

# MOTECUSOMA: MOnitoring The Energy Cycle for a better UnderStanding Of cliMate chAnge



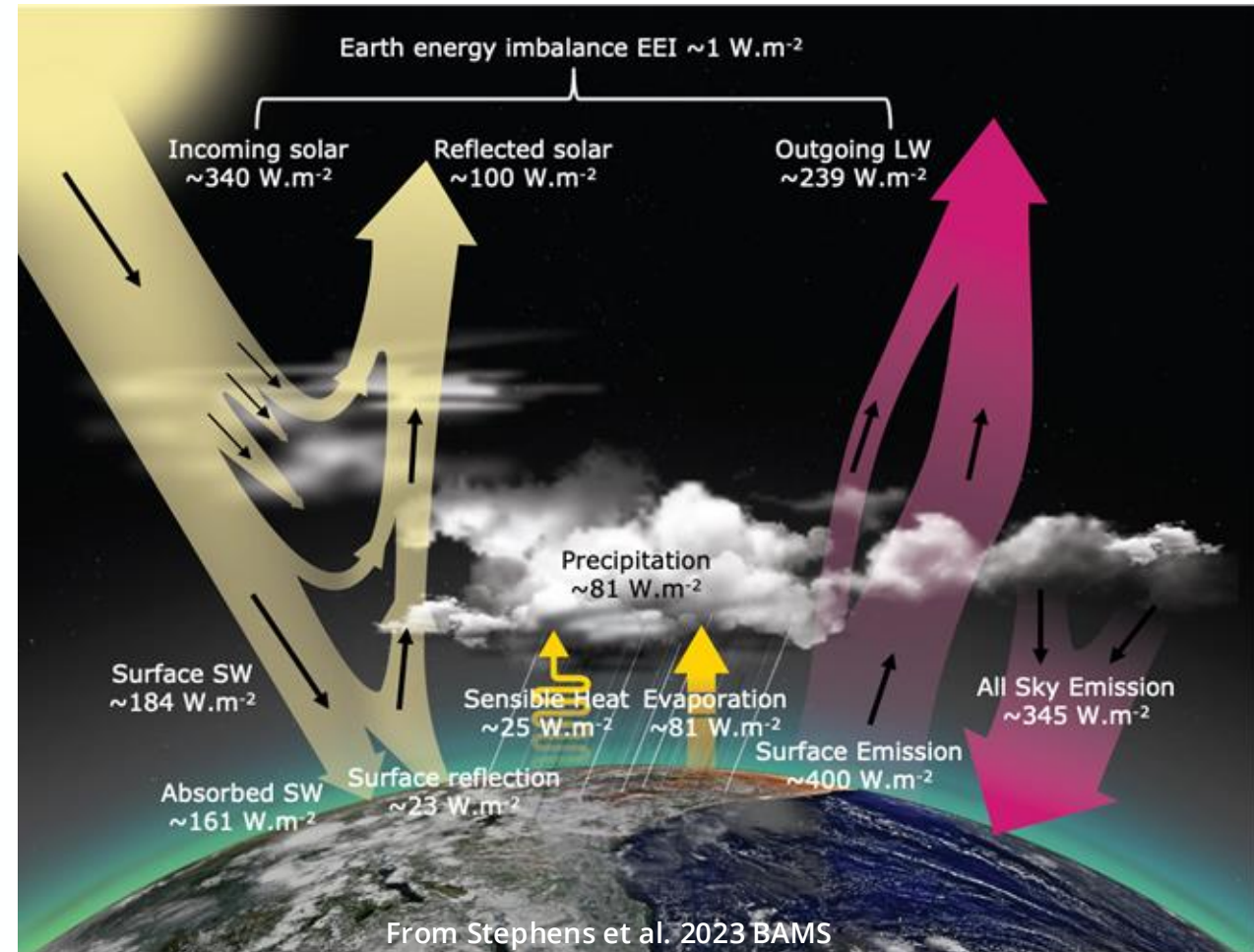
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\*: represented by Robin Fraudeau (Magellium)



