



# Satellite Application Facility in Support to Operational Hydrology and Water Management - Soil Moisture -

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<http://www.geo.tuwien.ac.at/>

# Overview

- EUMETSAT H-SAF
  - Satellites: METOP ASCAT
  - Federated ground segment: The SAF Network
- Soil moisture remote sensing – A vibrant research field
  - Basic principles
  - Sensors
- ASCAT surface soil moisture products
  - Service specifications
  - Data quality
- Value-added ASCAT soil moisture products

# EUMETSAT H-SAF

# SATELLITES

## SATELLITES

[CURRENT SATELLITES](#)
[FUTURE SATELLITES](#)
[PAST SATELLITES](#)
[LAUNCHES AND ORBITS](#)
[GROUND SEGMENT](#)
[SCIENCE ACTIVITIES](#)
[TECHNICAL DOCUMENTS](#)
[GLOSSARY](#)

## RELATED LINKS



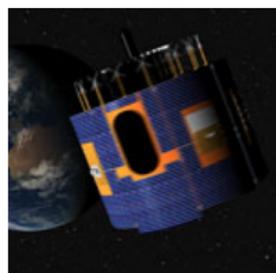
### SERVICE STATUS

Information on the status of our geostationary and polar satellites and the data.

[WMO OSCAR](#)

**EUMETSAT has been running a fleet of meteorological satellites, providing weather and climate data, for more than 25 years.**

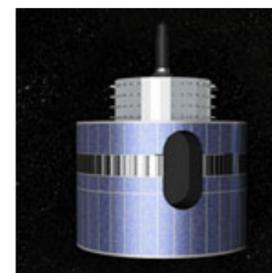
NAME	PERIOD	NUMBER
▶ <b>Meteosat First Generation (MFG)</b>	1977–2017	7 geostationary satellites
▶ <b>Meteosat Second Generation (MSG)</b>	2004–2025	4 geostationary satellites
▶ <b>Meteosat Third Generation (MTG)</b>	2021–2039	6 geostationary satellites
▶ <b>Metop</b>	2007–2024	3 polar satellites
▶ <b>EUMETSAT Polar System-Second Generation (EPS-SG)</b>	2021–2040	2 polar satellites
▶ <b>Jason</b>	2009–2036	3 marine satellites



CURRENT SATELLITES



FUTURE SATELLITES



PAST SATELLITES



GROUND SEGMENT

# METOP – EUMETSAT's Polar-Orbiting Satellites

## ■ METOP Satellite Series

- METOP-A
  - 19.10.2006
- METOP-B
  - 17.9.2012
- METOP-C
  - Planned for October 2018

