Using Machine-Learning to Evaluate and Understand our Capability to Model Tropical Wetland Methane Emissions

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Project Details/Updates

Work initially funded via CMUG has now grown into a wider research project:

- "The First Environmental Digital Twin Dedicated to Understanding Tropical Wetland Methane Emissions for Improved Predictions of Climate Change"
- Funded as part of my 4-year UKRI Future Leaders Fellowship

As part of CMUG project:

- Focused on Africa
- We're developing an emulator for JULES wetland methane
- Will use it's explainability to show which factors matter in the model
- Will drive the emulator with CCI EO data to generate wetland fluxes
- Compare those to a CH_4 inversions performed on GOSAT/TROPOMI ESA-CCI data

As part of FLF:

- Focused on whole Tropics
- We'll extend emulator to other models from Global Carbon Project
- Develop EO ML-based wetland extent datasets
- Combine hydrological models with our land surface models to better represent wetland dynamics
- Improve methane wetland emissions in UK Earth System Model for climate predictions (including ESMValTool recipes for evaluation)
- Develop "climate services" around this capability, providing decision support to stakeholders

The Problems

Complex

Unexplained Increases

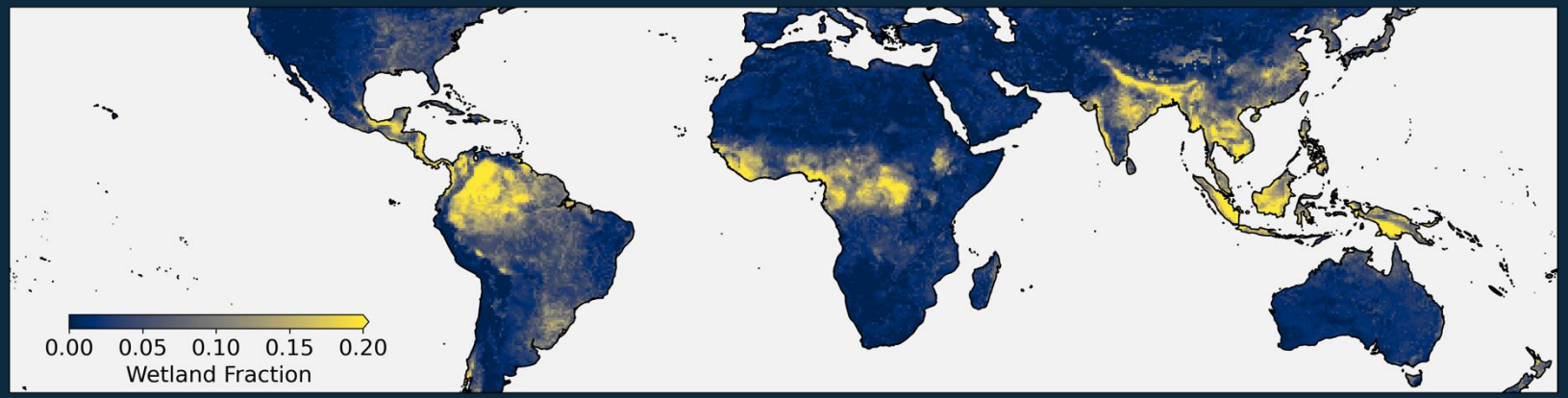
Alarming and Urgent

Tropical Wetlands?

Missing Knowledge

The First Problem.
Significant differences between the methane from models

The Second Problem.
Models fail at correctly simulating the size and location of wetlands



Parker et al., Biogeosciences, 2022

The key research questions that I will address:

1) ***How are tropical wetland methane emissions responding to climate change?***