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General objective: Understanding the global energy cycle to better understand climate change

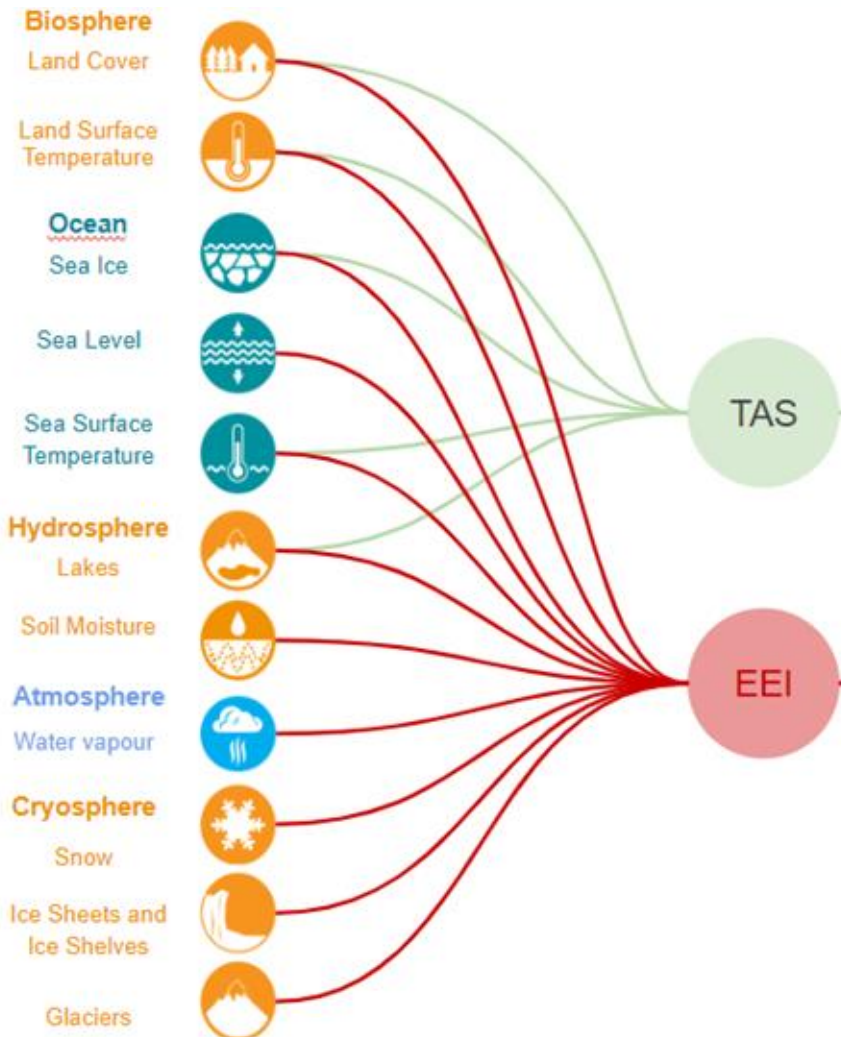
- Understand how changes in the atmospheric GHG impact climate variables (Earth Energy Imbalance (EEI) and Air Surface Temperature (TAS))
- Need to estimate the fluxes in the energy cycle and their changes (TAS and EEI trend, global warming Hiatus, etc...)
- Evaluation of the physical efficiency of mitigation policies



