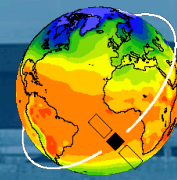


# CMUG Work Packages



# Task 4: CCI contributions to



ESMValTool  
Earth System Model Evaluation Tool



①

## Update/extend existing CCIs Implement new CCIs



aerosol  
cci



permafrost  
cci



biomass  
cci



snow  
cci



cloud  
cci



soil moisture  
cci



land cover  
cci



sst  
cci



land surface  
temperature  
cci



water vapour  
cci

②

## Implementation of uncertainty information

- Implement available **uncertainty information** into the ESMValTool (technical)
- Investigate possibilities to **propagate uncertainty information** to the spatial and temporal scales used by global models
- **Starting point:** work done on implementing uncertainty information for the CCI LAND SURFACE TEMPERATURE (Mittaz et al., 2019)

### Key personell

- **DLR**  
Axel Lauer
- **Met Office**  
Rob King
- **Univ. Reading**  
Claire Bulgin





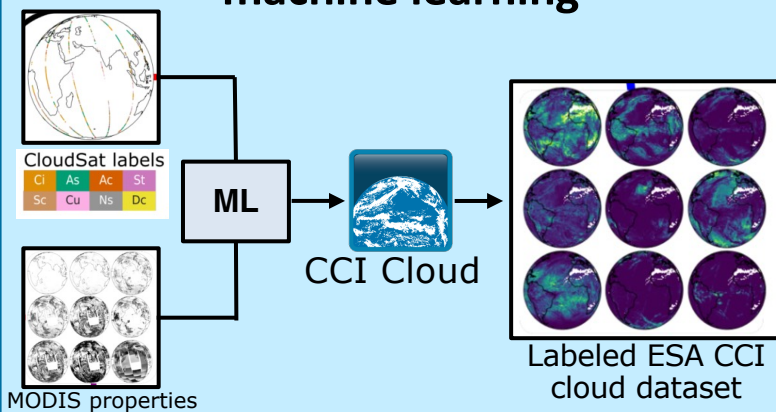
# WP5.1 Machine learning to advance climate model evaluation and process understanding



Lisa Bock, Axel Lauer and Veronika Eyring

## WP5.1.1

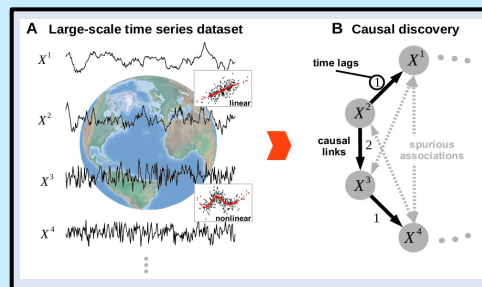
### Enhancing observational products for climate model evaluation with machine learning



- Developing ML-based approach to derive cloud classes from coarse-scale datasets
- Application of NN: timeseries of labelled ESA CCI Cloud data
- Evaluation of climate models

## WP5.1.2

### Causal model evaluation for cloud regimes and land cover types

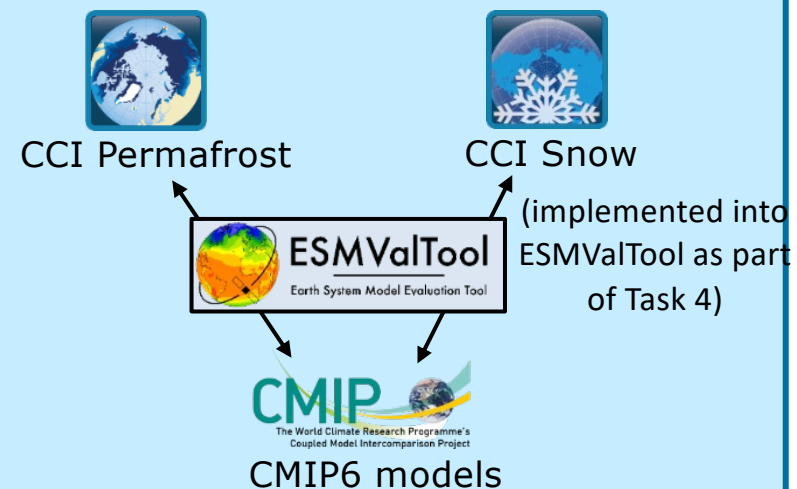


Causal inference (Runge et al., 2019)

- Investigate the causal connections among the cloud properties and their controlling factors (in ESA CCI data)
- Evaluation of global climate models

