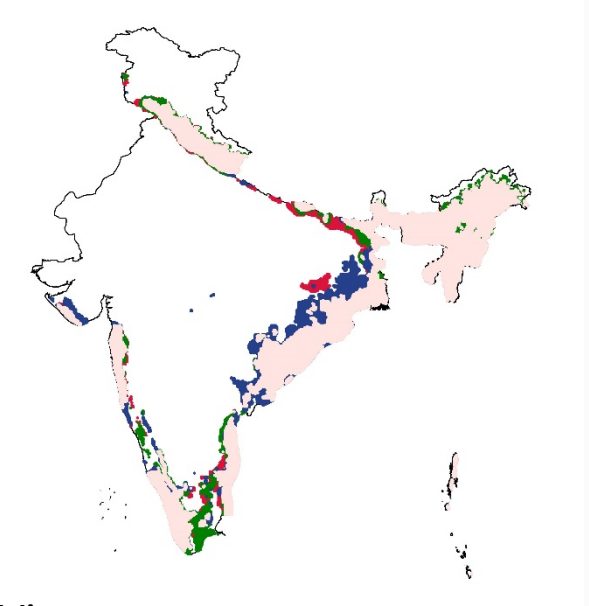
Post-Modelling Analysis

Results

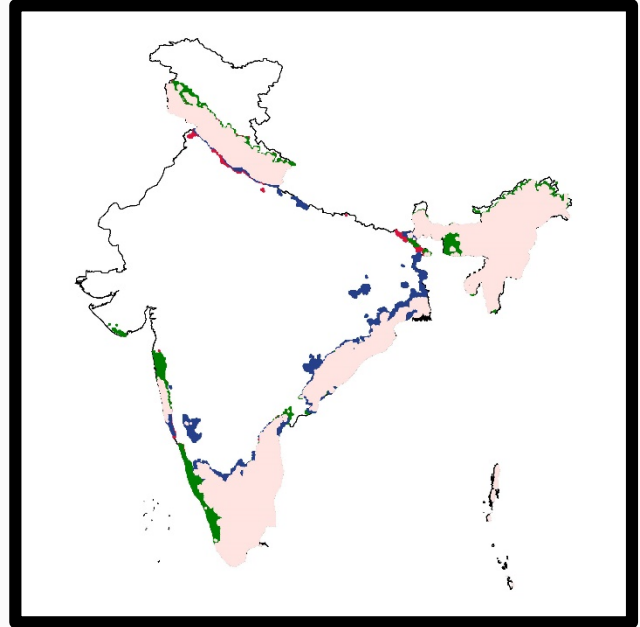
Growth Form : The 99% hotspot for current, 2050 and 2070 with overlapped region of current, 2050 and 2070 outlined as conserved region.



