The Ecosystem Monitoring and Management Application (EMMA) Workflow:

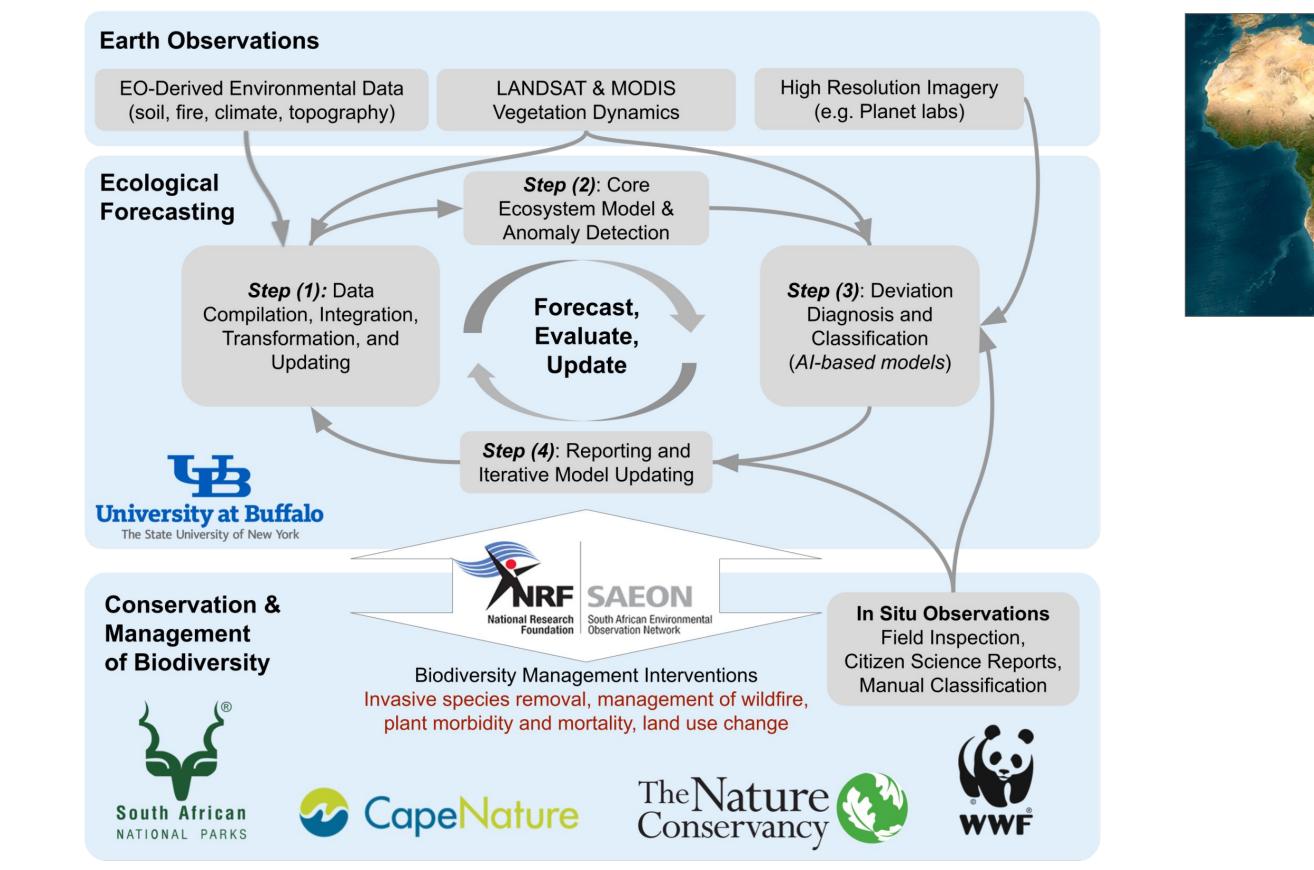
Automating Change Detection and Reporting in the Hyperdiverse Fynbos of South Africa