## WP5.1.3

### Evaluation of CMIP6 models with the ESMValTool



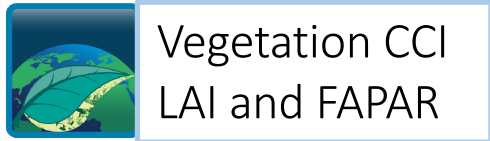
- Evaluation of CMIP6 models
- Investigate use of uncertainty information





### Task 5.2.1

*Testing and feedback on preliminary LAI datasets*



### Task 5.2.2

*Analyses of relationships between phenology and land-atmosphere processes*



Snow



LAI and FAPAR



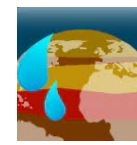
Land cover



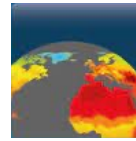
Water vapour



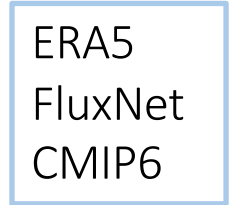
biomass



soil moisture



Land surface temperature



In the development phase of the CCI Vegetation and **interaction** with the CMUG team will provide **testing** and **feedback** on preliminary LAI and FAPAR data.

In this tasks, the **CMUG team** will:

- define a core set of **phenology indicators** at global and habitat scale;
- Quantify the influence of phenology on **land-atmosphere** interactions;
- **Compare** with model and observed values.

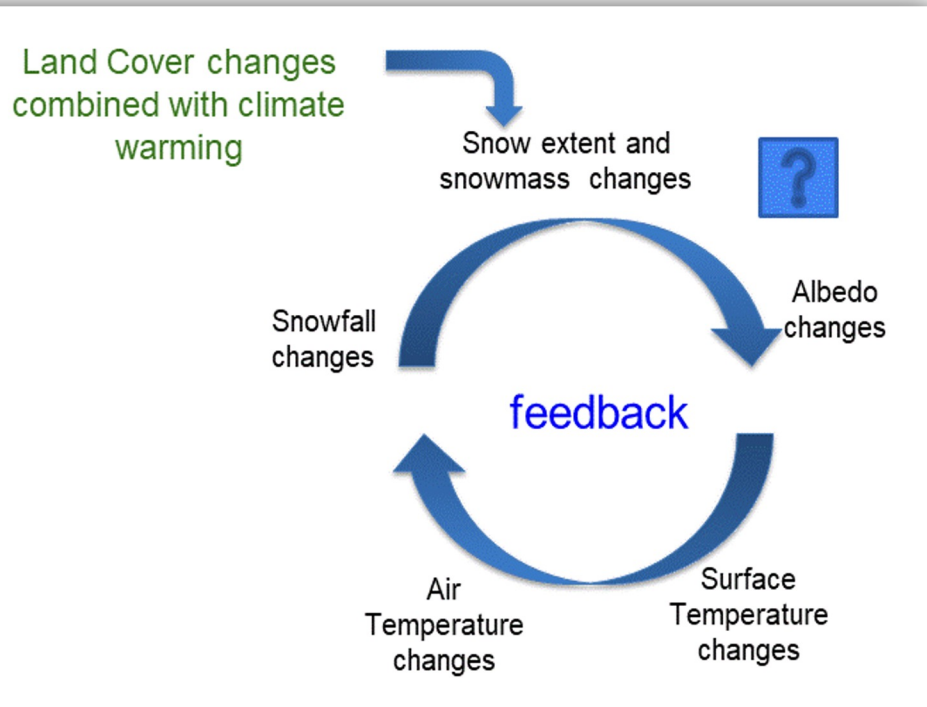


# Snow dynamics impacts on temperate / high latitude climate

**Objective:** Improve our understanding and modelling of snow-vegetation-atmosphere feedback, with the IPSL climate model and CCI products (i.e. Snow, LC,..)

*IPSL (P. Peylin, C. Ottlé, F. Chéruy)*

Potential feedback loop induced by land cover and albedo changes



## Planned work

### 1. **Data Analysis**

Consistency analysis btw Snow Cover (mass & extent) and LC dynamics & other products (LST, Fire, Biomass, Albedo)

### 2. **ORCHIDEE model evaluation**

### 3. **Model developments**

Shrubs – Snow interactions (parallel developments)

### 4. **Snow parameters calibration**

Multi-site Bayesian calibration

### 5. **Coupled model simulations**

LMDZ-ORC historical simulations (prescribed SST)



