

# Towards Scalable Deep Species Distribution Modelling Using Global Remote Sensing

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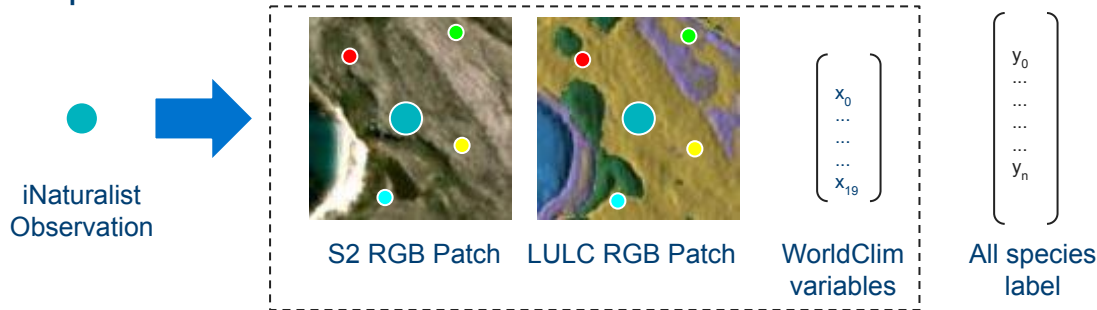
# Introduction

- Destruction of natural habitats and anthropogenic climate change are threatening biodiversity globally.
- Addressing this loss necessitates enhanced monitoring techniques to assess the impact of environmental shifts and to guide policy-making efforts.
- **Aim: to provide an improved, scalable method for species distribution modelling by incorporating global remote sensing data and using this with CNN-based architecture**
- Investigate performance using Protea species in the Cape Floristic Region of South Africa

# Methods & Models

## - Data:

- Adapt dataset pipeline [1] that creates patches centred on each observation. Each patch is labelled with all the species occurring in the patch.
- Data point example:



## - Models:

- *Deepbiosphere* [1], a multimodal version of a T-ResNet CNN model
- Maxent, classic statistical model used for species distribution modelling

# Training splits

- Spatial autocorrelation between evaluation and training sets can cause overly optimistic results.
- Create representative train/validation splits while addressing spatial autocorrelation

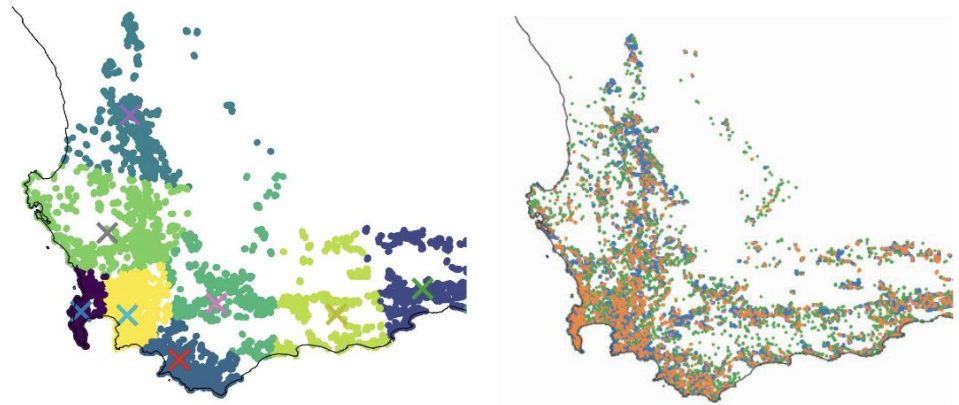


Figure 3: Example clusters for part of the Cape Floristic Region used in creating the validation sets (*left*) and separation of points into train (blue), validation (orange) and test (blue) sets for the region (*right*).

# Results & Conclusions

- The CNN models perform better on average for  $AUC_{ROC}$ :
  - Image + environmental mean  $AUC_{ROC}$ :  **$0.8849 \pm 0.046$**
  - Image + LULC + environmental mean  $AUC_{ROC}$ :  **$0.8822 \pm 0.048$**
  - Maxent mean  $AUC_{ROC}$ :  **$0.8550 \pm 0.059$**
- Overall we show that we can train a model to make predictions based on global remote sensing data that has fine-grained spatial and temporal scales.
- Results need to be further investigated using a field study and ground truth data.

[1] Gillespie, L., Ruffley, M. and Expósito-Alonso, M., 2021. An Image is Worth a Thousand Species: Scaling high-resolution plant biodiversity prediction to biome-level using citizen science data and remote sensing imagery. *Biodiversity Information Science and Standards*, 5, p.e74052.