# Project Objectives & Approach

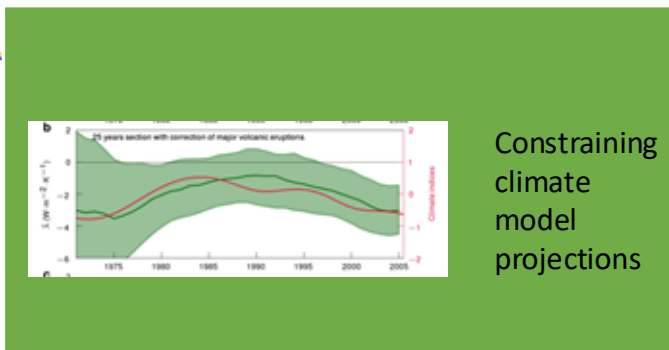
- Projected started in September 2024
- 3 years duration

ECVs



## Scientific analyses

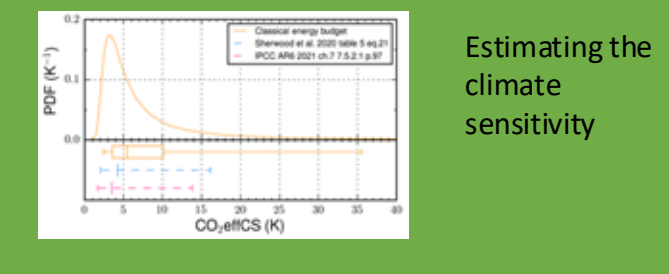
- TAS Cross-ECV Analysis
- WCRP/GCOS Earth's energy cycle and budget
- Climate feedback & climate sensitivity
- Evolution of temperature over the Ice zone
- EEI Response to Mitigation policies