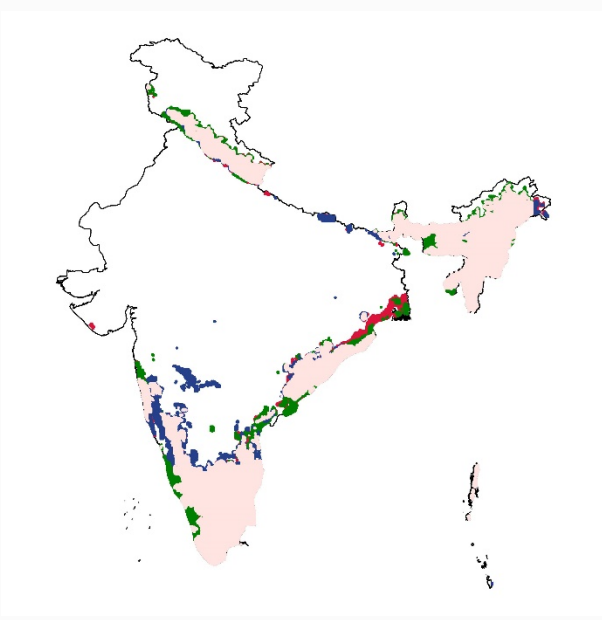
Graminoid



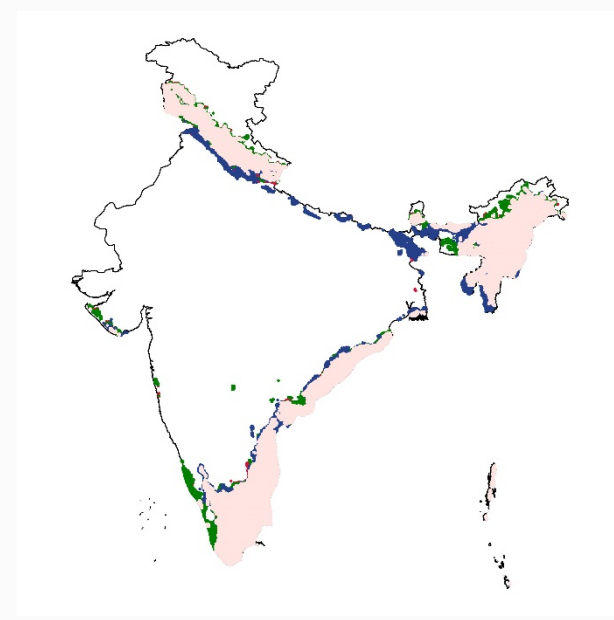
Herb



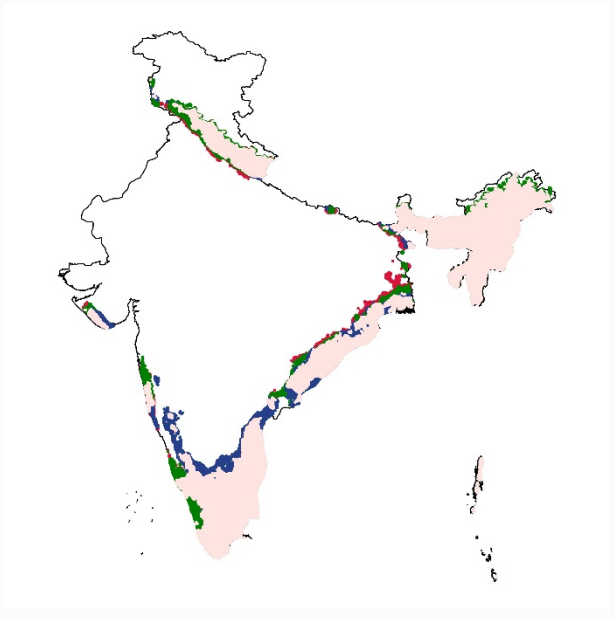
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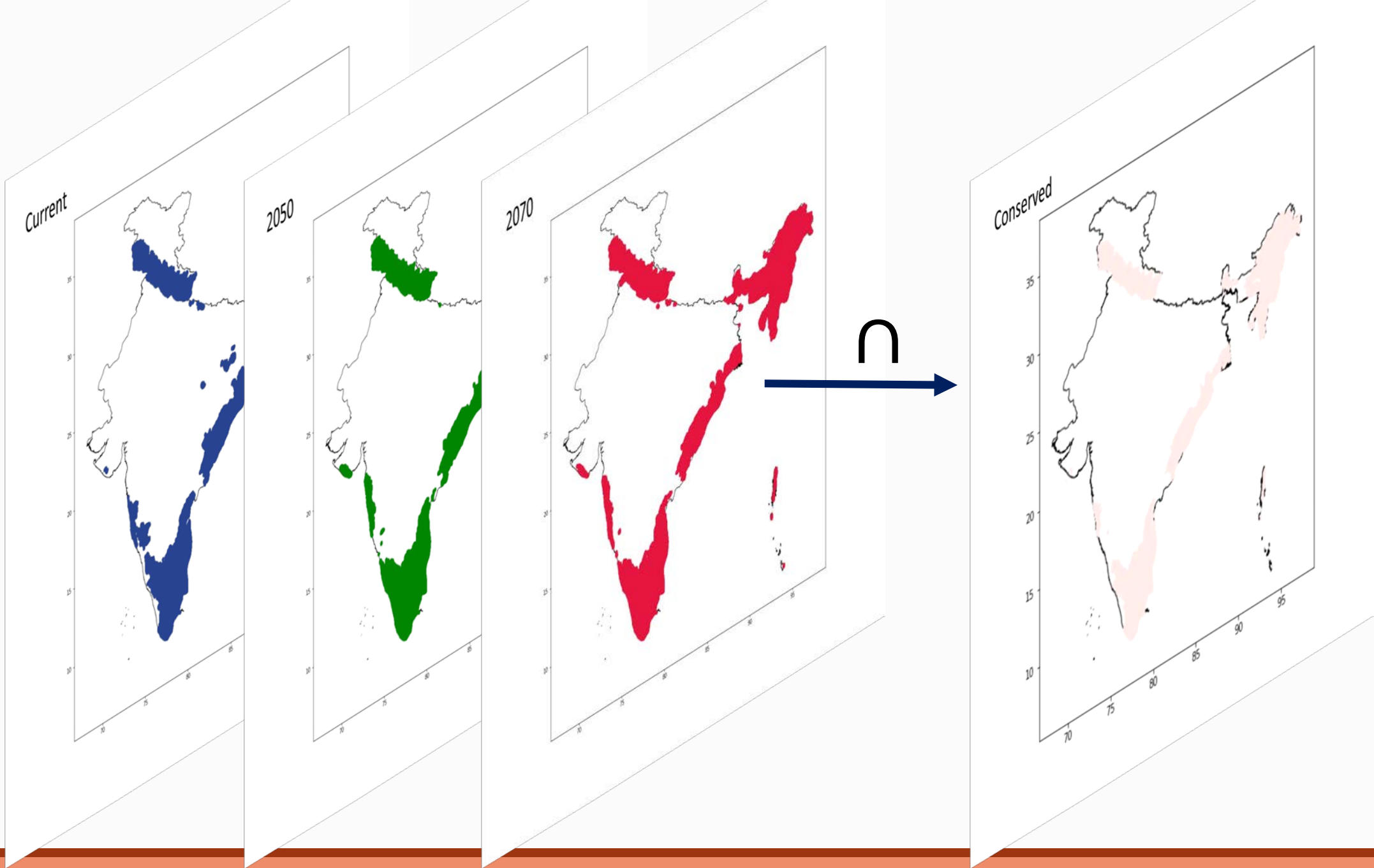