2) ***How will they continue to do so under future climate scenarios?***

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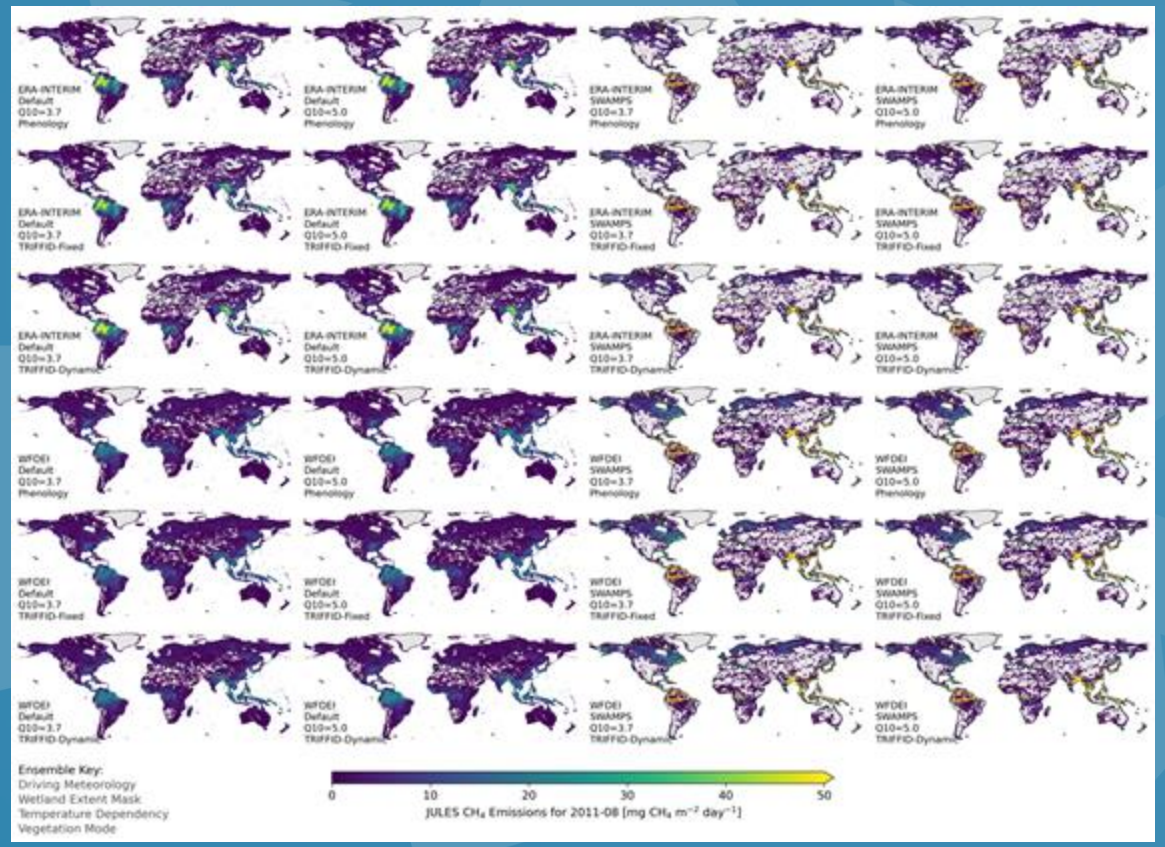
- Models disagree

  "Models demonstrate extensive disagreement in their simulations of wetland areal extent and CH_4 emissions, in both space and time" – Melton et al., 2013

- Intercomparisons are challenging

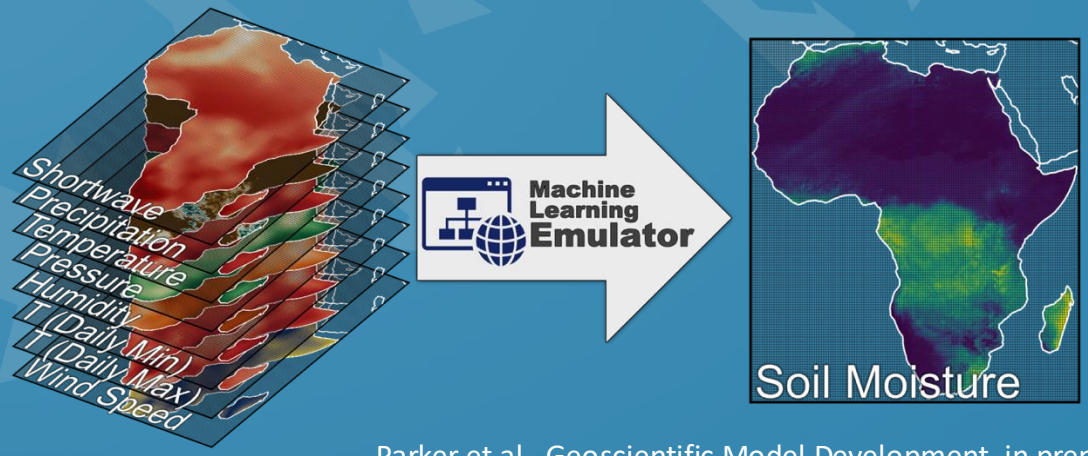


Parker et al., Biogeosciences, 2022

WP 1 – Emulating and Explaining Wetland Models (Years 1-2)

Results in new capabilities to model and explain wetland methane emissions

- Emulators allow novel comparisons
- Explainable AI can be powerful
- => New understanding!