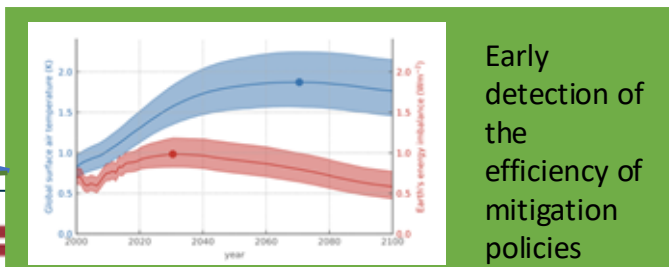
Constraining climate model projections



Estimating the global energy cycle fluxes



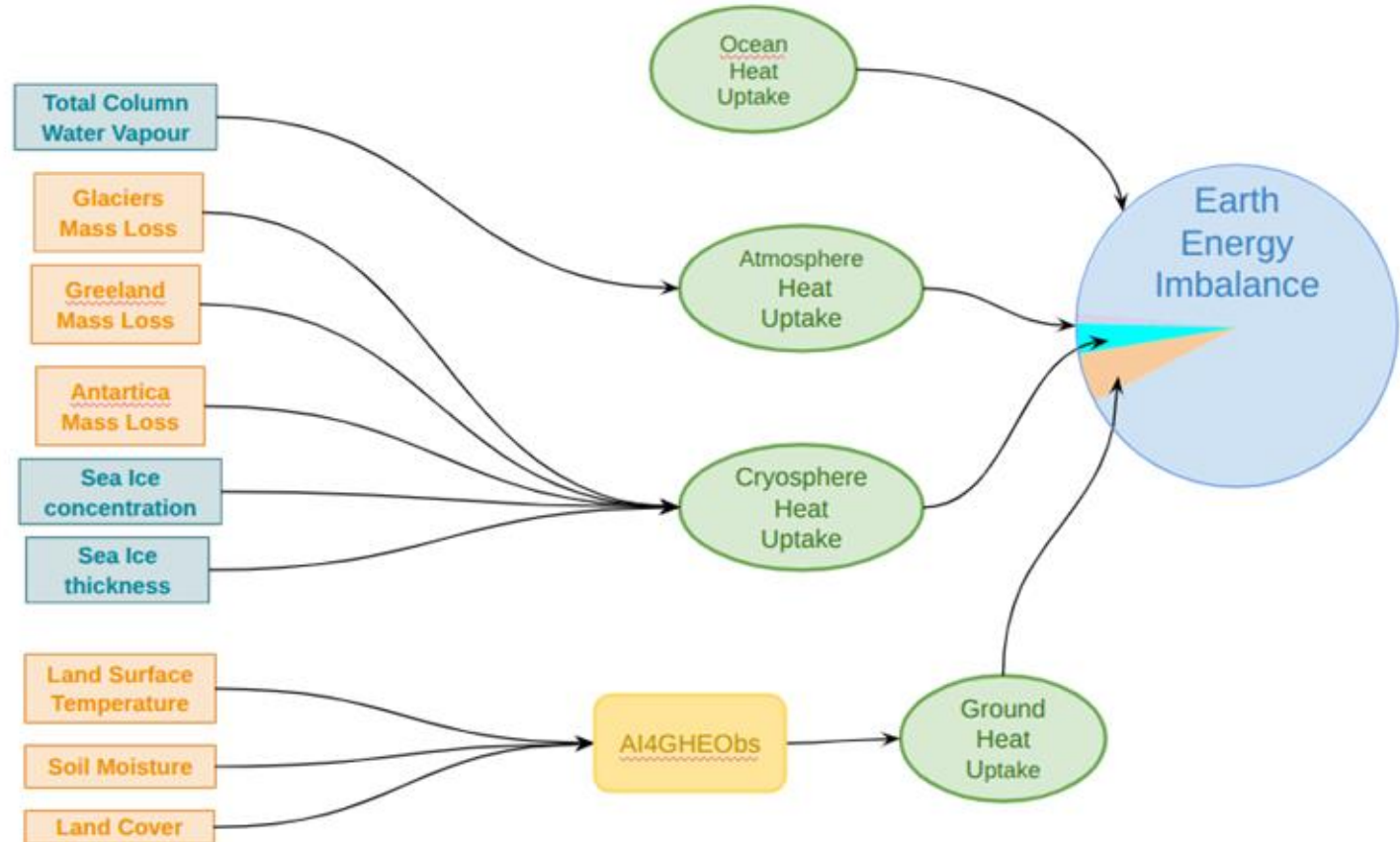
Estimating the climate sensitivity



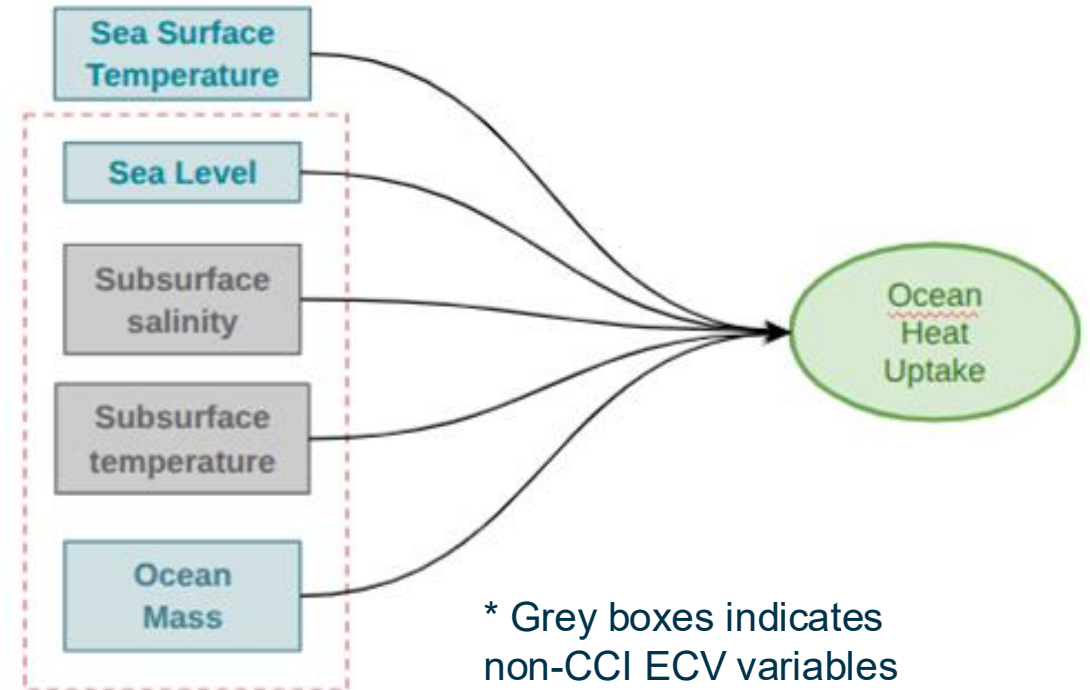
Early detection of the efficiency of mitigation policies