## ■ METOP Second Generation

- 2 x 3 satellites
- First launches in 2021 and 2022

Artistic view of METOP





METOP Display at EUMETSAT in Darmstadt

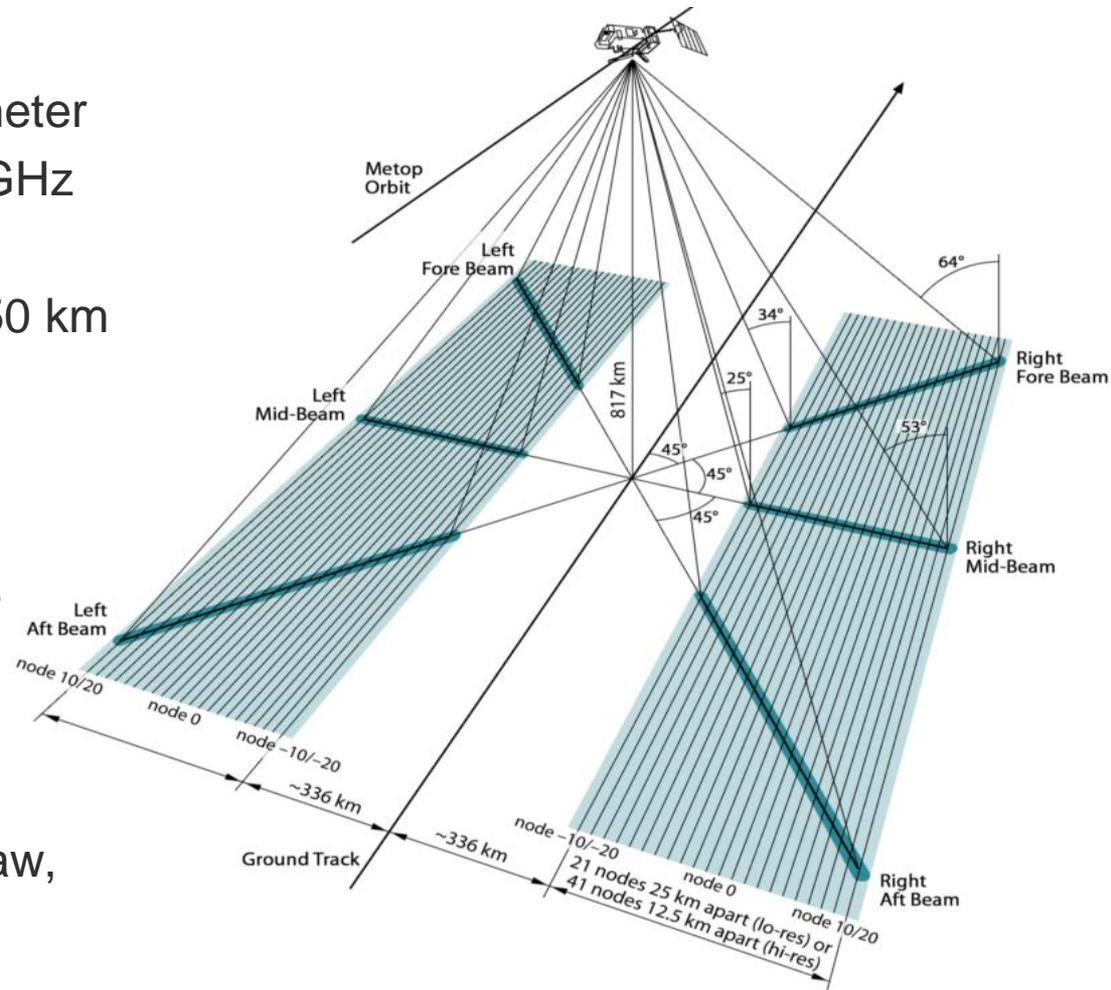
# METOP Advanced Scatterometer (ASCAT)

## ■ Sensor characteristics

- Active microwave scatterometer
- Frequency: C-band, 5.255 GHz
- Polarisation: VV
- Spatial Resolution: 25 km/ 50 km
- Antennas: 2 x 3
- Swath: 2 x 500 km
- Multi-incidence: 25-65°
- Daily global coverage: 82 %

## ■ Main applications

- Wind measurements, soil moisture, sea ice, freeze/thaw, vegetation dynamics



## SAFS

## SATELLITES

## CURRENT SATELLITES

## FUTURE SATELLITES

## PAST SATELLITES

## LAUNCHES AND ORBITS

## GROUND SEGMENT

## CENTRAL FACILITY

## MISSION CONTROL

## SAFS

## LAND SURFACE ANALYSIS

SUPPORT TO NOWCASTING  
AND VERY SHORT RANGE  
FORECASTING

## CLIMATE MONITORING

NUMERICAL WEATHER  
PREDICTIONATMOSPHERIC COMPOSITION  
MONITORINGSUPPORT TO OPERATIONAL  
HYDROLOGY AND WATER  
MANAGEMENT

## OCEAN AND SEA ICE

RADIO OCCULTATION  
METEOROLOGY

## Utilising specialist expertise from the Member States, Satellite Application Facilities (SAFs) are dedicated centres of excellence for processing satellite data. They form an integral part of the distributed EUMETSAT Application Ground Segment.

