Machine-Learning Emulators of Land Surface Model 'JULES' as Digital Twin building blocks

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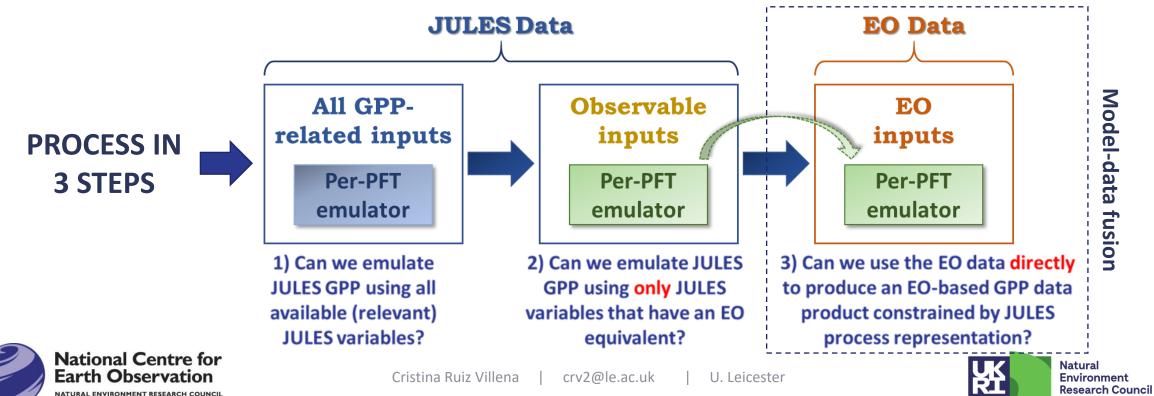