# ECVs being used for EEI

- EEI is estimated based on the different reservoirs contribution (ocean, atmosphere, land and cryosphere)
- Depending on the needs, CCI ECVs are used in purpose of estimation, validation or uncertainty assessment
- ~ 91% of the EEI is coming from the oceans contribution



- Ocean Heat Uptake (OHU) is estimated from the ocean heat content time tendency
- OHU is derived from different CCI-ECV and non-CCI products
- MOTECUSOMA and SLBC\_cci+ are **strongly connected** since the consistency of the sea level components is assessed in the SLBC\_cci+ project (red box)

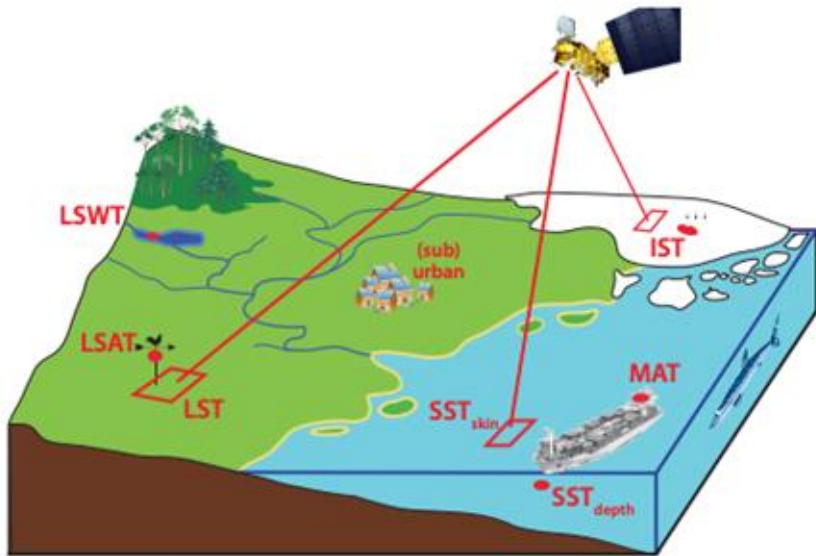


# ECVs being used for EEI

- Summary of the ECV datasets used for the EEI generation or validation, with their limitations :

	ECV	CCI Dataset usage (product/validation)	CCI dataset limitation	Dataset used
Ocean Heat Uptake	Sea surface temperature	EEI product		
	Sea level, ocean mass, etc...	Presented in SLBC_cci+ presentation		
Atmosphere Heat Uptake	Total Column Water Vapour	EEI validation	time coverage ( $\geq 2002$ ) no ocean coverage	ERA5
Cryosphere heat Uptake	Sea ice (concentration/ Thickness)	EEI validation	time coverage ( $\geq 2002$ )	GIOMAS
	Glaciers	No	Mass Balance not provided, only provide glacier area	Zemp et al. Dussailant et al.
	Ice sheet	EEI validation	time coverage ( $\geq 2002$ )	IMBIE
Ground Heat Uptake	Land surface temperature	EEI product		
	Soil moisture	EEI validation		ERA5
	Land cover	EEI product		

# ECVs being used for TAS



Merchant et al, 2013

LSAT = land surface air temperature

MAT = marine air temperature

TAS = the global combination of LSAT and MAT

TAS is a GCOS “surface temperature” ECV but not directly retrievable from satellite.