Parker et al., Geoscientific Model Development, in prep

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- Wetland extent = huge uncertainty

  "Our simulated wetland extents are also difficult to evaluate due to extensive disagreements between wetland mapping and remotely sensed inundation datasets." – Melton 2013

- Partnering with Planet

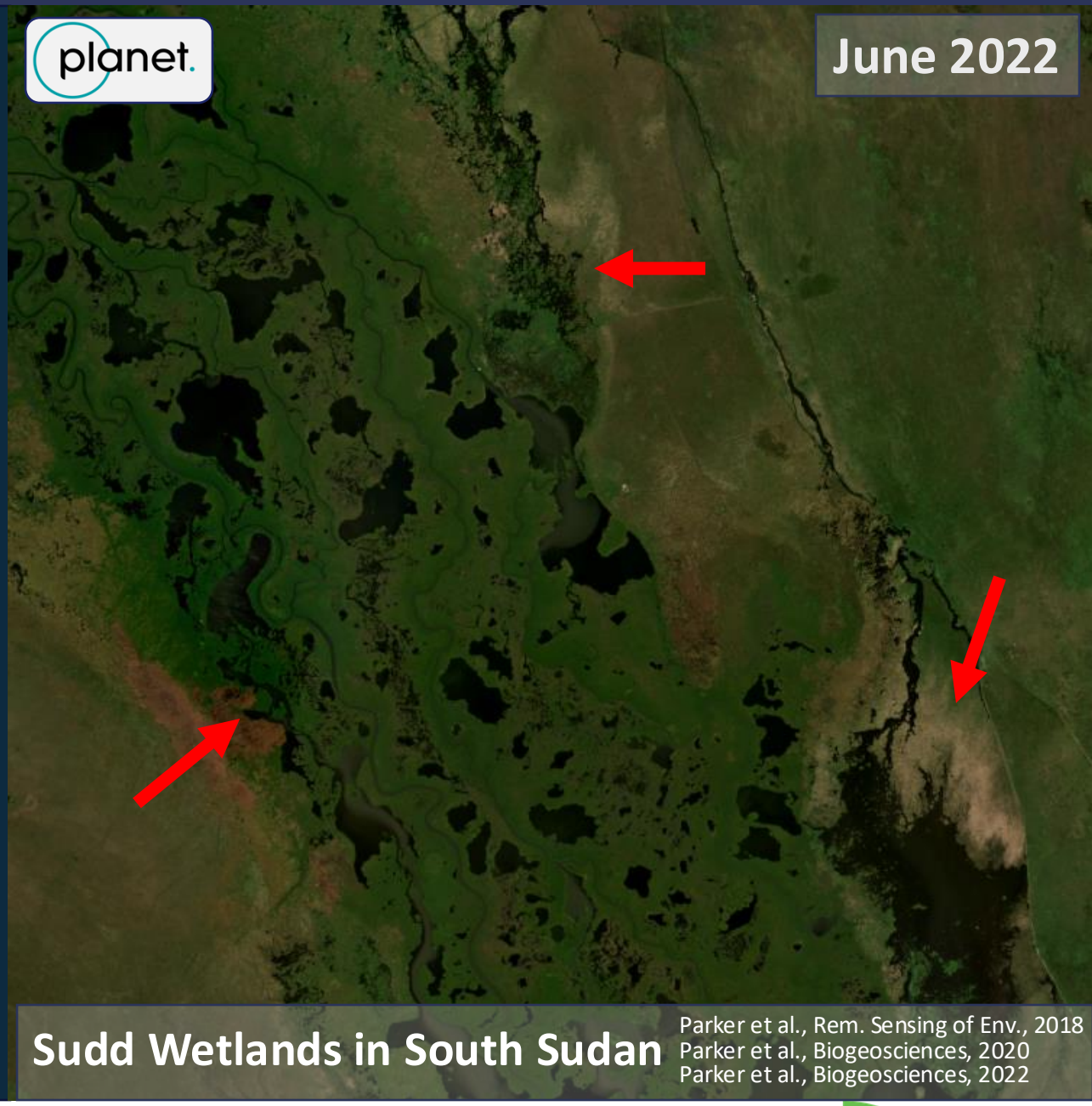
WP 2 – New and Improved Wetland Extent Estimates (Years 1-2)