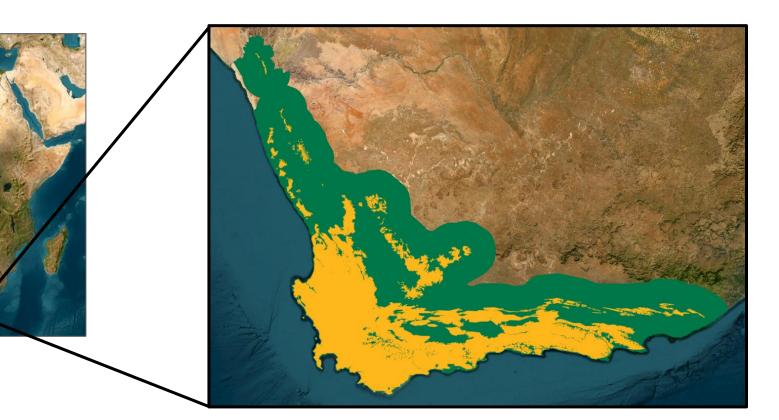
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The challenge:

• Much of the world is naturally open ecosystems, where trees are rare. Change detection in these ecosystems can be challenging due to dynamism from fire and seasonality.





- Here, we present a workflow to automate ecosystem change detection and reporting in one of the world's biodiversity hotspots, the Fynbos of South Africa (Figure 1), in an Open Science framework.
- The project has been co-developed with stakeholders, ensuring relevance to management needs.
- The workflow is publicly available via GitHub, and readily modifiable for other regions or modelling needs, providing a generalizable solution to change detection in open ecosystems.

Figure 1. Conceptual diagram of the EMMA workflow. Our primary end-user is SAEON, but they will serve as a liaison to other provincial, national, and international organizations.



Figure 2. The study covers the BioSCape domain including the fynbos biome (yellow) and surrounding areas within 50 km (green). Photo shows an example landscape on the Cape of Good Hope in the southwestern corner of the region.

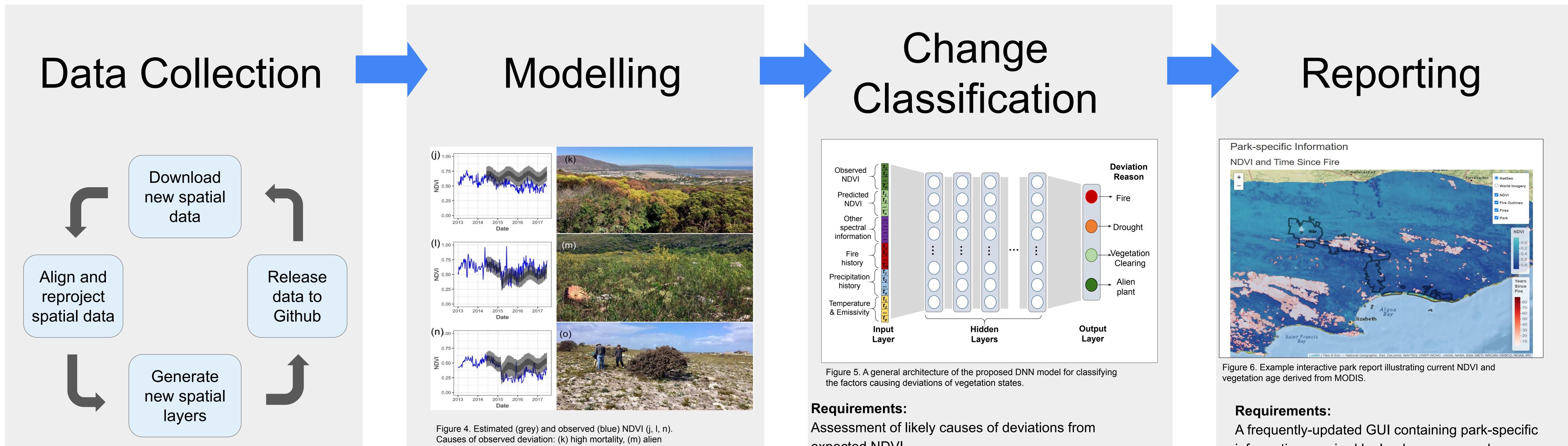


Figure 3. Diagram of data collection workflow.