But satellite surface temperature measurements can constrain the values of TAS using skin-TAS relationships developed per domain

CCI CDR	Use for satT-to-TAS	Limitations
Sea surface temperature (ESA SST CCI)	Training: L4 with in situ MAT and NWP winds (ERA5)  TAS (MAT): L3 SST	Daily cycle not resolved fully
Land surface temperature (ESA LST CCI)	L3 LST with station TAS and various auxiliary inc. ERA5 winds, training and data	Daily cycle not resolved fully prior to SEVIRI or for central Asia
Sea ice surface temperature (C3S/AASTI)	In situ, L4 IST, L3 IST locations, Sea ice concentration	Daily cycle not fully resolved
(Land) Ice surface temperature (C3S/AASTI + LST CCI)	In situ, L4 IST, L3 IST locations	Daily cycle not fully resolved

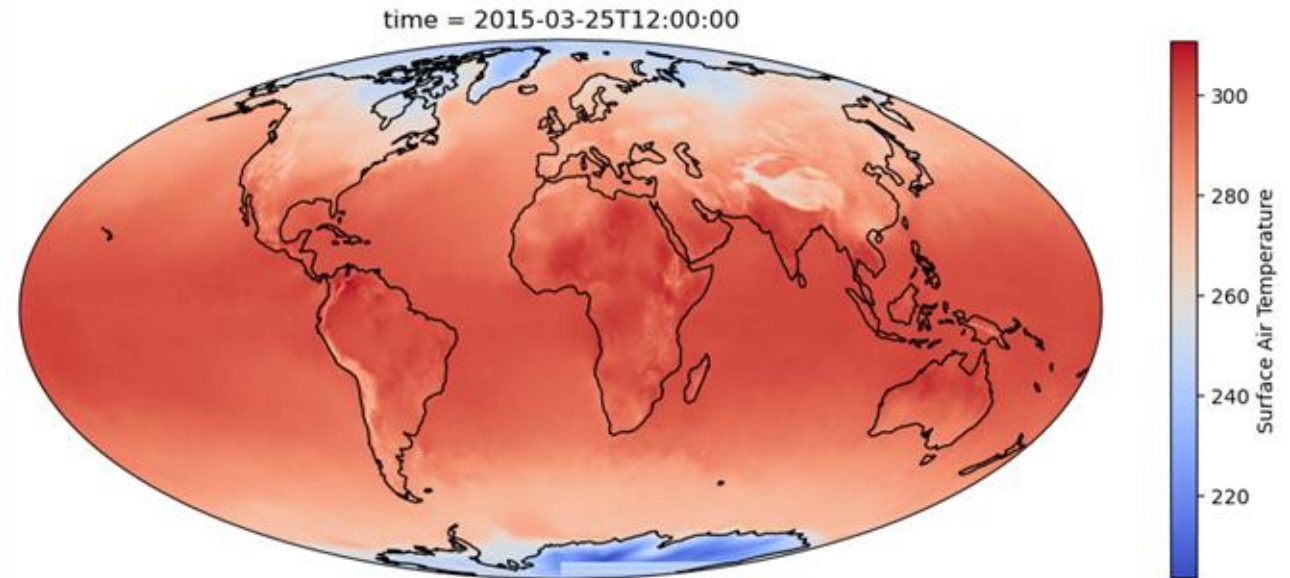


# Results & relevance to the CCI community

A trial analysis of TAS has been generated using satellite and in situ TAS inputs, for 2015 and 2016

The analysis system seeks to balance optimally the observational constraints from all the different sources of data, and using spatial structures of correlation.