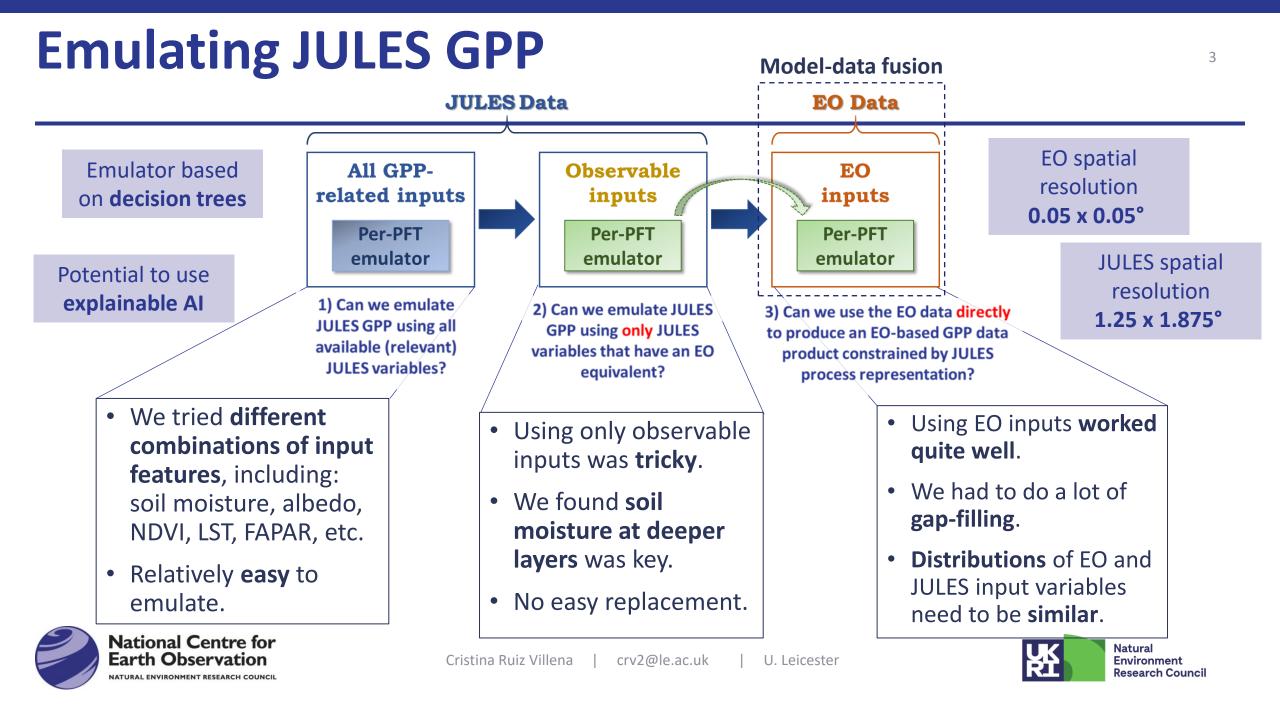




Emulating Europe's Carbon Cycle

- The **carbon cycle** over Europe is still **highly uncertain** and neither observations nor models alone are capable of addressing these issues.
- Emulators allow greater **understanding** of the model behaviour and let us explore the different relationships between the drivers and carbon fluxes.
- We can then use **emulator with Earth Observation data** to derive **new** datasets that are explicitly tied to observations and can make use of their uncertainties.





Results: Emulating JULES – maps of statistics

GPP Statistics for Emulator Performance for Validation Period (2020) Mean Absolute Error (MAE)

95th Percentile

Root Mean Square Error (RMSE)

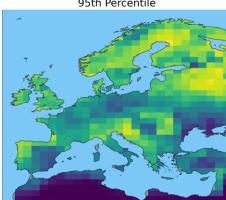
INPUTS

- Min LST
- Max LST
- Soil moisture (top layer)
- FAPAR
- Soil moisture availability factor

OUTPUT

GPP







 \mathbb{R}^2

0.0

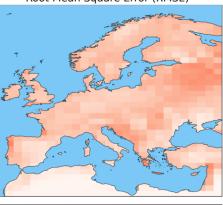
0.2

0.4

0.6

0.8





0.00 0.25 0.50 0.75 1.00 1.25 1.50 1.75 2.00 0.00 0.25 0.50 0.75 1.00 1.25 1.50 1.75 2.00 1e-7 kg m⁻² s⁻¹ 1e-8 ka m⁻² s⁻¹ 1e-8

MAE (scaled)

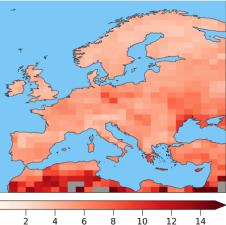
10

12

14

ò

RMSE (scaled)



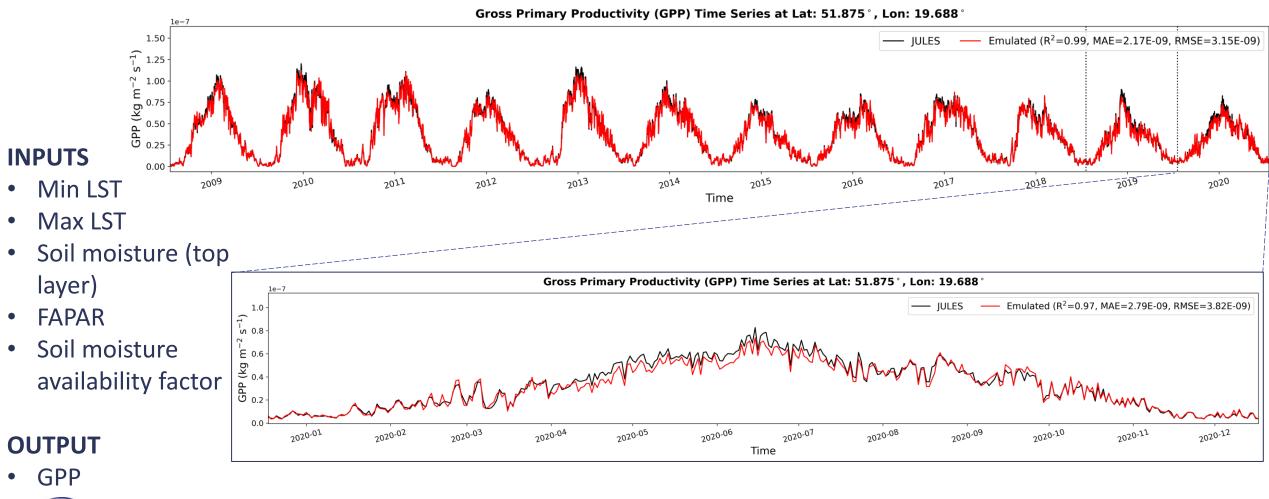
6 RMSE/95th percentile (%)