Requirements:

Continuously-updated and analysis-ready spatial data for the modeling domain (Figure 1).

Implementation:

- R scripts running in a docker environment
- Automatically processes new data every 6 hours
- Downloads new data
- Transforms to a common resolution and projection

invasion following fire, and (o) drought. See tps://doi.org/10.1016/j.isprsjprs.2020.05.017 for more details

Requirements:

1) A model of current NDVI as a function of past NDVI, season, fire history, and environmental covariates

2) Quantification of deviation from expected NDVI

Implementation:

- Hierarchical Bayesian Model
- Current NDVI modeled as a function of site age

expected NDVI

Implementation:

- Deep neural network model that classifies the abnormal deviation locations (pixels) into several categories of interest.
- Model inputs are observed NDVI time series and estimated NDVI
- Other model covariates include:
- additional spectral information
- fire or burned area detection products

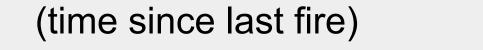
information required by land manager and decision makers

Implementation:

- R scripts in a docker environment
- Data and park reports are updated daily
- Reports contain interactive spatial and temporal data visualizations requested by park managers.
- Integrating reports into decision-making processes with partners and stakeholders

- Quality checks are performed
- Derived products are produced (e.g., time since fire)
- Module is memory conservative and can be run on any operating system
- Data are served freely via Github releases.

Code at: https://github.com/AdamWilsonLab/emma_envdata



- Additional level models the parameters of the negative exponential curve as a function of environmental variables
- Observed and Expected NDVI compared to detect changes
 - Change likelihood assessed using the Continuous Ranked Probability Score



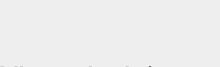
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Code at: https://github.com/AdamWilsonLab/emma_model

- weather data
- MODIS-derived land Surface Temperature and Emissivity
- Model outputs are land change categories:
- fire (or ploughing)
- drought

Code at:

- vegetation clearing
- alien plant invasion



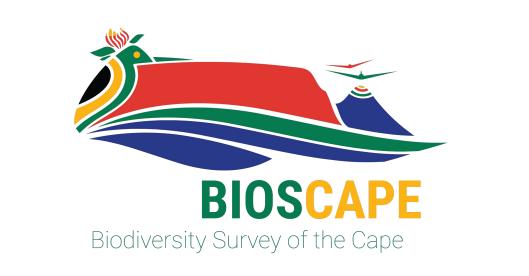
https://github.com/AdamWilsonLab/emma_change_classification

 Collecting feedback and revising system for decision-makers

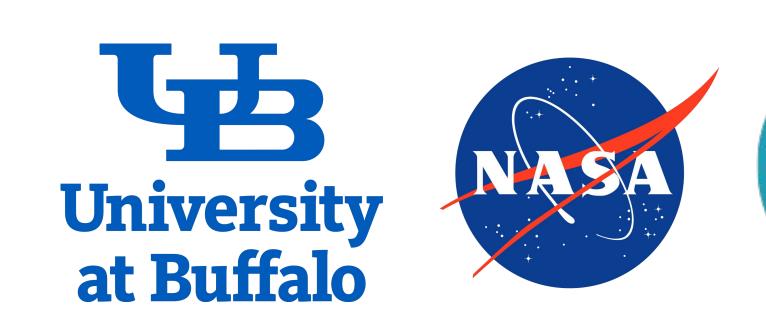


Code at: https://github.com/AdamWilsonLab/emma_report









Biological **Diversity** & Ecological Conservation