

Using DNA barcoding and machine learning to investigate macroecological relationships between plants and arthropods across South Africa.

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Relevance

- What is climate change doing?
 - Baseline data
 - Is it altering ecological relationships?
 - timing shift and range shifting
- Challenges
 - There are millions of species!
 - Lack base line data on insects
 - Lack baseline data with the flowering times



Insect Diversity Overview

- Described insect diversity.
 - About 1 million.
 - Estimates range between 8-100 million.

(Mora et al., 2011)

- Challenges with describing insect biodiversity.
 - Time consuming | taxonomic expertise | funding.



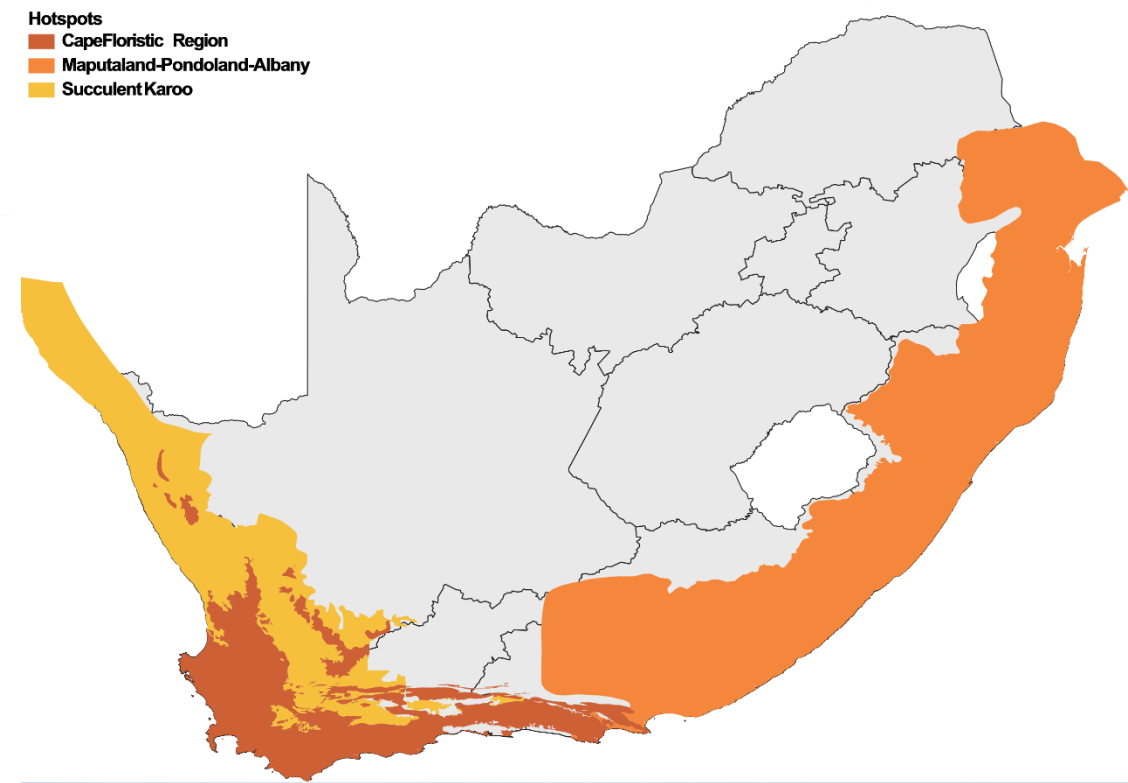
Importance of Insects

- Important ecosystem roles
 - The Good – Pollination.
 - The Bad – Plant pests.
 - The Ugly – Vectors for pathogens (Polyphagous Shot Hole Borer beetle – PSHB).
- Key ecological indicators for climate change
 - Species richness | abundance | predator-prey ratios | sensitive species | specialists vs. generalists.
- Insect apocalypse (Hallmann et al., 2017)



South African Biodiversity

- Mega biodiversity, estimated
 - 21,000 plant species.
 - 44,000 insect species in Southern Africa.
- Three biodiversity hotspots
 - Cape Floristic Region – 6,210 endemic plants.
 - Succulent Karoo – 2,542 endemic plants.
 - Maputaland-Pondoland-Albany – 1,900 endemic plants.



Insects in South Africa

The last study done on insects was done in southern Africa in:

- 2016 – Scholtz
- 1995 – Scholtz & Chown
- 1985 – Scholtz & Holm

Nearly 40 years!

44,000 species, 7750 genera, 569 families, and 25 orders

New ways to update this baseline data





Flowering Phenology

~Represents the timing of biological events.

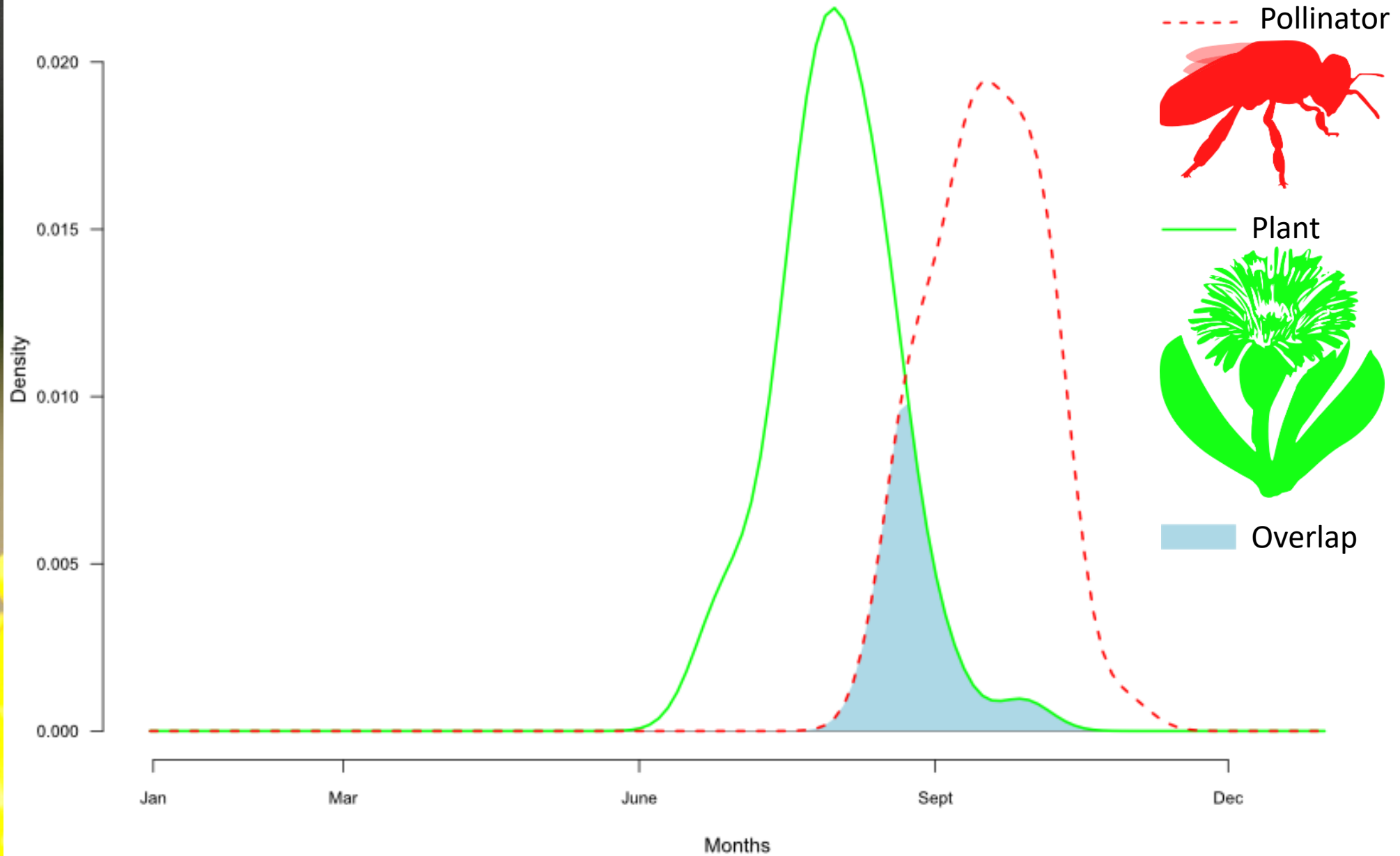
- Flowering time
 - Crucial part of a plant's life cycles
- UN Sustainable Development Goals
- Fingerprint to climate change (Fei et al., 2017)
- Ecological interactions
 - Pollination
 - Insect Armageddon (Hallmann et al., 2017)

SUSTAINABLE DEVELOPMENT GOALS




10 Years from now


Overlapping phenology



--- Pollinator



— Plant



■ Overlap

Insects in South Africa

Prosoeca marinusii



Babiana framesii





Phenology

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Plants and Insects

- Plant community defines the niche space for insects – expect insect species richness to reflect the plant species richness.
- Large scale plant diversity is well-studied
 - Climate | soil properties | species interactions.
- Lack of macroecological studies for insects.
- What will happen to the plant-insect interactions in SA?



Aims and Objectives

To investigate macroecological relationships between plants and arthropods across South Africa.

- Generate baseline data on insect diversity in South Africa and explore the diversity across different biomes.
- Find the flowering period of all the plants in different parts of South Africa.
- Investigate the relationship between plants and insect diversity.