Tree



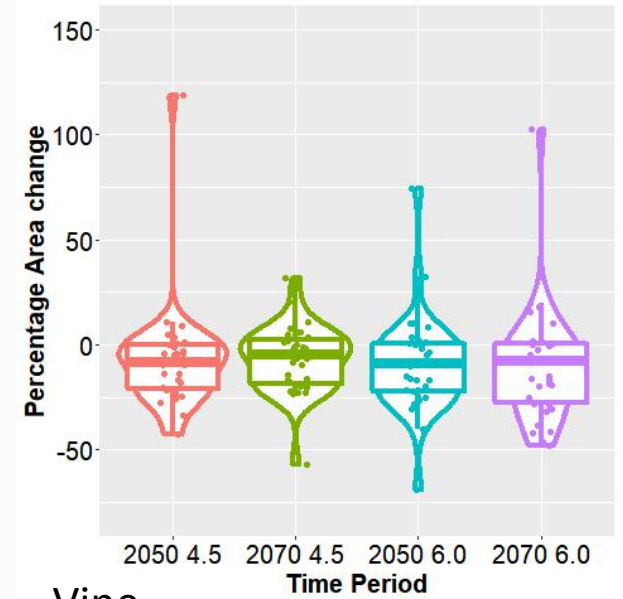
Vine



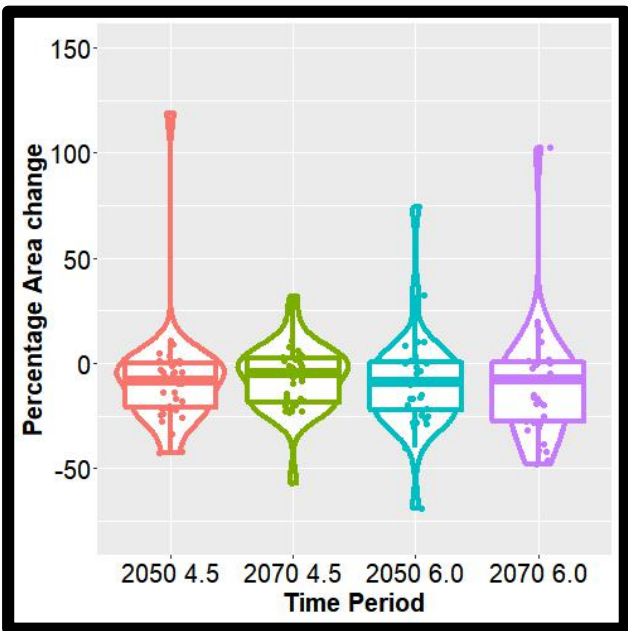


Growth Form : The comparison of overall area changes emerged in Time period 2050 and 2070 at RCP 4.5 and RCP 6.0 respectively.