Results in vastly improved estimates of wetland extent and change

planet. esa GLOBAL CARBON project NEODAAS UK Centre for Ecology & Hydrology

- New ML-based wetland extent dataset

- Improve estimates of wetland extent



Sudd Wetlands in South Sudan

Parker et al., Rem. Sensing of Env., 2018
Parker et al., Biogeosciences, 2020
Parker et al., Biogeosciences, 2022

Vision

We will develop a **new world-class capability in Environmental Digital Twins**, enabling cutting-edge science and truly impacting on climate policy decision-making.

Host Institute

University of Leicester

National Centre for Earth Observation – Natural Environment Research Council

Space Park Leicester

Project Partners

Met Office

esa – European Space Agency

planet.

Global Carbon Project

CGI

UK Centre for Ecology & Hydrology

Stakeholders

WCRP – World Climate Research Programme

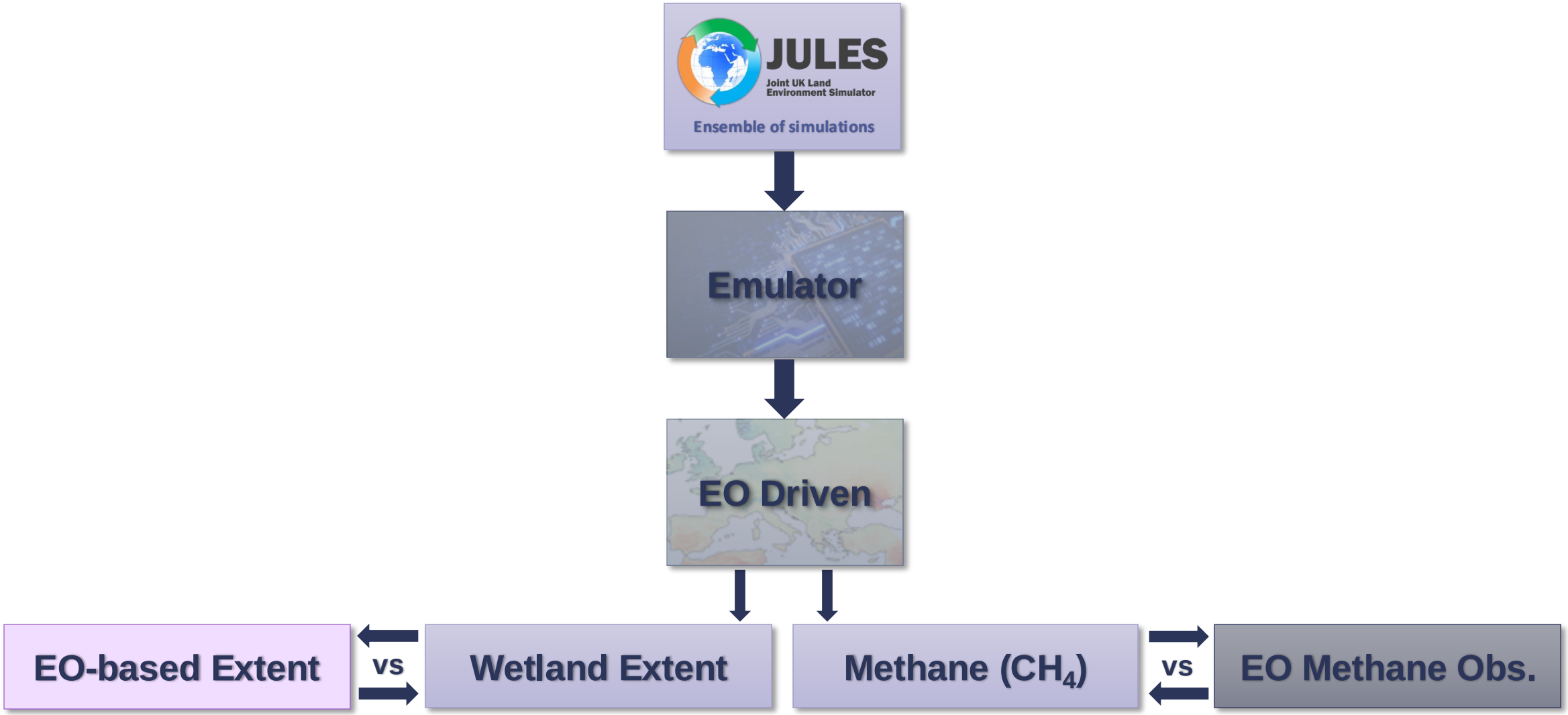
ICPAC – IGAD Climate Prediction & Applications Centre