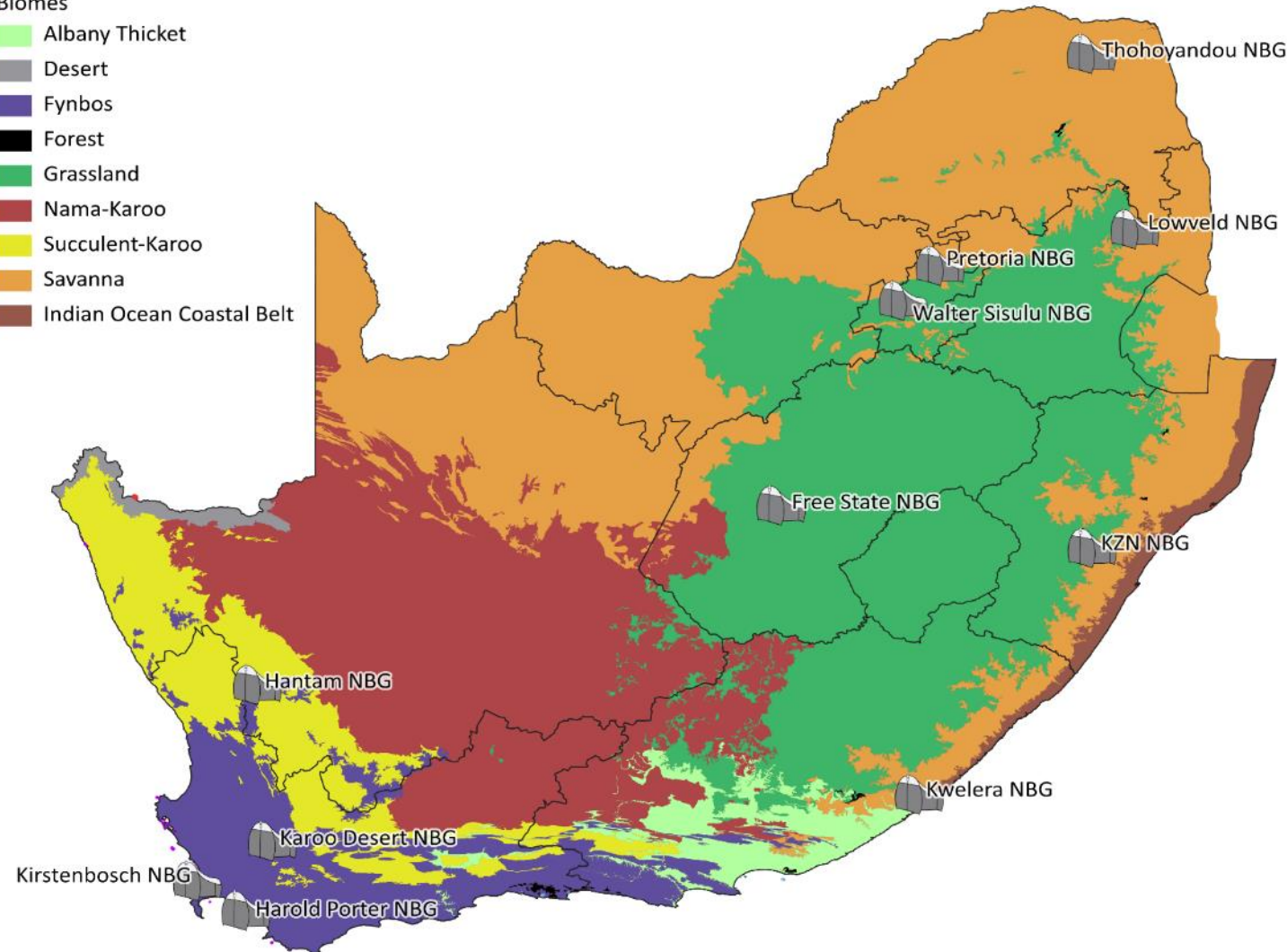


Methods – Insect Composition



Biomes

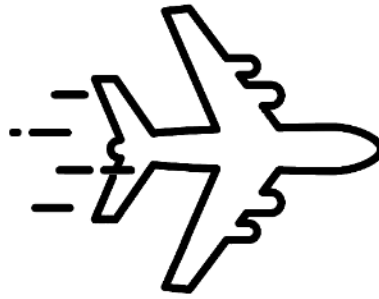
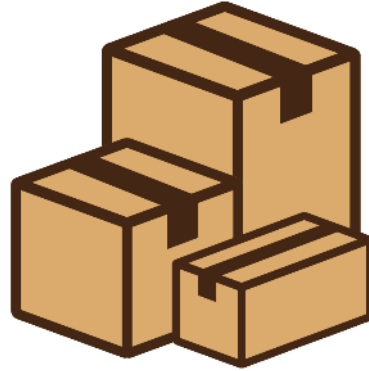
- Albany Thicket
- Desert
- Fynbos
- Forest
- Grassland
- Nama-Karoo
- Succulent-Karoo
- Savanna
- Indian Ocean Coastal Belt



Methods – Insect Composition



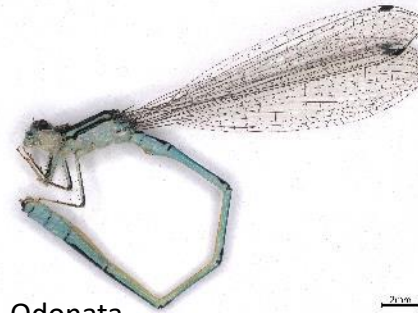
Methods – Insect Composition



Amphipoda



Pseudoscorpiones



Odonata



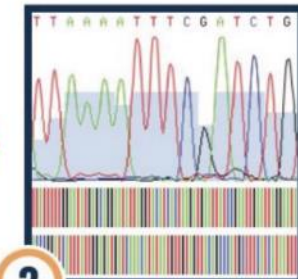
Araneae



1



2



3



4

DNA Barcoding

- It is a molecular technique that uses a **short, standardised region** of DNA (*COI*) to identify and classify species.
- It is great because it can distinguish between species when you **don't have morphological characters** or a species ID.
- It is widely used in biodiversity research, **conservation**, and **ecological monitoring**
- **“Dark Taxa”** – Species with barcode sequence data but **missing formal taxonomic classification**.





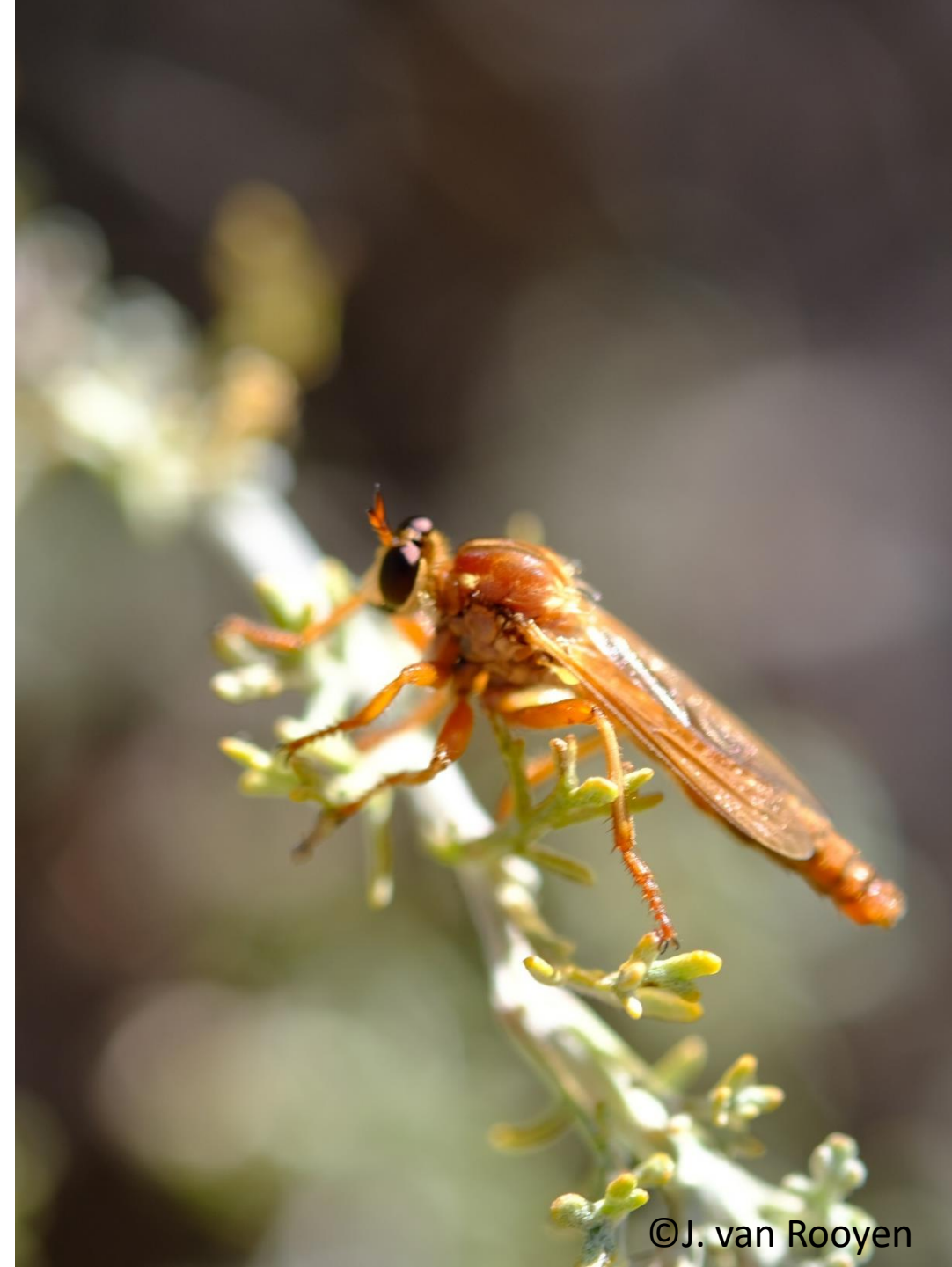
Informatics and phylogenetics

All informatics were performed in R using several packages, including:

- Species estimates, Mantel tests, phylogenetic rarefaction curves, assessed the β diversity and investigated the phylogenetic turnover.
- Performed phylogenetics analyses on insects (29,919 unique species proxies) and Plants (ca. 14,000 unique species).