The eight EUMETSAT SAFs provide users with operational data and software products, each one for a dedicated user community and application area.

EUMETSAT Secretariat supervises and coordinates the overall activities of the SAF network, ensuring that the SAFs in operations are providing reliable and timely operational services related to the meteorological and environmental issues.

The SAF Network manages and coordinates interfaces between the SAFs themselves and between SAFs and other EUMETSAT systems, overseeing the integration and operations of SAFs into the overall ground segment infrastructure. During this process EUMETSAT ensures that services are delivered in the most reliable and cost-effective way.



### SAF NETWORK

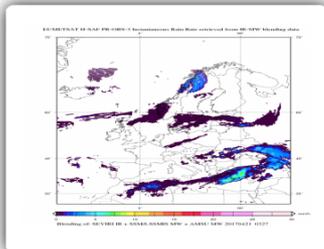
The [Product Navigator](#) has a full list of EUMETSAT centrally produced and SAF products.

### SATELLITE APPLICATION FACILITIES

The Satellite Application Facilities (SAFs) are a distributed network of thematic application facilities responsible for necessary research, development, and operational activities not carried out by the [central facility](#). The SAFs are located within the National Meteorological Services (NMS) of EUMETSAT Member States, or other agreed entities linked to a user community.

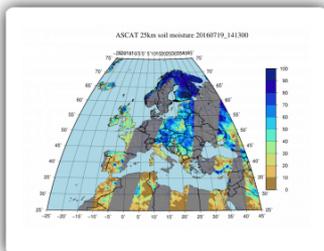
SAFs deliverables can be a specific piece of software to be made available to users for use in their own environment, or data and products made available in near real-time or offline.

The map below shows the deployment of the SAF Operational Architecture. Distributed architecture are indicated by a red dot, by selecting it, the relevant operational subsystem details will be displayed.



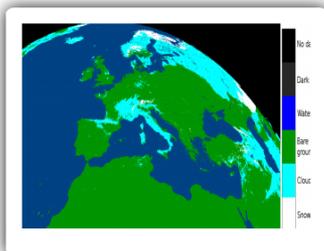
## PRECIPITATION

Images	Descriptions	Quality Monitoring	User Documents	Visiting Scientist	References
	<b>PR OBS 1 - H01</b> Precipitation rate at ground by MW conical scanners (with indication of phase)				
	<b>PR OBS 2 - H02</b> Precipitation rate at ground by MW cross-track scanners (with indication of phase)				
	<b>H02B-H03B-H05B-H15B-H17-H18-H23</b> H-SAF Full Disk Precipitation Products				
	<b>PR OBS 3 - H03</b> Precipitation rate at ground by GEO/IR supported by LEO/MW				
	<b>PR OBS 4 - H04</b> Precipitation rate at ground by LEO/MW supported by GEO/IR (with flag for phase)				
	<b>PR OBS 5 - H05</b> Accumulated precipitation at ground by blended MW and IR				
	<b>PR-OBS-6A - H15A</b> Blended SEVIRI Convection area / LEO MW Convective Precipitation				



## SOIL MOISTURE

Images	Descriptions	Quality Monitoring	User Documents	Visiting Scientist	References
	<b>SM OBS 2 - H08</b> Small scale surface soil moisture by radar scatterometer				
	<b>SM DAS 2 - H14</b> Profile Index in the roots region by scatterometer data assimilation				
	<b>H25-H108-H109-H110</b> Surface Soil Moisture ASCAT Data Record Time Series				
	<b>SM DAS 3 - H27</b> Soil Wetness Index in the roots region by ERS/SCAT and Metop ASCAT-A Scatterometer assimilation in a Land Data Assimilation System				
	<b>H101-H102-H16-H103</b> Surface Soil Moisture ASCAT A/B IIRT Orbit				