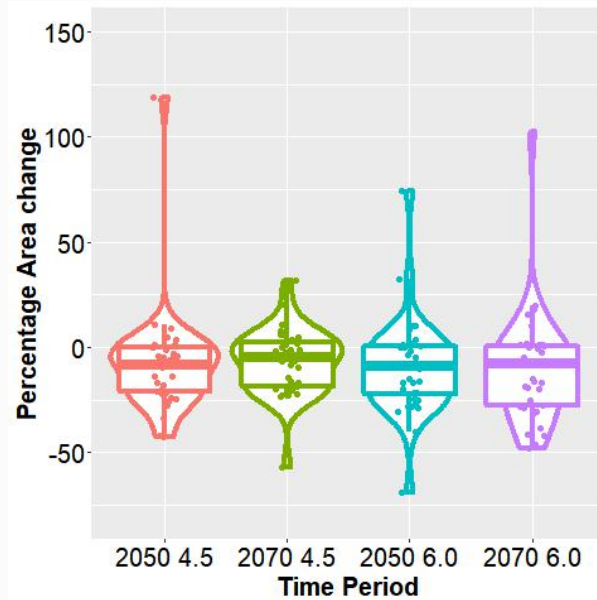
Graminoid



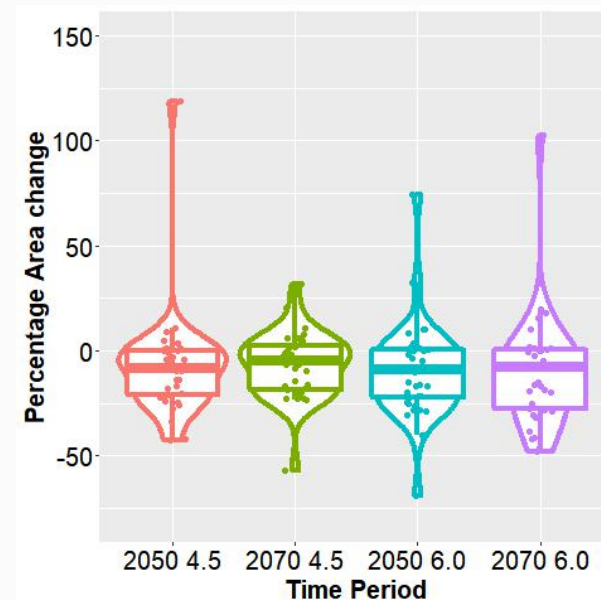
Herb



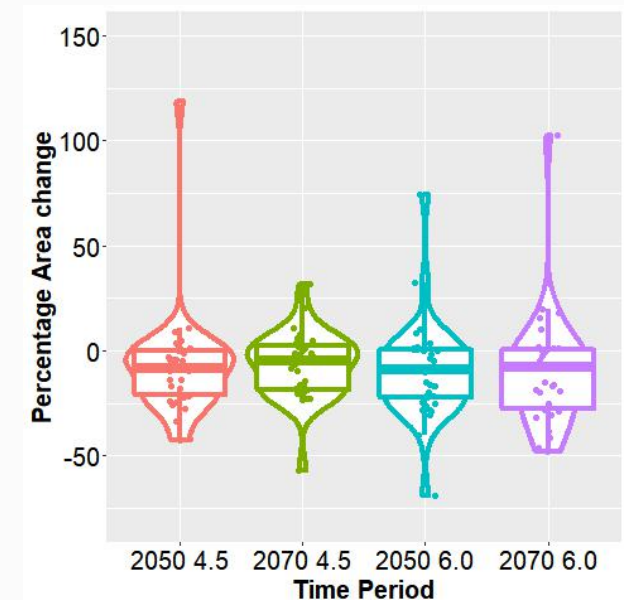
Shrub



Tree



Vine



List of Species showing expansion and contraction

GROUPS	RAPID EXPANSION (RCP4.5)	RAPID CONTRACTION (RCP 4.5)
GRAMINOID	Bromus catharticus Vahl	Pennisetum polystachyon Schult.
HERB	Oxalis pes-caprae L.	Youngia japonica (L.) DC.
SHRUB	Stachytarpheta cayennensis (Rich.) Vahl	Cylindropuntia ramosissima (Engelm.) F.M.Knuth
TREE	Acacia melanoxylon R.Br.	Prosopis juliflora (Sw.) DC.
VINE	Calopogonium mucunoides Desv.	Passiflora edulis Sims
ANNUAL	Acmella radicans (Jacq.) R.K.Jansen	Youngia japonica (L.) DC.
PERENNIAL	Oxalis pes-caprae L.	Cylindropuntia ramosissima (Engelm.) F.M.Knuth
FIRST	Oxalis pes-caprae L.	Cylindropuntia ramosissima (Engelm.) F.M.Knuth
SECOND	Calopogonium mucunoides Desv.	Oxalis latifolia Kunth
THIRD	Acmella radicans (Jacq.) R.K.Jansen	Cirsium arvense (L.) Scop.
FOURTH	Stachytarpheta cayennensis (Rich.) Vahl	Citharexylum spinosum L.

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Thank You !