Results: The Numbers

- 22 Traps
- 337,809 Specimens
- 29,919 unique BINs
- 37 unique orders
- 522 unique families
- 40,994 specimens to genus
- 18,251 specimens to species

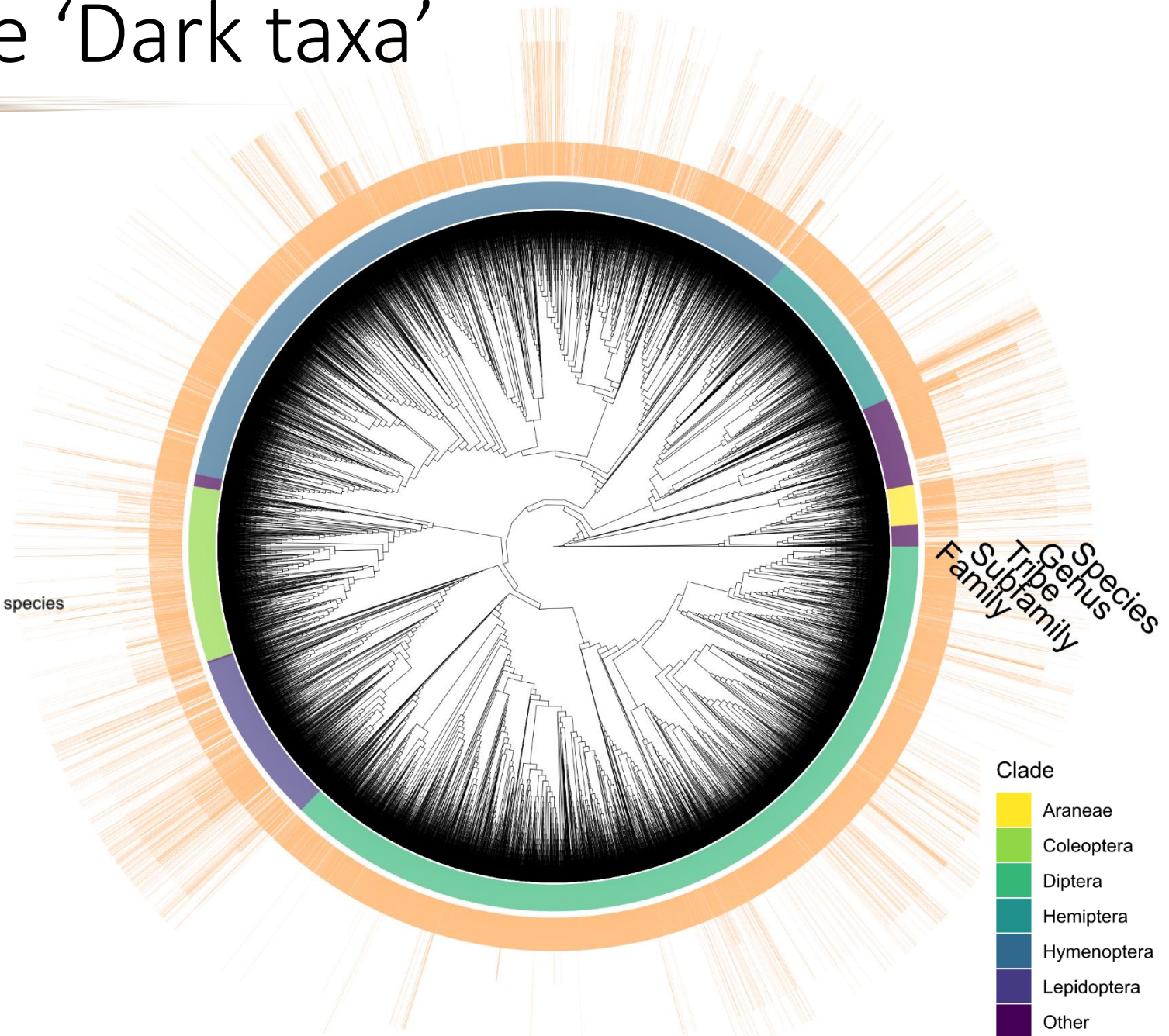




Results: Where are the 'Dark taxa'



to species



Take Home Message

- Plant species richness doesn't really predict the insect species richness.
- 'Dark taxa' are growing rapidly in South Africa – we'll need solutions.
- We have a lot more to learn about insects than we thought!



Aims and Objectives

To investigate macroecological relationships between plants and arthropods across South Africa.

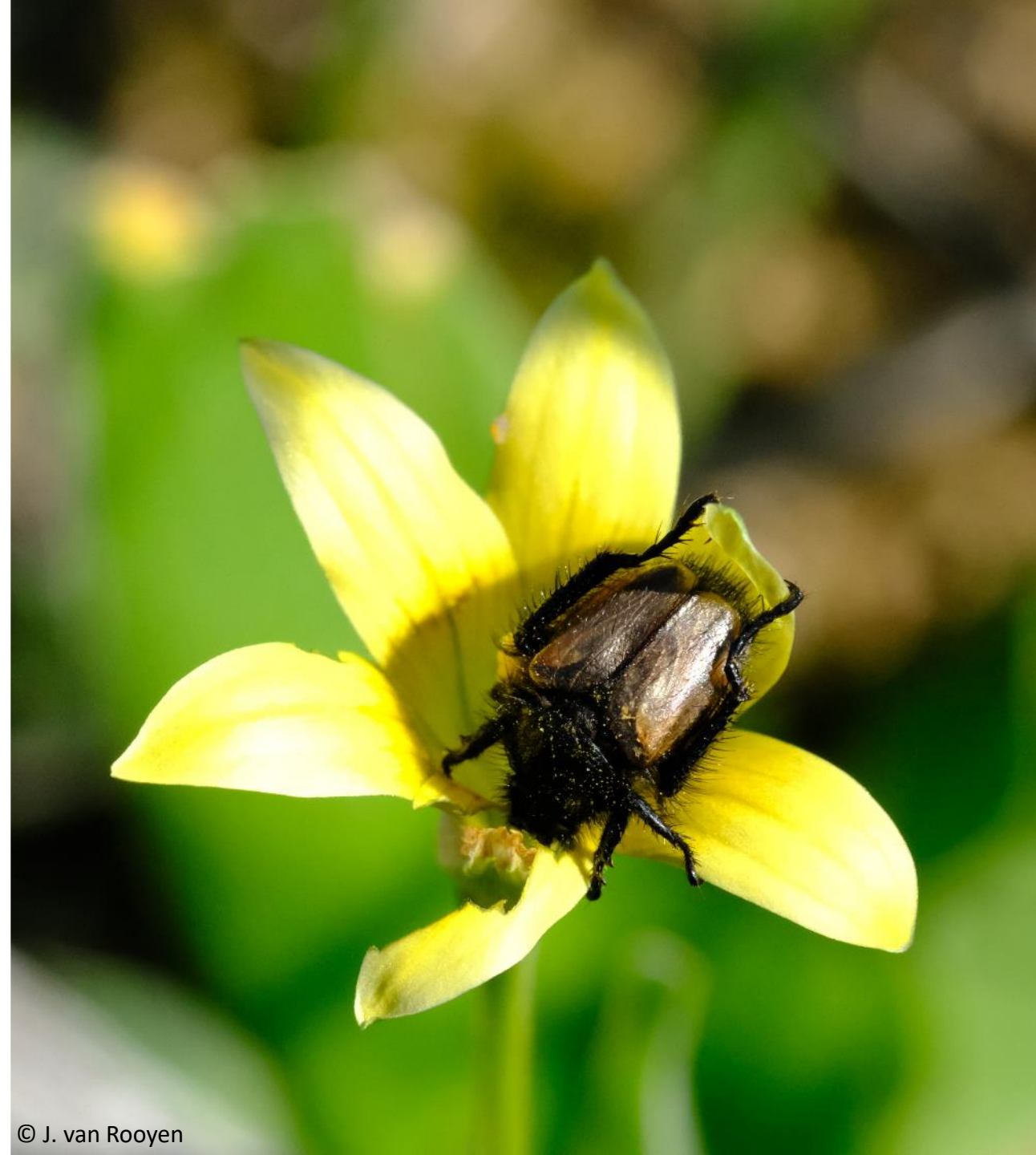
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South Africa

~Is the third most biodiverse country in the world – Making it one of the richest countries in the world

- Ca. 21,000 Plants. (Zengeya et al., 2020)
- Not enough taxonomists.
- Phenological data gap.
- Need to accelerate data collection.



Machine learning

“In the age of big data, machine learning is the key to unlocking the secrets hidden within the numbers.”