## SNOW

Images	Descriptions	Quality Monitoring	User Documents	Visiting Scientist	References
	<b>SN OBS 1 - H10</b> Snow detection (snow mask) by VIS/IR radiometry				
	<b>SN OBS 2 - H11</b> Snow status (dry/wet) by MW radiometry				
	<b>SN OBS 3 - H12</b> Effective snow cover by VIS/IR radiometry				
	<b>SN OBS 4 - H13</b> Snow water equivalent by MW radiometry				

# Soil Moisture Remote Sensing

# Soil Moisture

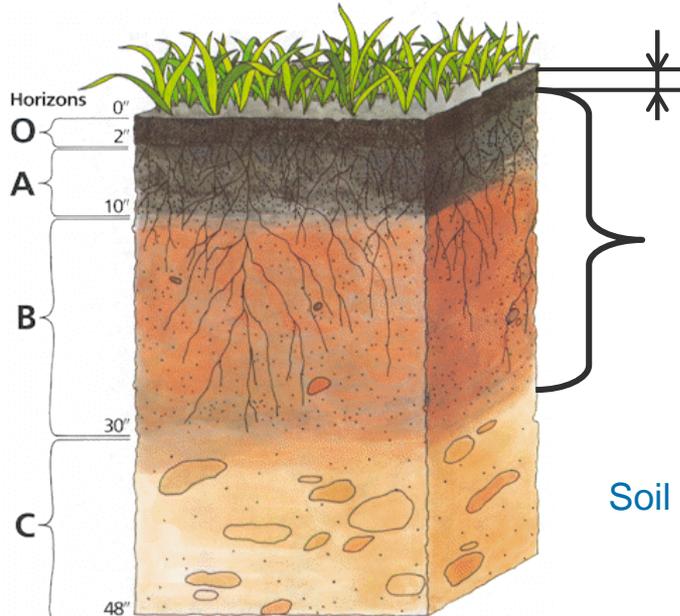
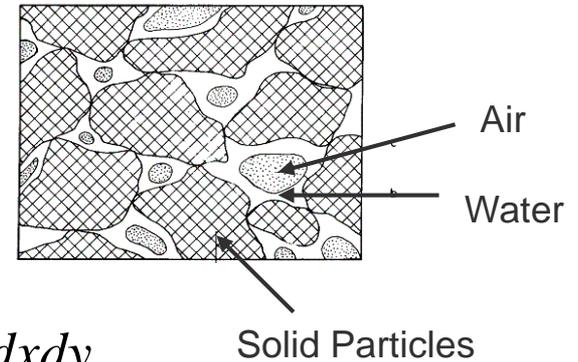
- Definition, e.g.

$$\theta = \frac{\text{Water Volume (m}^3\text{)}}{\text{Total Volume (m}^3\text{)}}$$

- Average

$$\langle \theta \rangle = \frac{1}{\text{Area} \cdot \text{Depth}} \int_{\text{Area}} \int_{\text{Depth}} \theta(x, y, z) dz dx dy$$