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1.0 0 8 6 MAE/95th percentile (%)

Results: Emulating JULES – example time series







5

Results: (Emulator + EO) vs JULES – example map

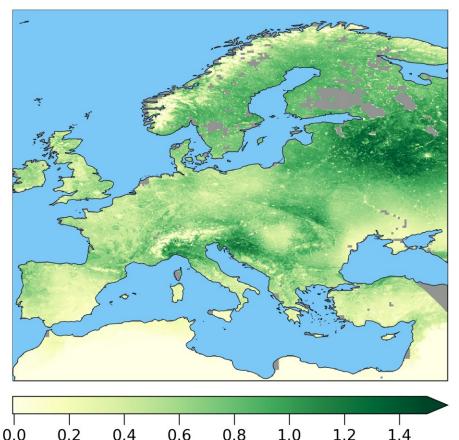
Monthly Average - Emulator vs JULES GPP over Europe on 06-2020

1e-7

Emulator

Driving emulator with EO data:

- Constrains physics from JULES with real observations.
- Provides higher
 detail owing to
 higher resolution
 of the EO data.

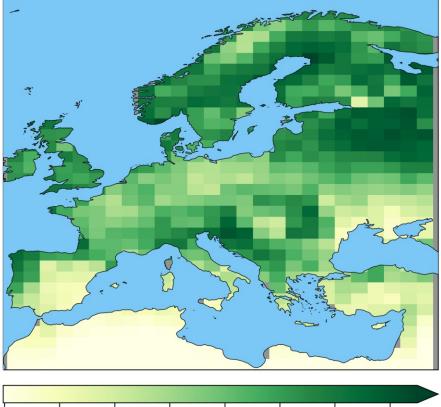


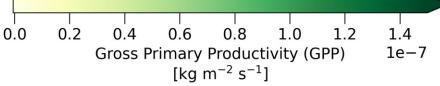
Gross Primary Productivity (GPP)

 $[ka m^{-2} s^{-1}]$

JULES

6





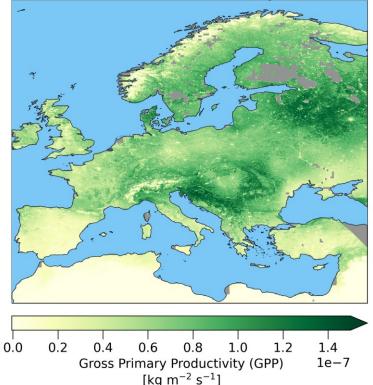


Results: Comparison with other GPP products – example maps

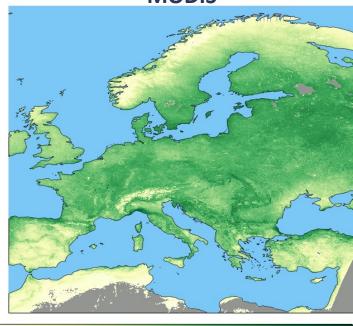
Our data compares well with existing satellite GPP products

GPP monthly average 06-2019

Emulator with EO inputs



MODIS

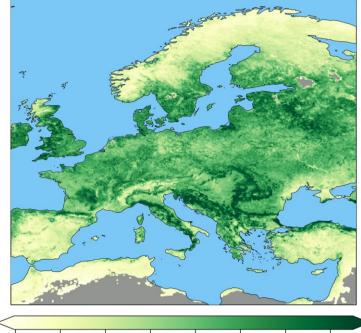


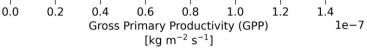
0.2 0.4 0.6 0.8 1.0 1.2 1.4 Gross Primary Productivity (GPP) [kg m⁻² s⁻¹]