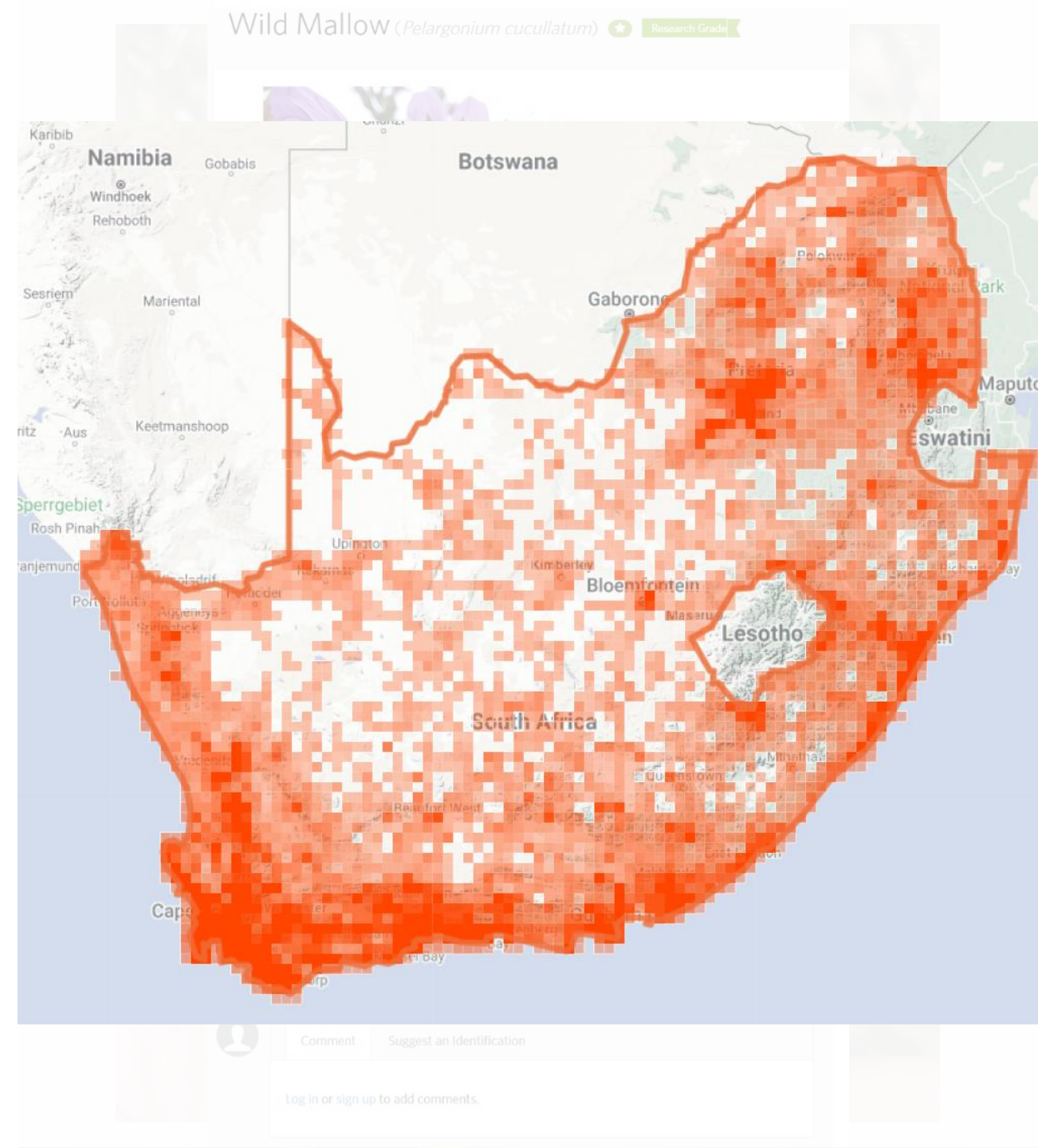
- Hal Varian

- Manual image categorization can be time-consuming.
- Convolutional neural networks (CNNs).
- Has been used to classify herbarium specimen for single species.



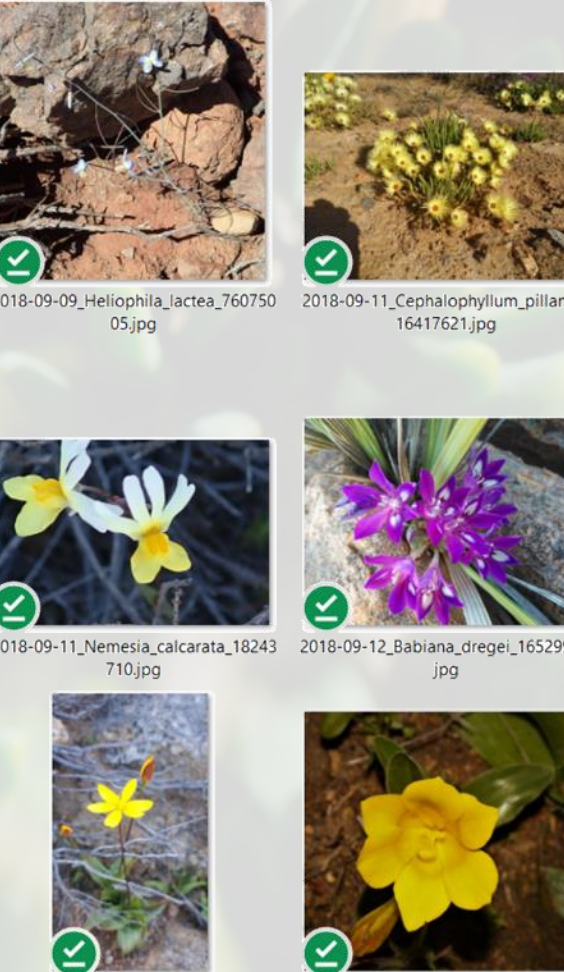


- A community that takes pictures of species for identification.
 - Non-standardized images.
- What is a Research Grade observation?
 - Suggested ID (ML identification), two more suggested ID's (confirmation).
- 1,797,903 research grade observations in SA.