# ESA CCI data assimilation impact on seasonal predictability of ocean biogeochemistry



- WP1: Assimilation of ESA CCI variables (SST, Sea Ice, SSS, Sea Level, Ocean Color) to produce reconstructions
  - Subtask 1.1: assimilate only physical CCI variables
  - Subtask 1.2: assimilate physical and biogeochemical CCI variables
- WP2: Impact of assimilation choices of these reconstructions on physical and biogeochemical properties
  - Subtask 2.1: evaluate physical properties of reconstructions
  - Subtask 2.2: identify best strategy to reconstruct ocean biogeochemistry
- [Option, unfunded] WP3: Impact of assimilation choices of these reconstructions on seasonal predictions
  - Subtask 3.1: production of seasonal predictions
  - Subtask 3.2: evaluation of seasonal predictions (e.g., ACC, RMS Skill Score)

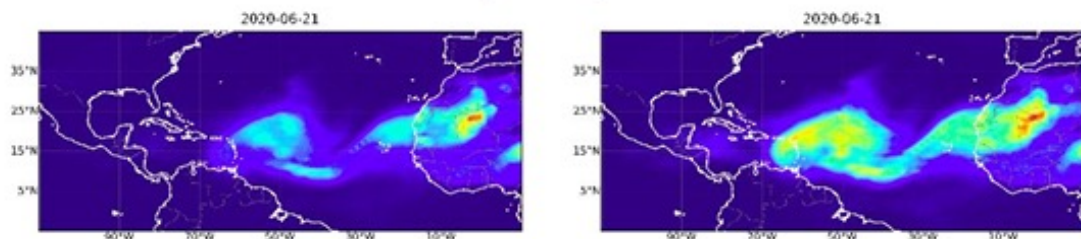


**Aerosol ECVs** : Aerosol Optical Depth (Dust AOD, FM AOD, AOD)

**Cloud ECVs** : Cloud Optical Depth (Cloud Top Height, Cloud Fraction, Ice Water Path, Liquid Water Path)

## WP5.5.1 Dust aerosol analysis with the BSC system

Jeronimo Escribano (BSC)

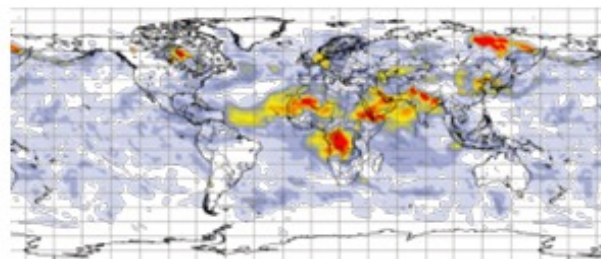


Constrain global **dust** aerosol simulations from the BSC MONARCH model with CCI data to produce dust analyses during the extraordinary event of June 2020.

→ Explore pixel-level uncertainties, Coarse AOD vs DOD, Comparison with DOMOS results.

## WP5.5.2 Cloud/Aerosol analysis with the ECMWF system.

Angela Benedetti and Kirsti Salonen (ECMWF)



Joint assimilation of **aerosol** and **cloud** ECVs in the ECMWF IFS during June 2020 and September 2021 with the IFS 4DVar scheme in CAMS configuration.  
→ Impact of COD and AOD level 2 data on the 4D-Var analysis

## OWP5.5 Cloud and Aerosol Analysis Validation Study:

Evaluation using the ESMValTool and internal tools at BSC/ECMWF

**Soil Moisture, Water Vapour ECVs.**

A. Benedetti and K. Salonen (ECMWF), Axel Lauer (DLR), J. Escribano (BSC)



## Scientific questions to be addressed:

- ⚙ Can regional (global) climate models represent accurately the atmospheric and surface processes affecting the Greenland and Antarctic ice-sheets?
- ⚙ Where, when and why do the surface mass (energy) balance of the models processes perform least & most well?

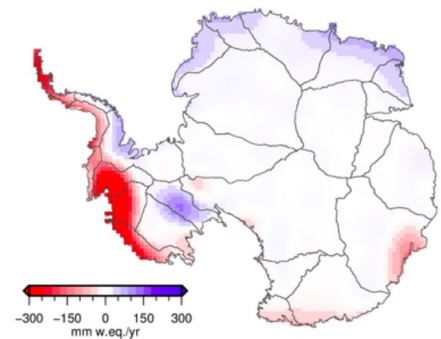
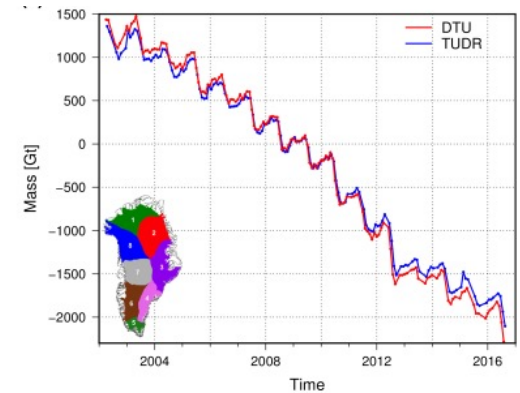
## CCI ECV's:

- ⚙ Gravimetric Mass balance and Surface Elevation Change
- ⚙ Land Surface Temperature
- ⚙ Total Column Water vapour
- ⚙ Cloud (fraction, ice/water path) and radiation fluxes (surface and TOA)
- ⚙ *Snow on Greenland ice-free margins*

## Partners:

**SMHI** (Swedish Meteorological, Hydrological Institute) Ulrika Willén, Klaus Wyser

**DMI** (Danish Meteorological, Hydrological Institute) Ruth Mottram, Shuting Yang



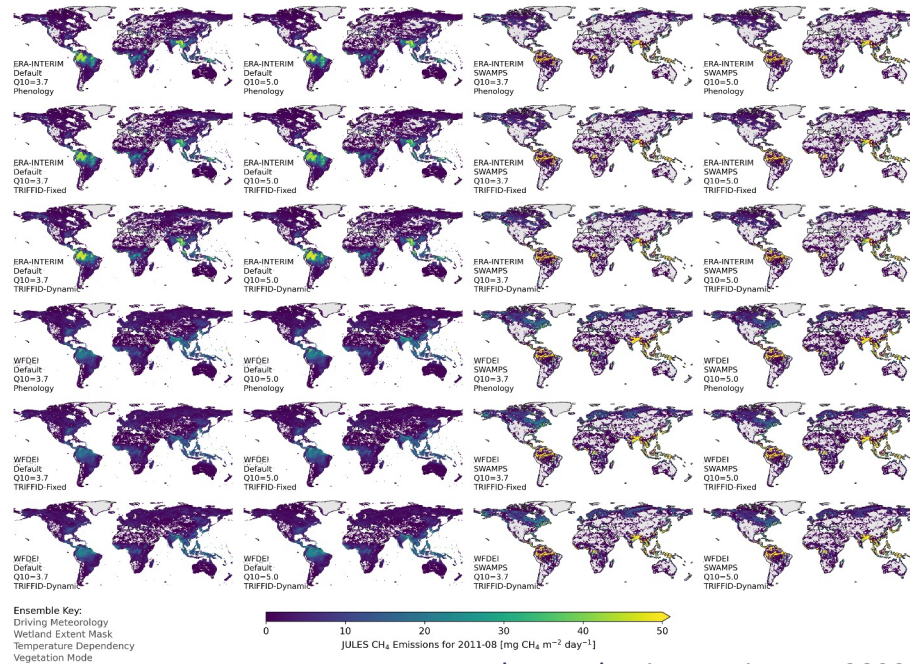
GMB\_figure.png



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

- ❑ Team: Rob Parker and Cristina Ruiz Villena (NCEO-Leicester), Nic Gedney (Met Office), Paul Palmer (NCEO-Edinburgh)
- ❑ In short: We'd develop an emulator for JULES wetland methane, use it's explainability to show which factors matter in the model, drive the emulator with CCI EO data to generate wetland fluxes and compare those to a CH<sub>4</sub> inversions performed on GOSAT/TROPOMI ESA-CCI data.

Figure: Ensemble of JULES simulations with different driving data, temperature dependency, vegetation and wetland mask show massively different methane fluxes!



Parker et al., Biogeosciences, 2022

## CCI Datasets

- ❑ GHG (methane)
- ❑ Land Surface Temperature
- ❑ Soil Moisture
- ❑ Land Cover
- ❑ + Vegetation (?)

## Models

- ❑ JULES (land surface)
- ❑ GEOS-Chem (atmospheric)

# Break out rooms

## 10-11am First Session:

- Room 1 = WP5.1 Machine learning for process understanding (moon room)  
**Cloud, SST, WV, LST, SM, LC, Snow, Permafrost**
- Room 2 = WP5.8 Machine learning for wetland methane emissions (mars room)  
**GHG, SM, LST, LC**

## 11:00 Coffee break 30'

## 11:30-12:30 Second Session:

- Room 1 = WP5.3 Land cover (moon room)  
**LC, Snow, SM, LST**
- Room 2 = WP5.6 Snow dynamics (mars room)  
**Snow, LC, Fire, AGB**
- Room 3 = WP5.7 Ice sheets (ECSAT 243)  
**AIS, GIS, LST, WV, cloud**
- Room 4 = Drop in for ESMValTool demo (ECSAT A013)

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