1e-7

- 0.0 0.2 0.4 0.6 0.8 1.0 1.2 1.4 Gross Primary Productivity (GPP) 1e-7 [kg m⁻² s⁻¹]

Sen4GPP (SIF-based)





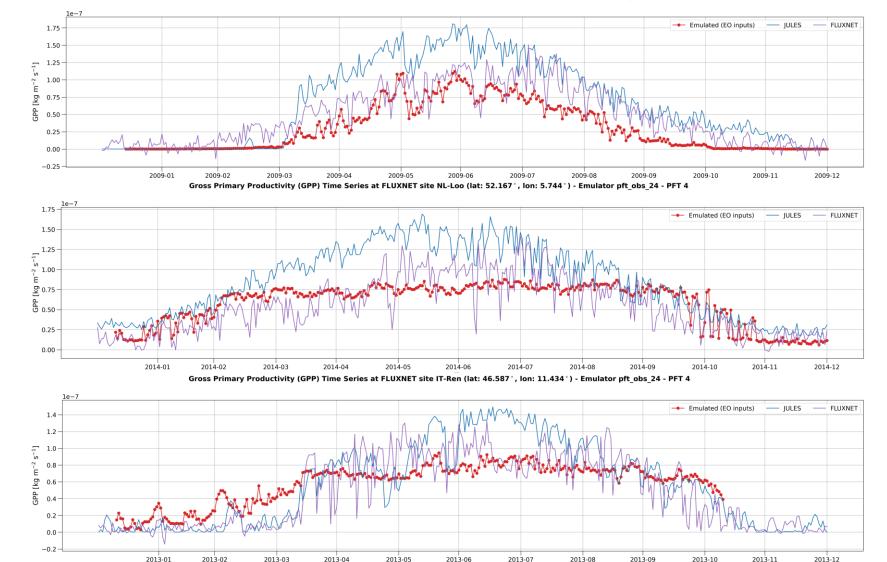
Results: (Emulator + EO) vs JULES vs FLUXNET

Gross Primary Productivity (GPP) Time Series at FLUXNET site BE-Bra (lat: 51.308°, lon: 4.520°) - Emulator pft_obs_24 - PFT 0

Evaluation against FLUXNET

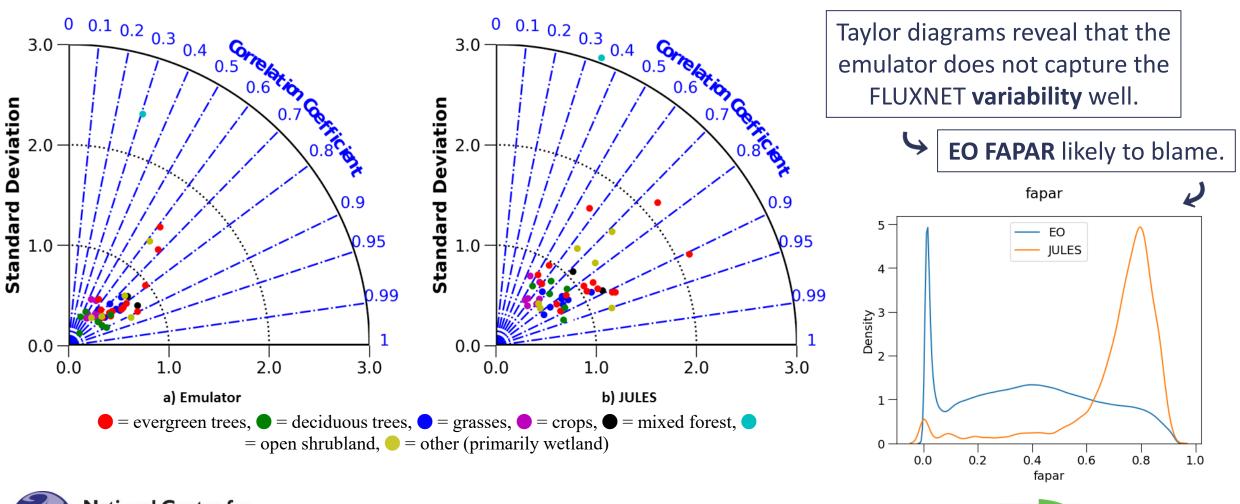
JULES does a good job, though datadriven emulator can do a better job at some sites.