Training Data – Primary Model

5000



2018-09-09_Heliophila_lactea_76075005.jpg

2018-09-11_Cephalophyllum_pillansii_16417621.jpg


2018-09-11_Nemesia_calcarata_18243710.jpg

2018-09-12_Babiana_dregei_1652996.jpg

2018-09-12_Pauridia_gracilipes_16894.jpg

2018-09-13_Moraea_kamiesensis_21111111.jpg

Flowering

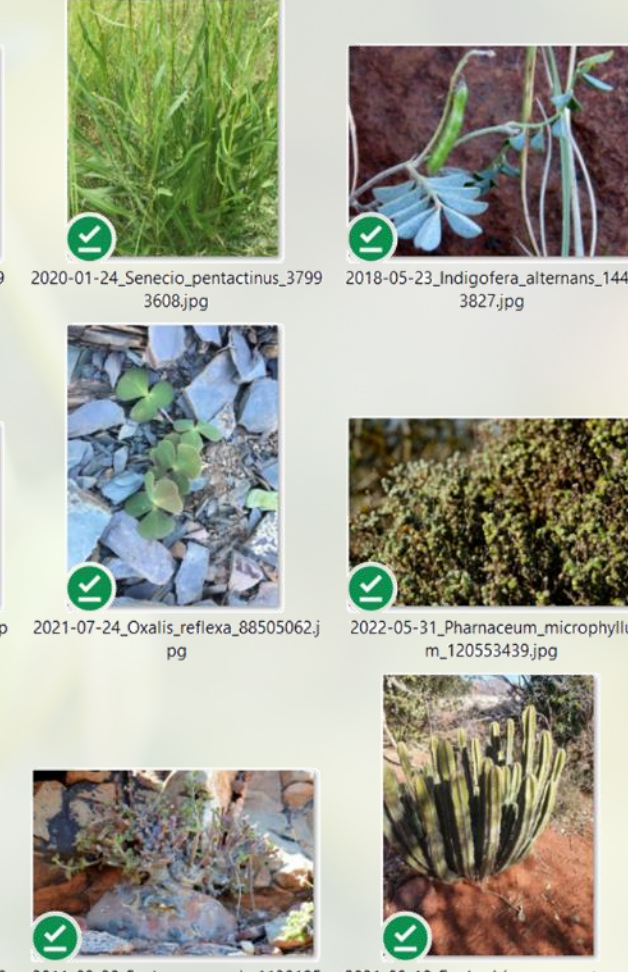


2018-09-11_Suaresia_paniculata_16417621.jpg

2018-09-12_Cleistanthus_radians_1652996.jpg

2018-09-14_Ambrosia_paniculata_16894.jpg

5000 Not Flowering



2020-01-24_Senecio_pentactinus_37993608.jpg

2018-05-23_Indigofera_alternans_14443827.jpg

2021-07-24_Oxalis_reflexa_88505062.jpg

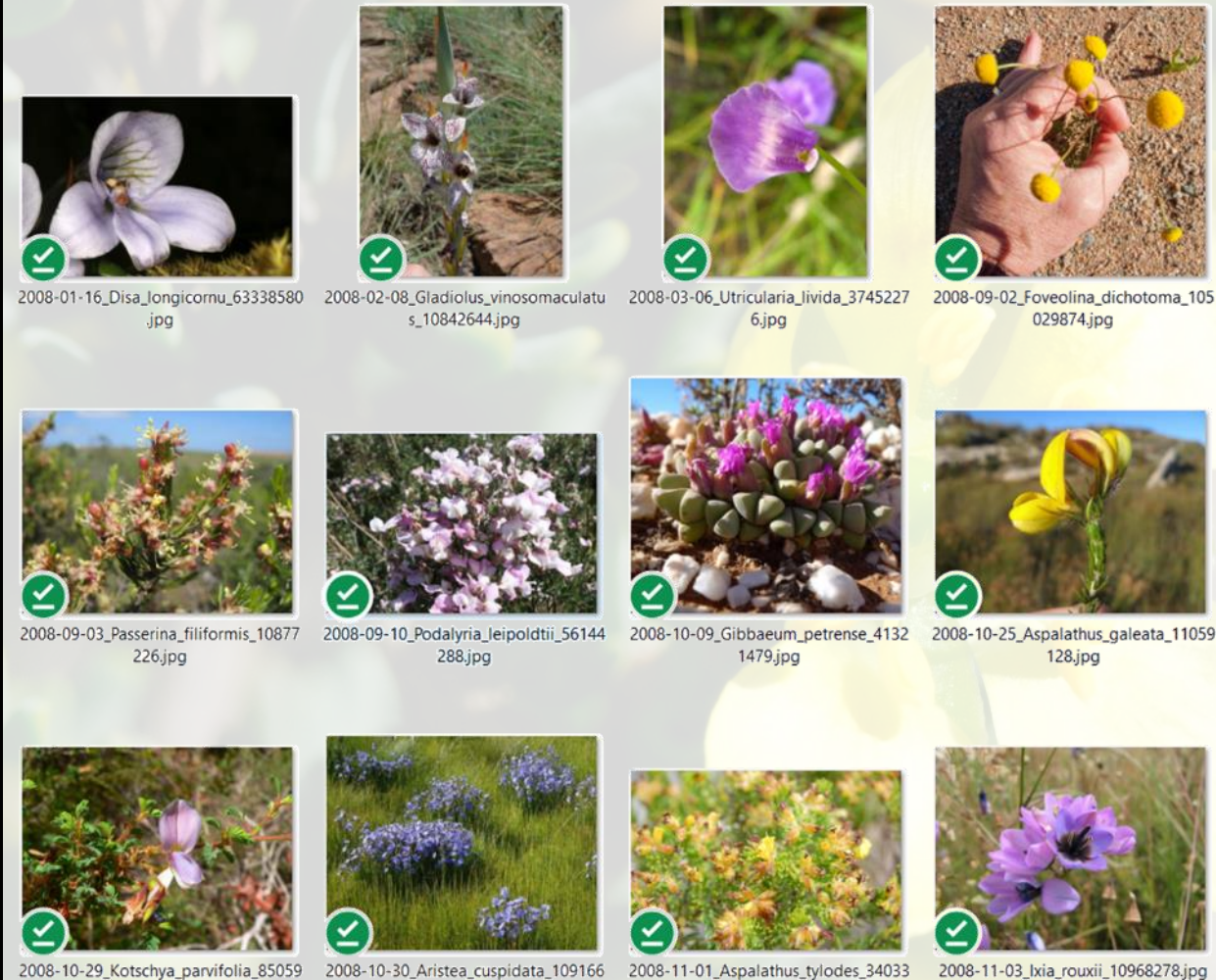
2022-05-31_Pharnaceum_microphyllum_120553439.jpg

2011-09-30_Fockea_capensis_11301956.jpg

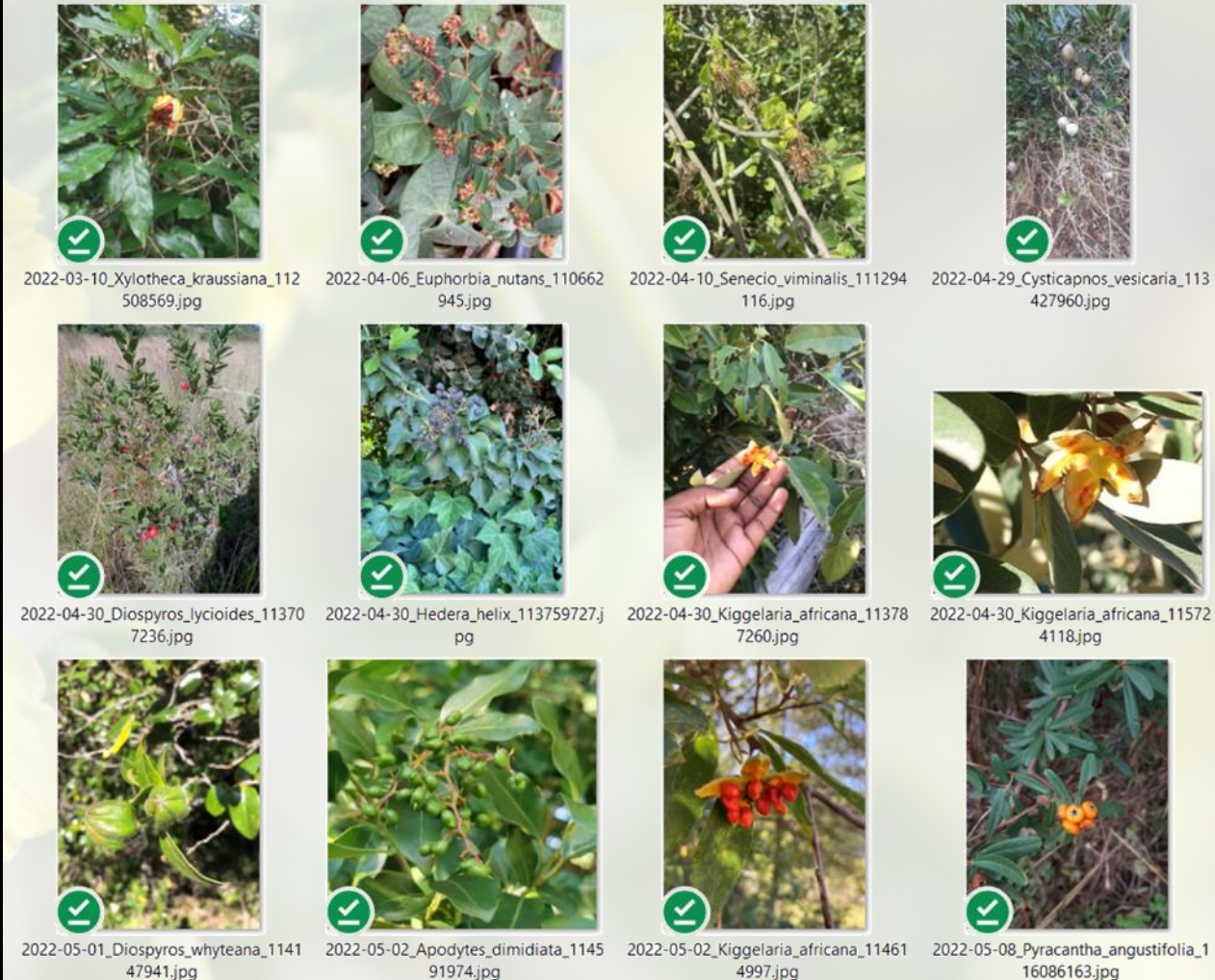
2018-08-10_Euphorbia_avasmontana_90872535.jpg

Training Data – Secondary Model

1000 Flowering



168 Fruiting

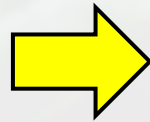




Determining the Phenology



Download Images
(1,8 million)



For the code scan below:

Test Primary Model
(Flowering)

Test Secondary Model
(Fruiting)

<https://github.com/rossdstewart/ML-Phenology-Code>

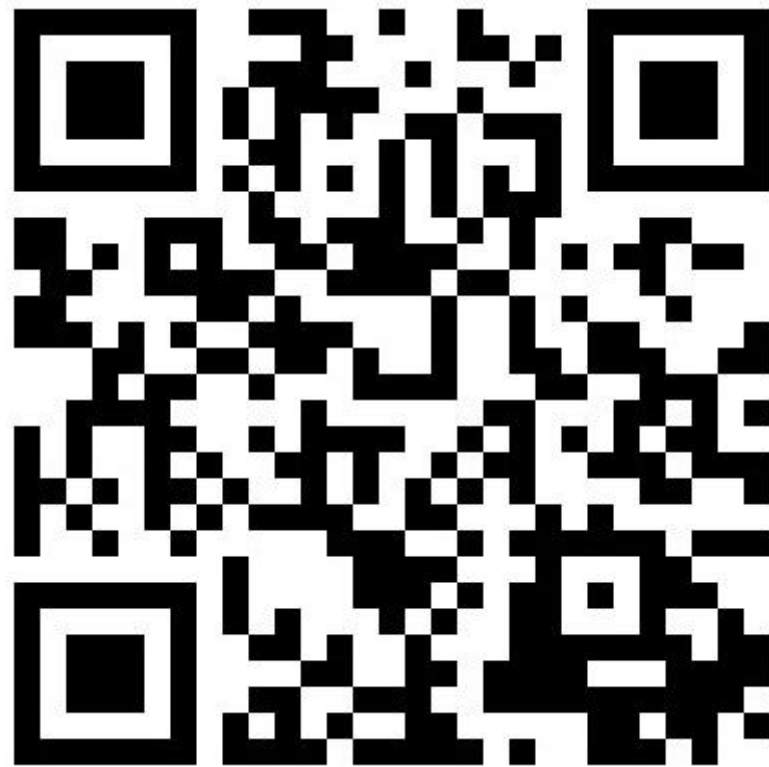
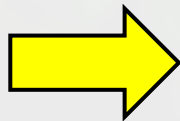


Image categorization

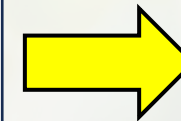


Extract the date
from each image.



Keep d
confirm

Extract the
Phenological pattern.