UN environment programme

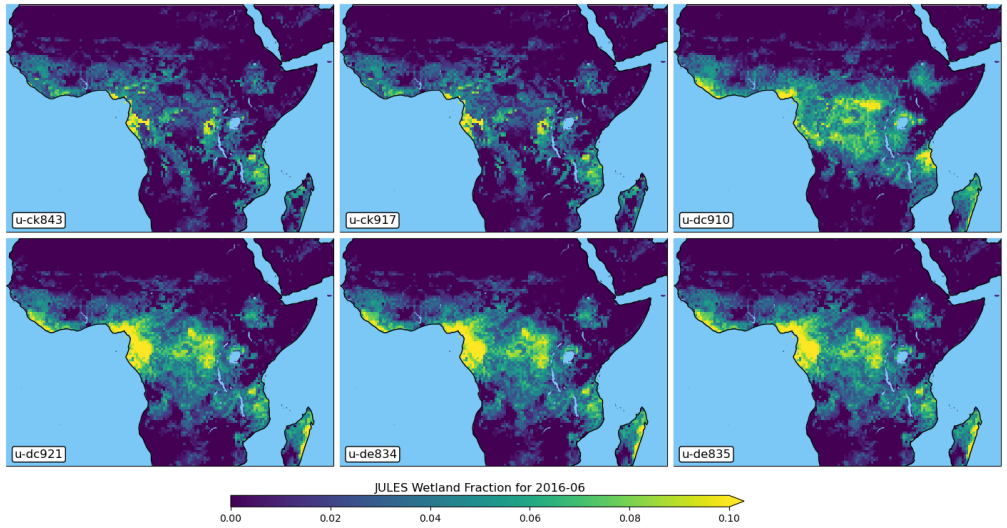
WMO

Ramsar Convention on Wetlands

Methodology

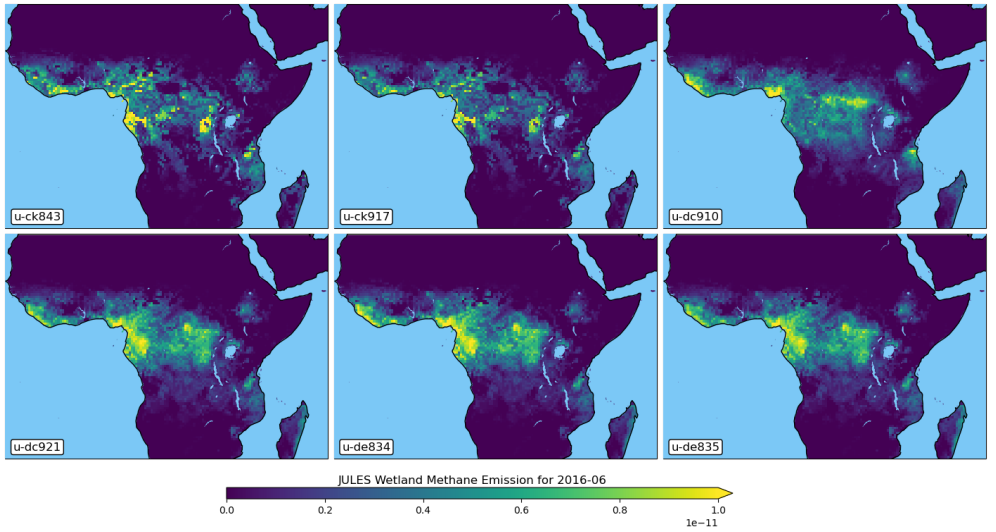


Methodology



Wetland Extent

JULES
Joint UK Land Environment Simulator
Ensemble of simulations

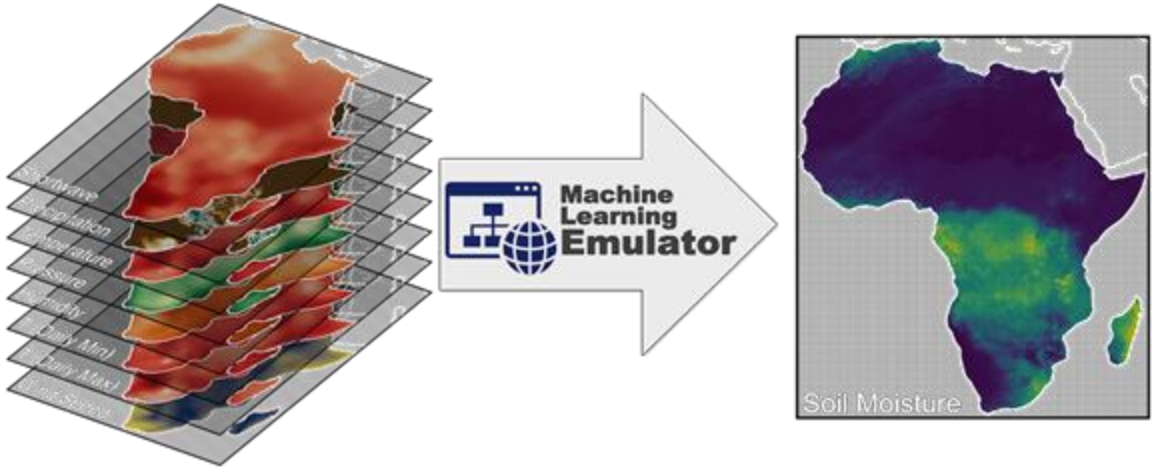


Wetland Methane

- Ensemble of simulations
- Currently 6 members but work ongoing
- Different forcing meteorology
- Different temperature dependencies
- Different soil types

Methodology

We train a **machine-learning** decision-tree model (*emulator*) using JULES data to reproduce wetland extent and methane emissions.