Cross-section of a soil



Thin, remotely sensed soil layer

Root zone: layer of interest for most applications

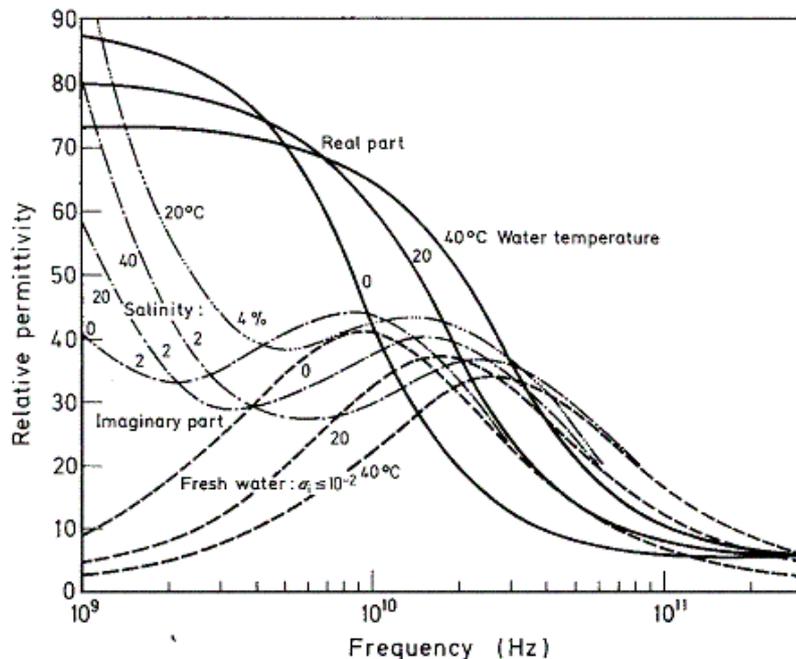
Soil profile

# Approaches to Remote Sensing of Soil Moisture

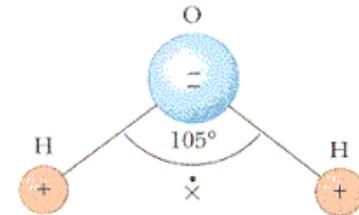
- Measurement principles
  - No direct measurement of  $\theta$  possible, only indirect techniques
- Optical to Mid-Infrared (0.4 – 3  $\mu\text{m}$ )
  - Change of “colour”
  - Water absorption bands at 1.4, 1.9 and 2.7  $\mu\text{m}$
- Thermal Infrared (7-15  $\mu\text{m}$ )
  - Indirect assessment of soil moisture through its effect on the surface energy balance (temperature, thermal inertia, etc.)
- **Microwaves (1 mm – 1 m)**
  - **Change of dielectric properties**

# Microwaves

- Microwaves (1 mm – 1 m wavelength)
  - All-weather, day-round measurement capability
  - Very sensitive to soil water content below relaxation frequency of water (< 10 GHz)
  - Penetrate vegetation and soil to some extent
    - Penetration depth increases with wavelength



Dielectric constant of water



The **dipole moment** of water molecules causes “orientational polarisation”, i.e. a high dielectric constant

# Measurement Principle

- Microwaves are highly sensitive to soil moisture due to the distinct dielectric properties of liquid water
- Observables
  - **Passive sensors:** Brightness temperature  $T_B = e \times T_s$  where  $e$  is the emissivity and  $T_s$  is the surface temperature
  - **Active sensors:** Backscattering coefficient  $\sigma^0$ ; a measure of the reflectivity of the Earth surface
- Active measurements are somewhat more sensitive to roughness and vegetation structure than passive measurements, but
  - are not affected by surface temperature (above 0°C)
  - have a much better spatial resolution
- Despite these differences both active and passive sensors measure essentially the same variables:
  - Passive and active methods are interrelated through **Kirchhoff's law**:
    - $e = 1 - r$  where  $r$  is the reflectivity