Abrus_precatorius



Example: *Abrus precatorius*

F – Flowering; NF – Not-Flowering; F2 – Flowering 2; Fu - Fruiting



F=54.47%; NF=45.53%
F2=14.85%; Fu=85.15%



F=12.19%; NF=87.81%



F=13.82%; NF=86.18%



F=2.48%; NF=97.52%



F=64%; NF=36%
F2=99.5%; Fu=0.5%



F=0.4%; NF=99.6%



F=86.19%; NF=13.81%
F2=99.95%; Fu=0.05%

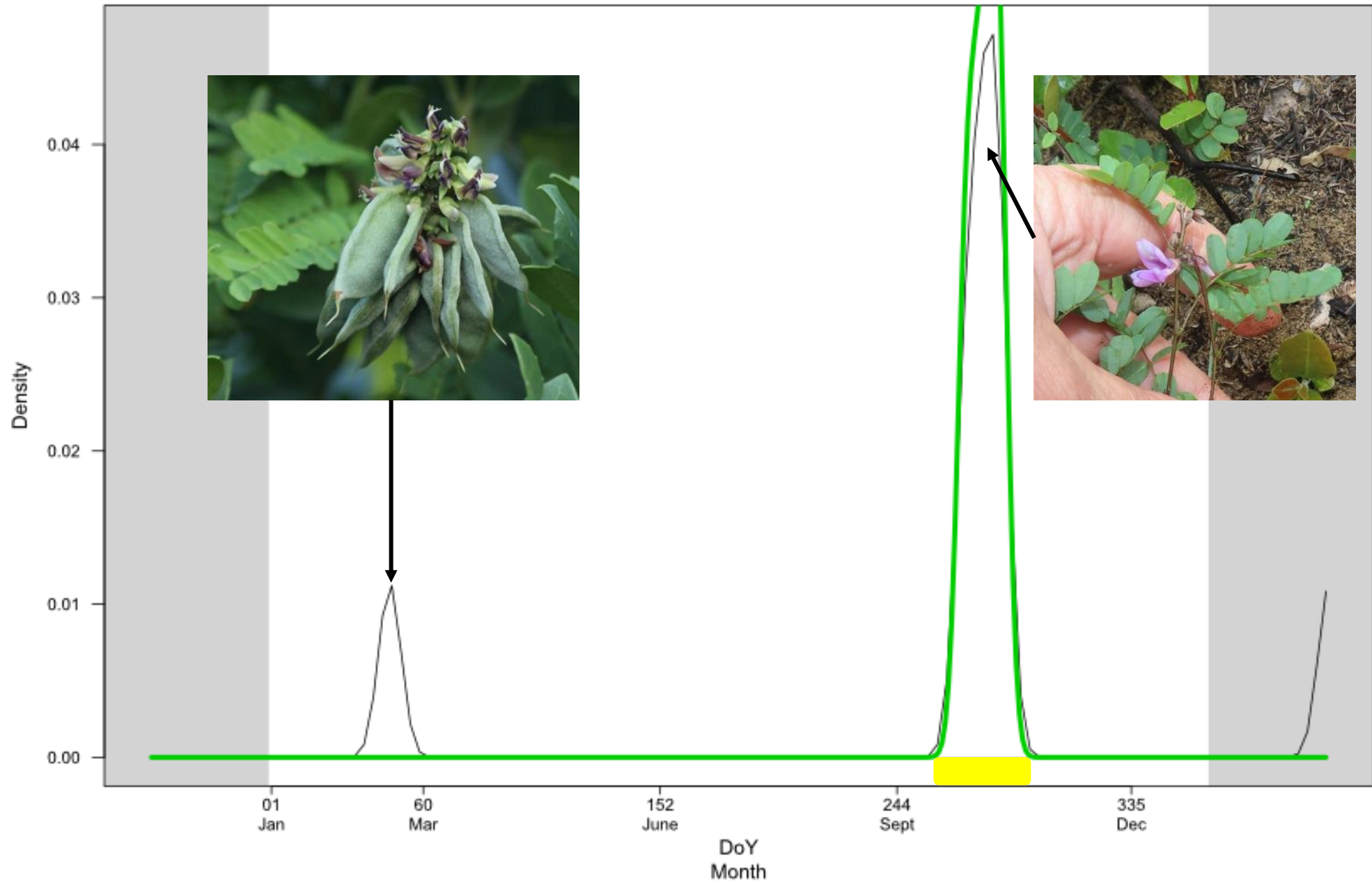


F=29.53%; NF=70.47%



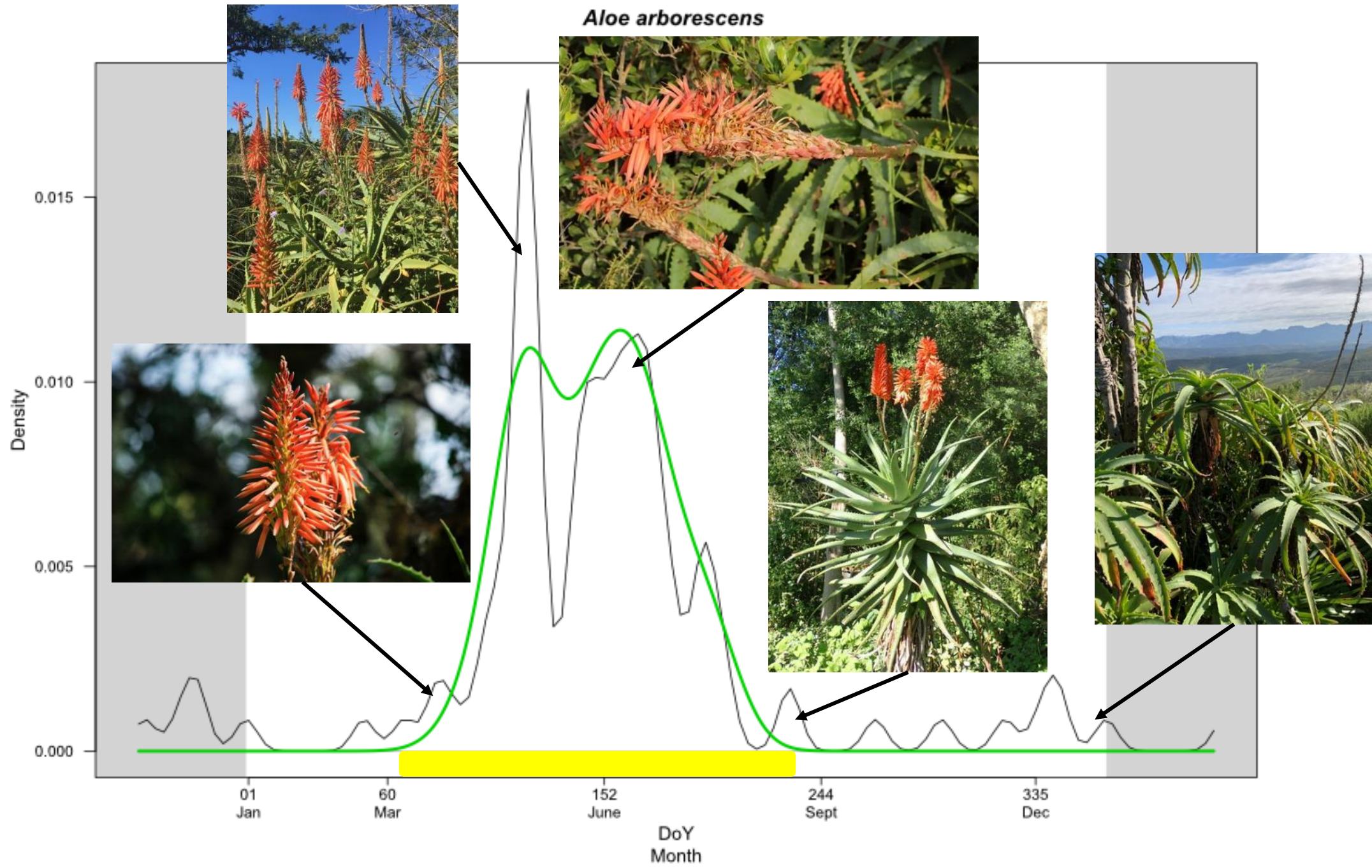
F=98.11%; NF=1.89%
F2=95.89%; Fu=4.11%

Abrus precatorius



n=7

Aloe arborescens



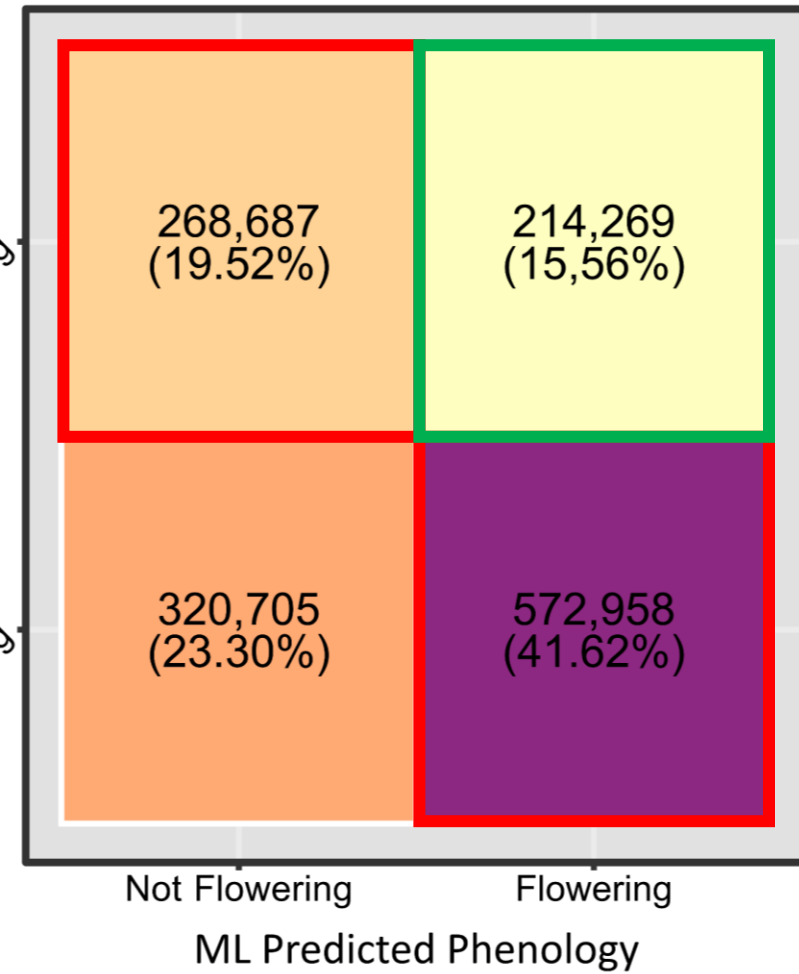


(b)