Realistic models of uncertainty and error correlation are crucial to creating the best possible TAS estimate:



MOTECUSOMA trial analysis, 25 March 2015

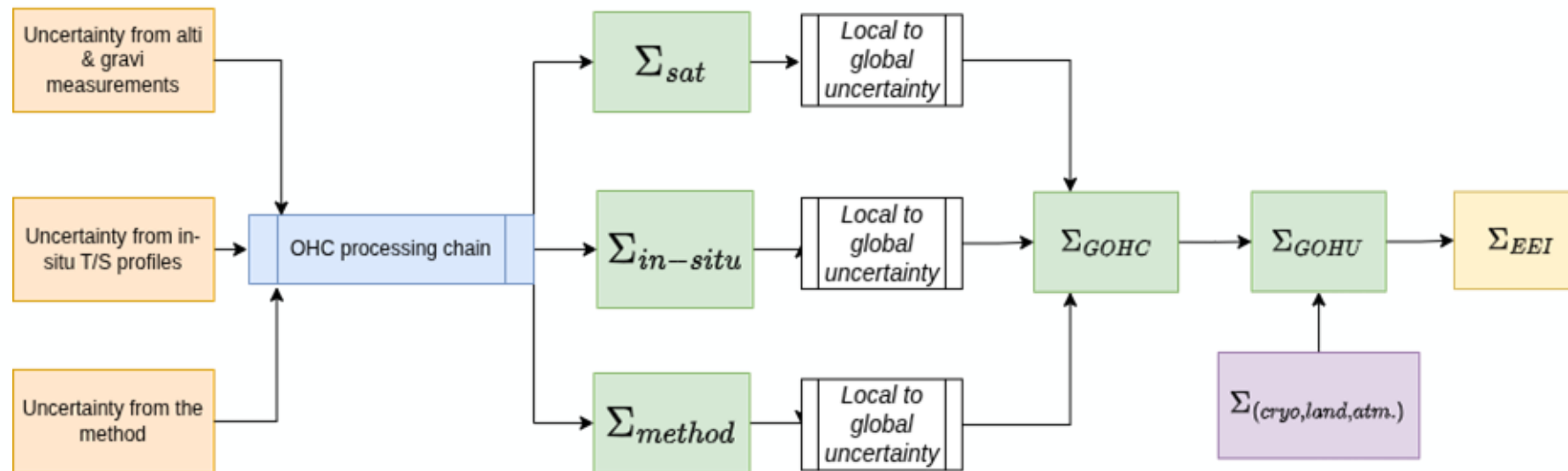
CCI CDR & auxiliary variable uncertainty → propagation to TAS uncertainty and error cov. estimates

**TAS uncertainty and error cov. structure → determine analysis best estimate combination of inputs**

Further trial analyses and then final analysis coming during 2026

## Uncertainties:

- Spatial and/or temporal structures of the errors in the EO data provided in the error covariance matrix.
- Error covariance matrix propagated from the EO data to the TAS and EEI, taking into account the effects due to the methods, formulations and approximations.
- Realistic uncertainties in the trend and acceleration, calculated using a formal mathematical approach (e.g. extended OLS or GLS).



Example of uncertainty propagation for EEI product

Paper produced within the MOTECUSOMA project:

- Review on knowledge on the EEI: Hakuba, M.Z., Fourest, S., Boyer, T., B. Meyssignac et al. Trends and Variability in Earth's Energy Imbalance and Ocean Heat Uptake Since 2005. Surv Geophys 45, 1721–1756 (2024).
- Data paper describing the strengths and weaknesses of the TAS product
- “Estimating Ocean Heat Content from Satellite Altimetry, Space Gravimetry, and In Situ Observations closes the Earth's Energy Budget on Interannual to Decadal Timescales” Fourest et al. 2026, submitted
- Scientific paper on Climate feedback & Climate sensitivity
- Scientific paper on the evolution of temperature over the Arctic Zone
- Scientific paper on the use of EEI index to assess the impact of mitigation policies