Advantages

- ✓ We can run many simulations **very fast**
- ✓ **No** need for **expert** knowledge
- ✓ **No** need for expensive **supercomputers**
- ✓ We can derive **useful metrics** for users
- ✓ They can be deployed on **web platforms**
- ✓ They can integrate **many types of data**
- ✓ **Explainable AI.**

Methodology

Model-data fusion

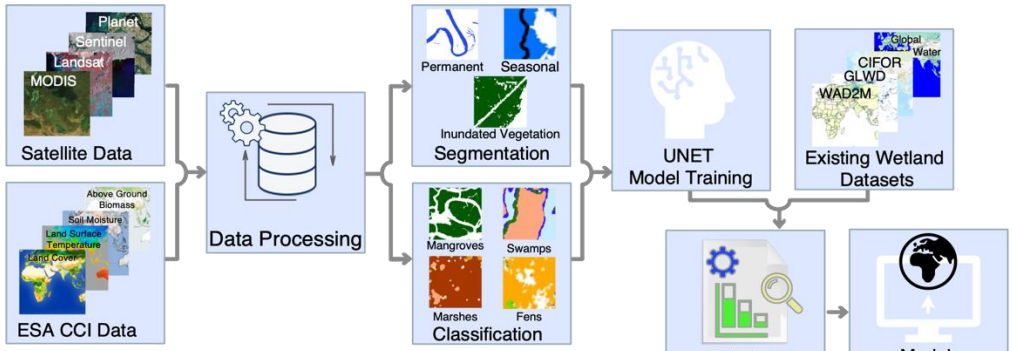
We will **drive the emulator** with input based on **ESA-CCI data** to produce new wetland CH_4 emissions, consistent with observed LST and soil moisture.