Expert Reported Phenology

True Not Flowering

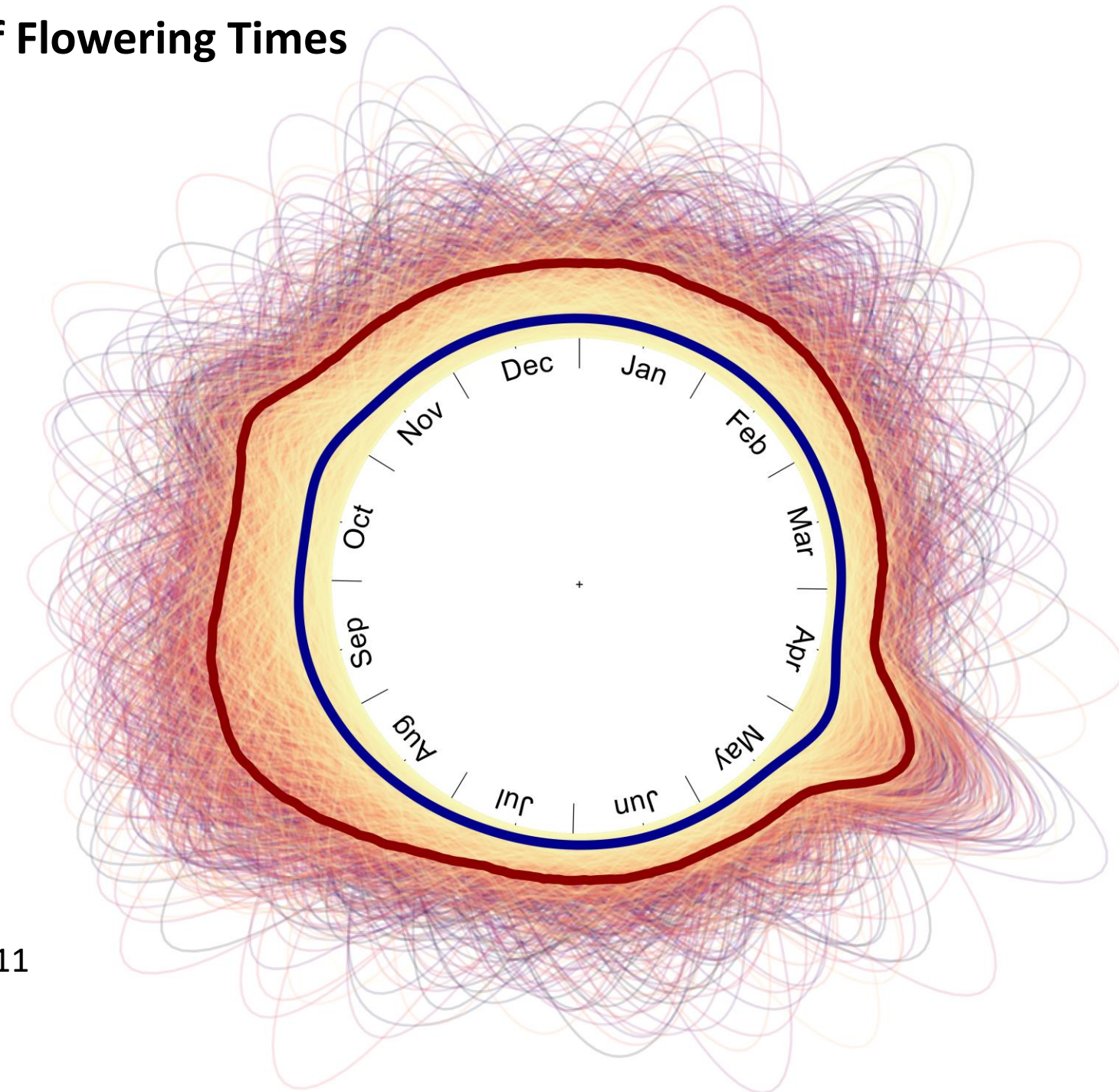
True Flowering





(a) – Accuracy of two-step prediction method

(b) – Accuracy of predicted phenologies

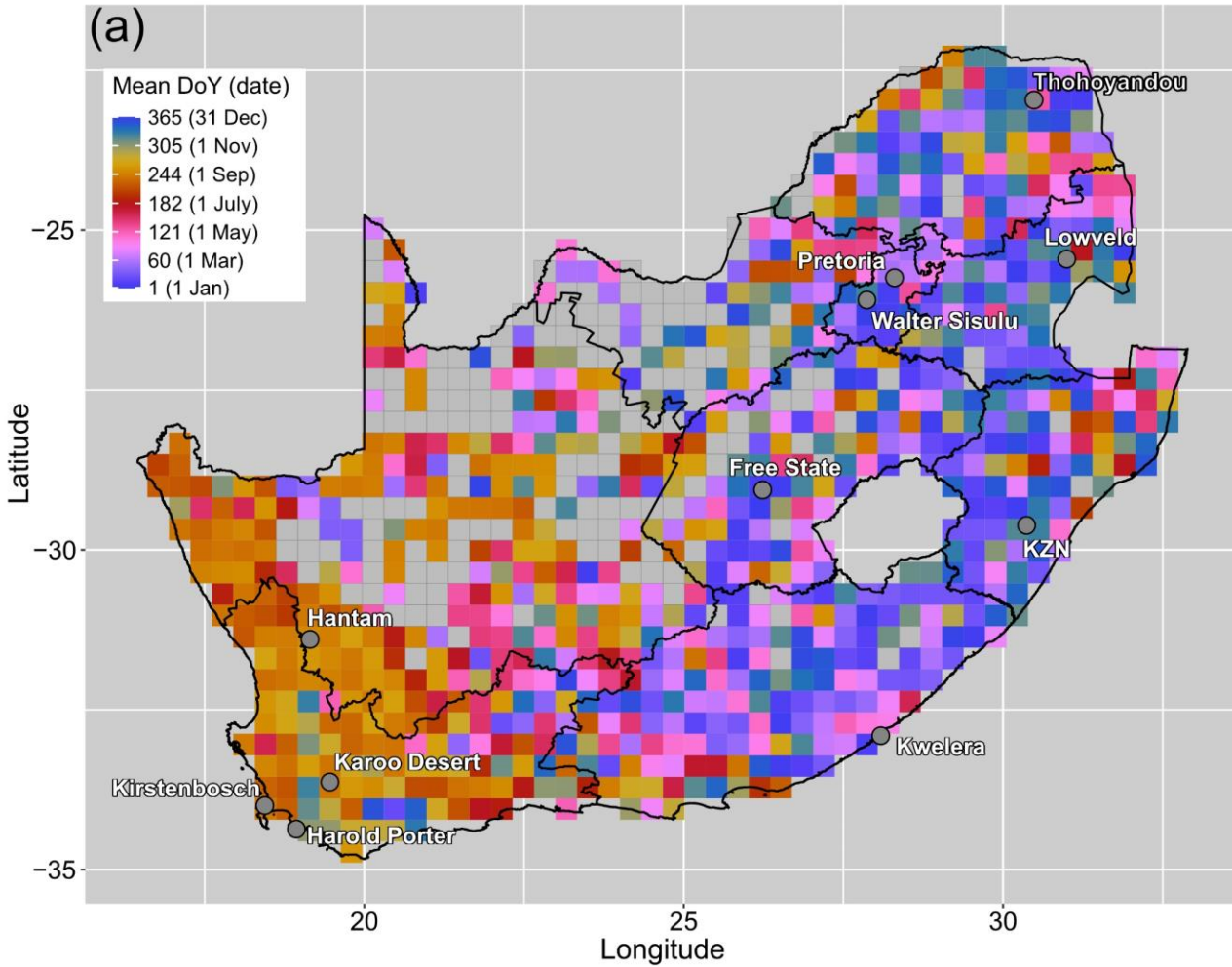
Density Diagram of Flowering Times



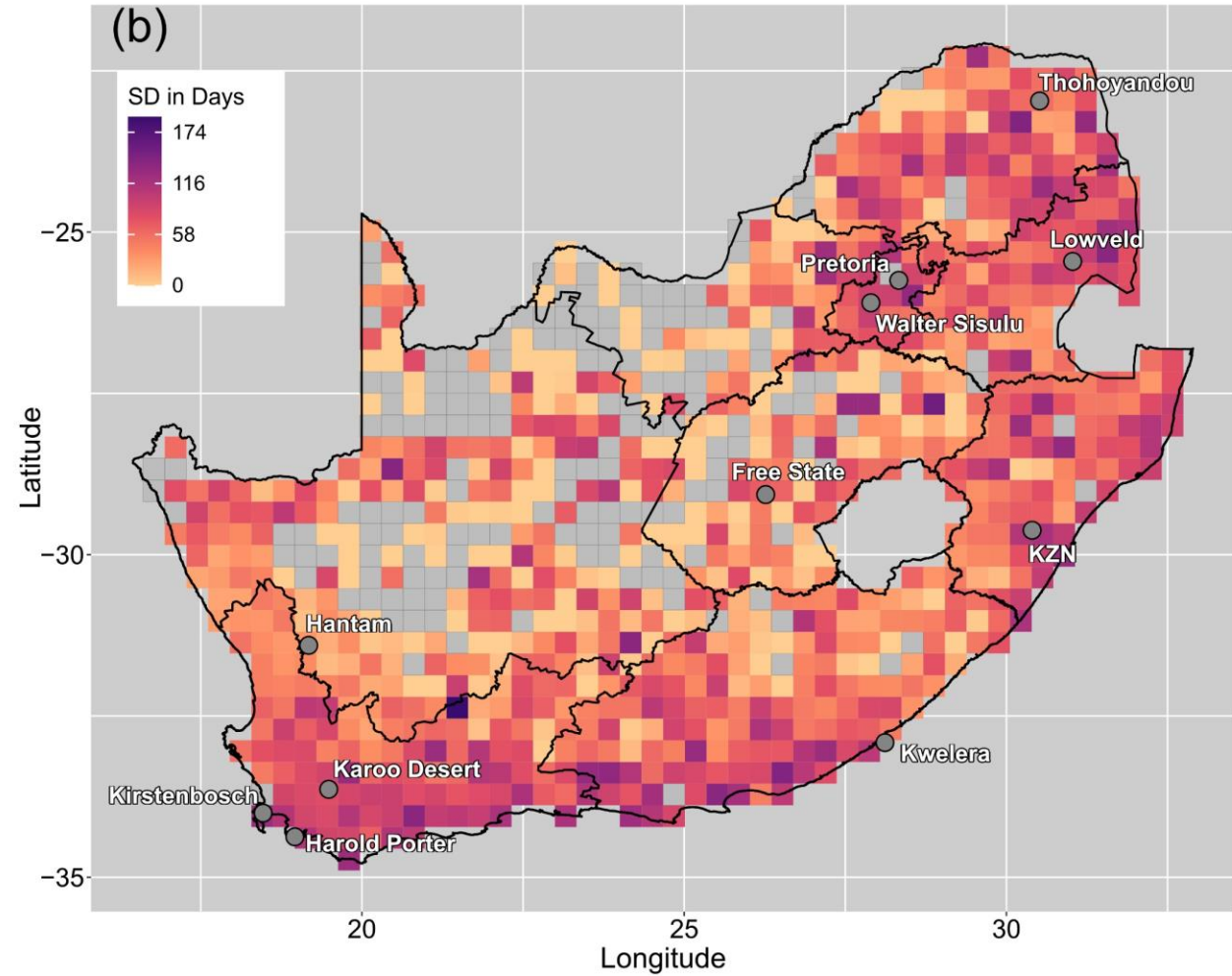
Species with > 5 records
Species = 6,986
Flowering images = 497,911

 95% quantile Flowering time
 Mean Flowering time

Flowering Times of South Africa




(a) – Mean phenology of South Africa



(b) – Variation of phenology of South Africa

Take Home

“All models are wrong, but some of them are useful.”
- George Box

- We have successfully used Machine Learning to determine plant flowering patterns at a large scale, and we can start investigating which species are at risk of extinction due to climate change.
- I want to encourage people to contribute to  iNaturalist

New Results

 [Follow this preprint](#)