Time

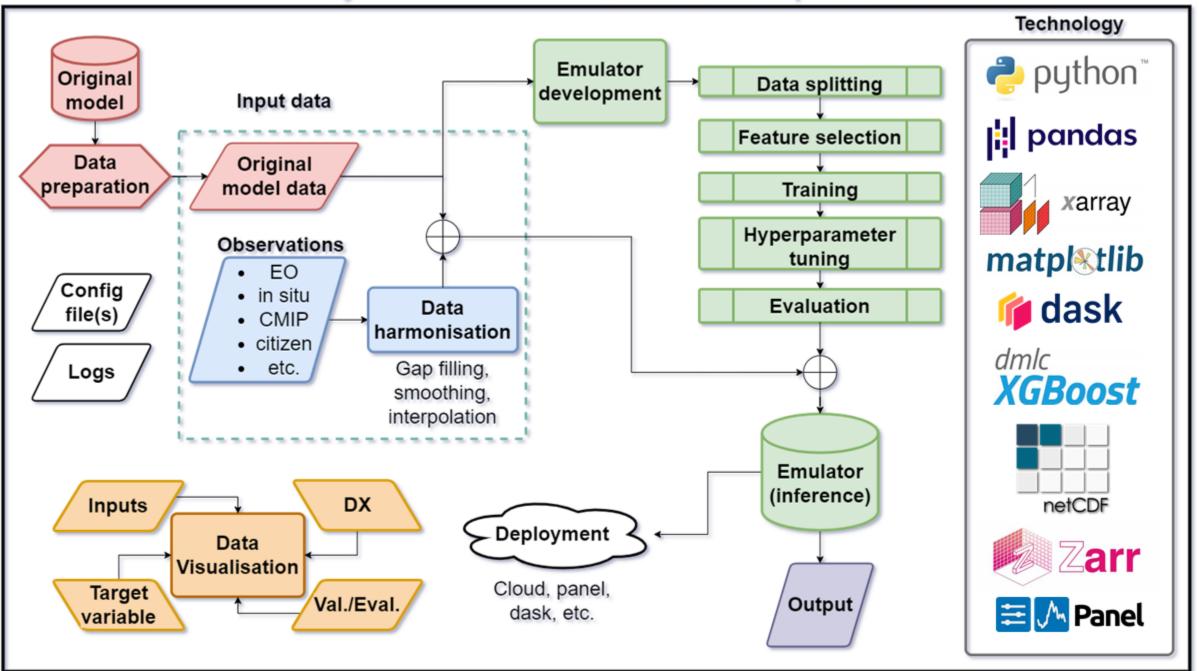
Results: (Emulator + EO) vs JULES vs FLUXNET







Python Framework for Emulator Development



Summary

- □ We developed machine-learning emulators of JULES GPP.
 - □ They are very good at emulating JULES.
 - □ Information on soil moisture profile was key.
- **We combined GPP emulators with EO inputs.**
 - All inputs were observable except fsmc.
 - Results comparable to MODIS and Sen4GPP, and reasonable agreement with FLUXNET (in some cases better than JULES).
 - Differences in FAPAR distribution caused reduced variability.

Future Work

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- □ We are developing a **framework** to speed up development.
- □ We are working on a **soil moisture emulator**, which in future will emulate the entire vertical profile.
- □ We will further explore the potential of **explainable AI**.









SPARE SLIDES





Results: Comparison with other GPP products – example time series

Results at some sites agree well with Sen4GPP.