land surface temperature cci

soil moisture cci

Methodology

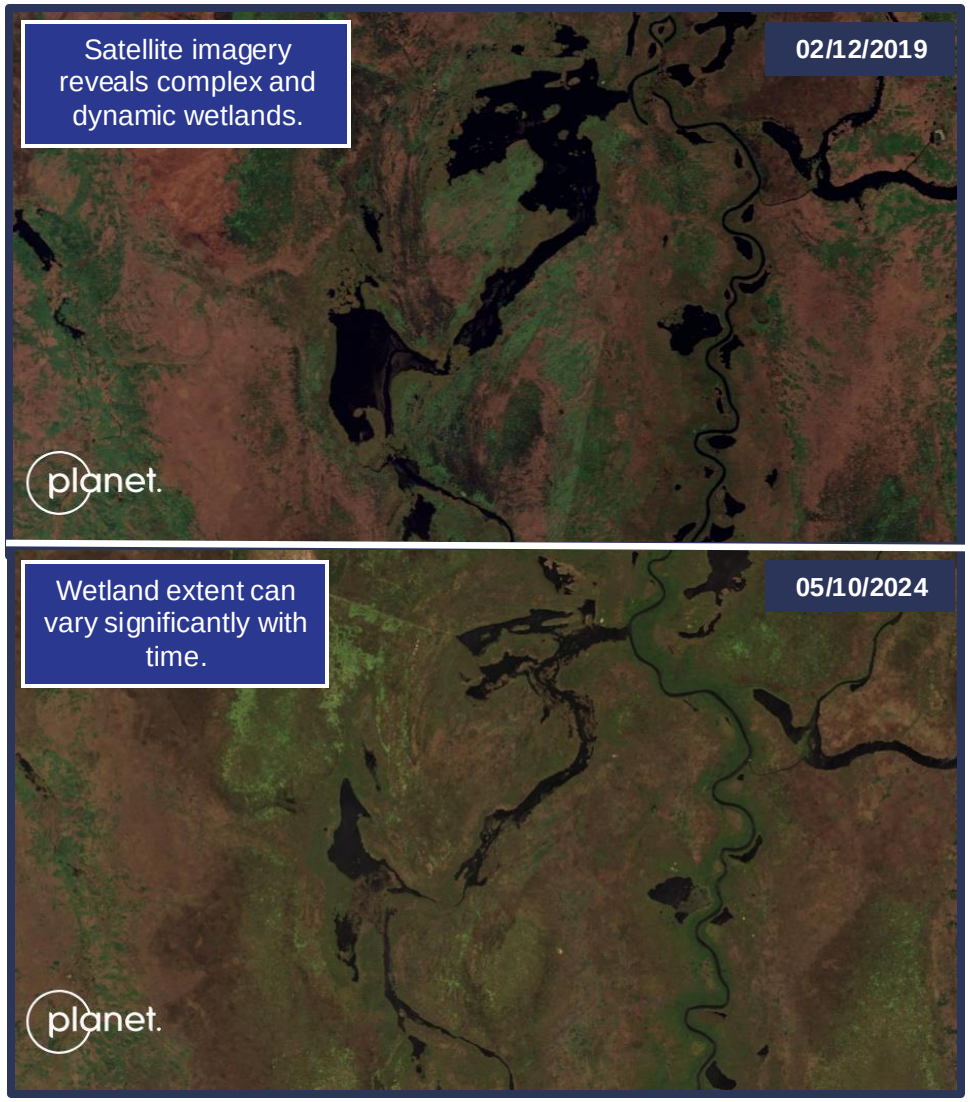
ML-based Architecture for Segmentation and Classification