Leveraging machine learning and citizen science data to describe flowering phenology across South Africa

 Ross Dylan Stewart,  Nicholas Bard,  Michelle van der Bank,  Jonathan Davies

doi: <https://doi.org/10.1101/2023.12.21.572952>



bioRxiv

THE PREPRINT SERVER FOR BIOLOGY

Aims and Objectives

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- Generate baseline data on insect diversity in South Africa and explore the diversity across different biomes. ✓
- Find the flowering period of all the plants in different parts of South Africa. ✓
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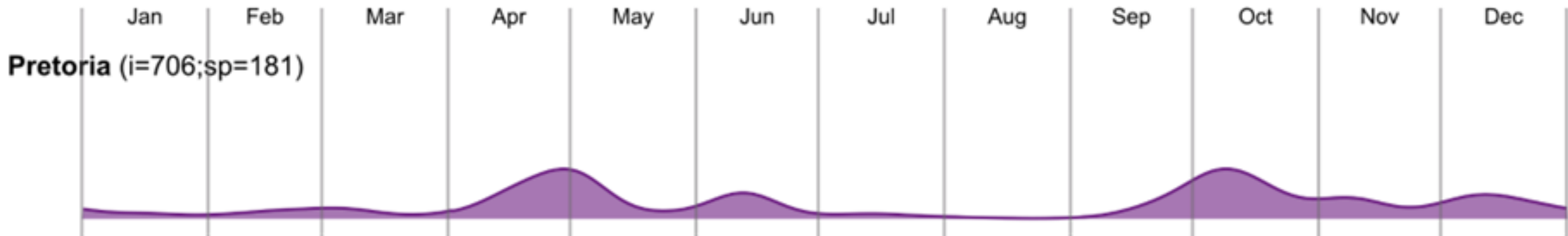
Pretoria Garden

Numbers:

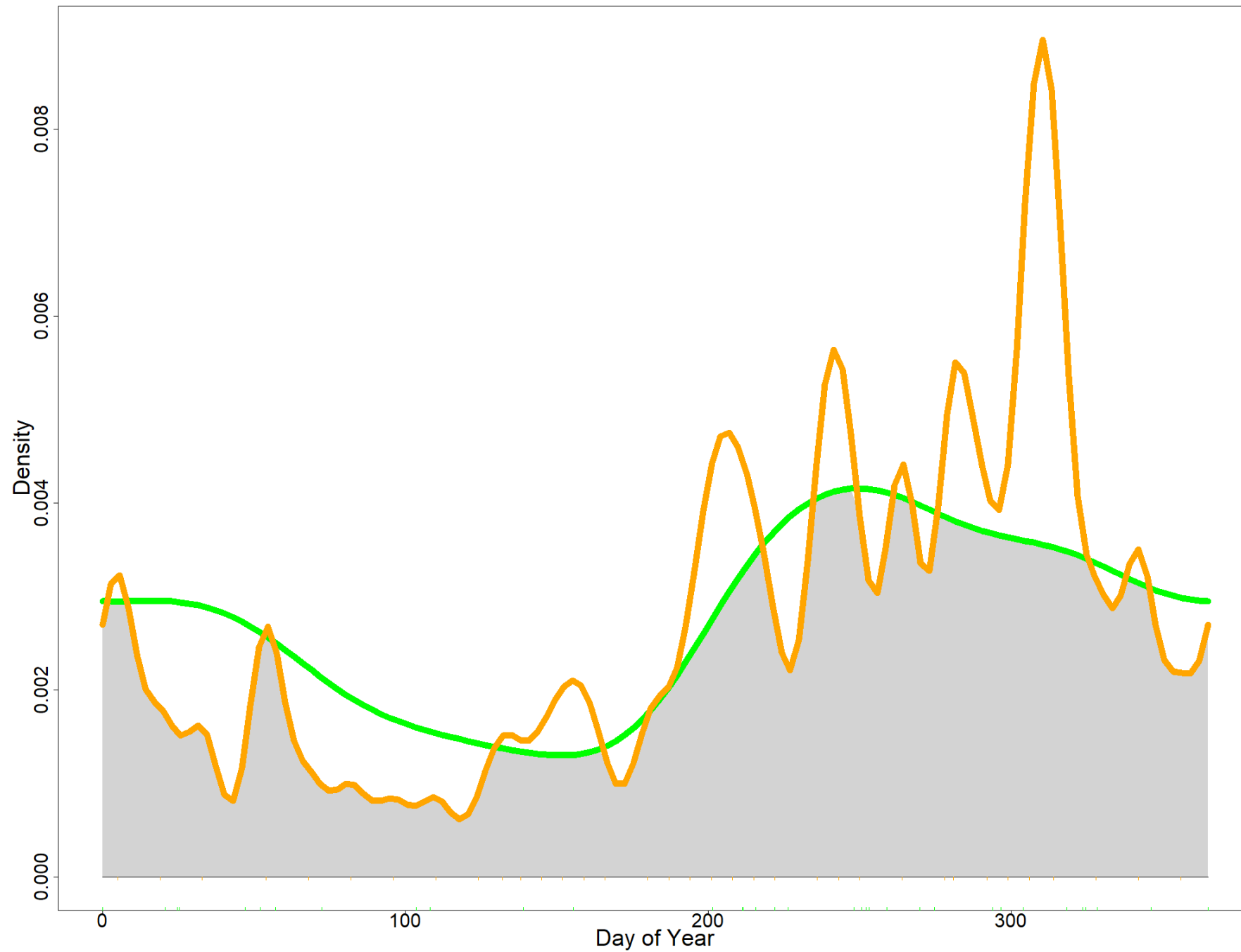
- Specimen: 45,210
- Species (BINs): 5,539
- ID'ed species: 324



Cultivated	Natural
Specimen: 22,350	Specimen: 22,860
Species (BINs): 3,617	Species (BINs): 3,946
ID'ed species: 214	ID'ed species: 246



Overlap_NDVI_insect_shifted_-90



Thank you to Everyone for Helping



The End



GBIF