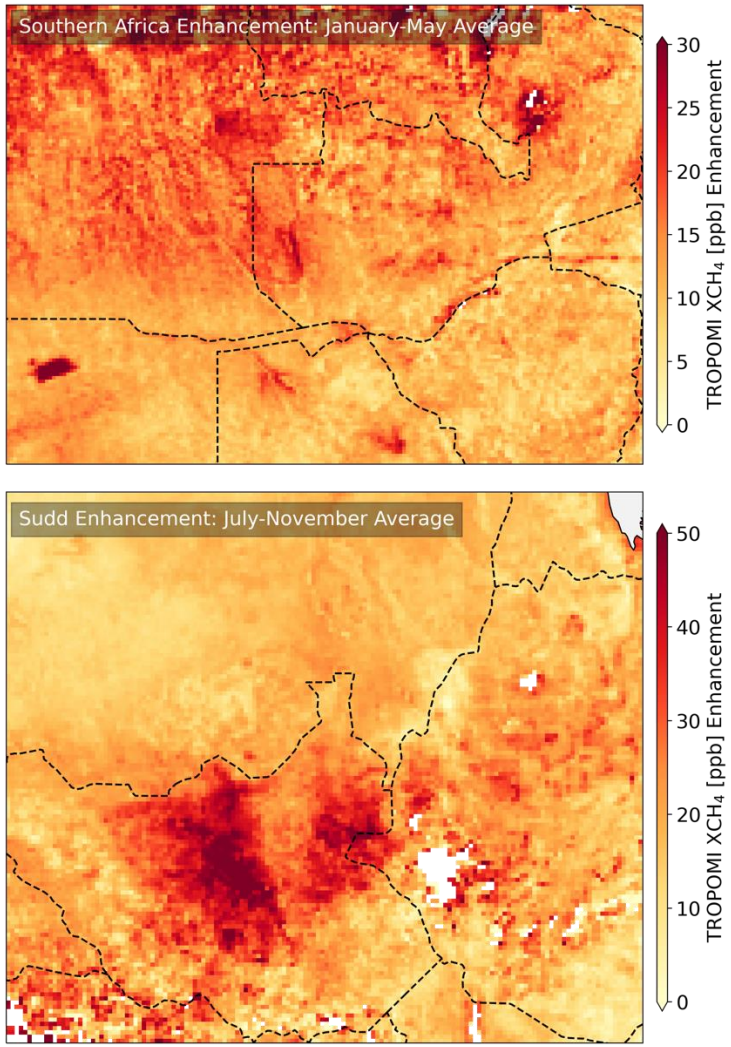
Emulator

EO Driven

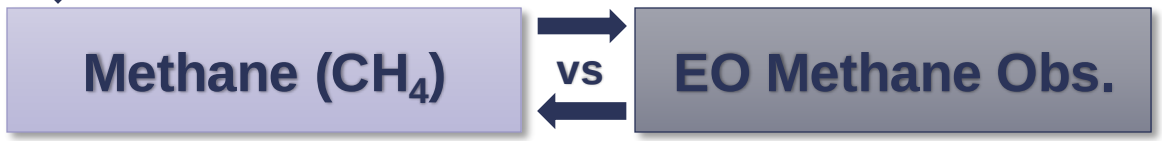
EO-based Extent vs **Wetland Extent**



Methodology



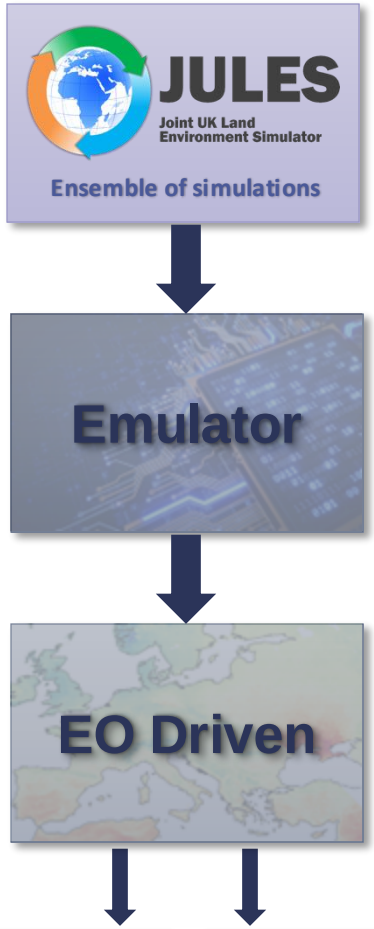
Parker et al. (2022)



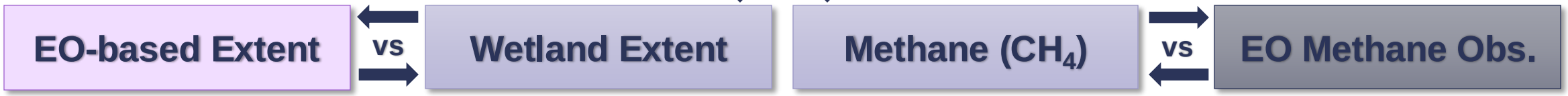
Emulator emissions will be evaluated against **atmospheric inversions** of ESA CCI CH_4 data

Next Steps

- Continue with additional JULES simulations to extend ensemble
- Discuss with CCI teams (LST, soil moisture) on most appropriate datasets to use to drive emulator
- Develop wetland extent datasets and make use of CCI land cover



- Continue to develop emulator
  - Fairly slow process as lots of potential combinations of input features
- Evaluate against GHG-CCI CH_4 data
  - Perform regional flux inversions



Questions: Slido.com #colocation24

For more details, please see poster and talk to Cristina, Khunsa and Chandana ☺