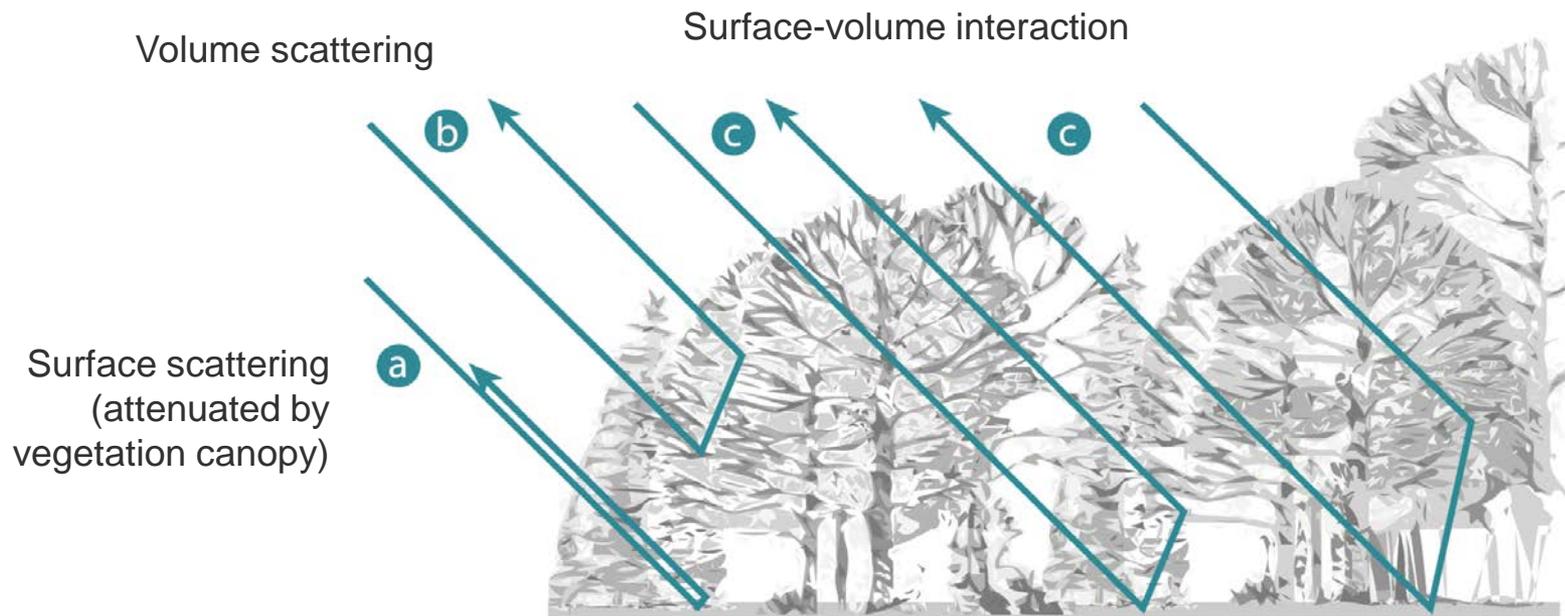
# Microwave Satellites used for Soil Moisture Retrieval



# Backscatter from Vegetated Surfaces

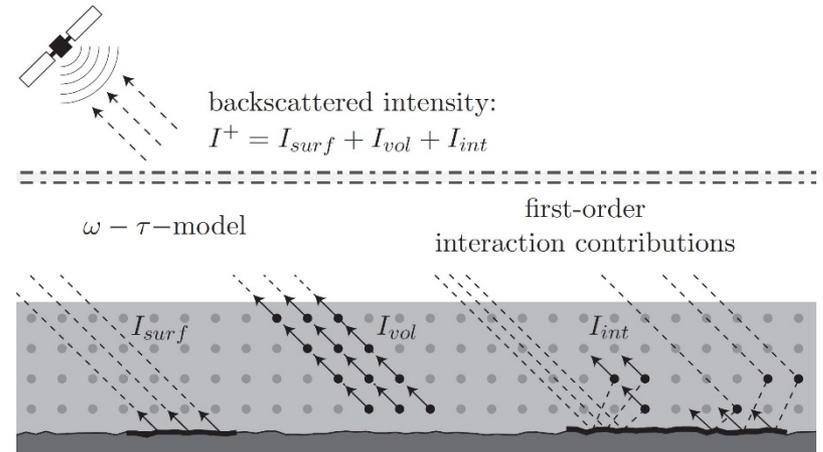
- Except for dense forest canopies, backscatter from vegetation is due to surface-, volume- and multiple scattering

$$\sigma_{total}^0 = \sigma_{volume}^0 + \sigma_{surface}^0 + \sigma_{interaction}^0$$



# Theoretical Backscatter Models

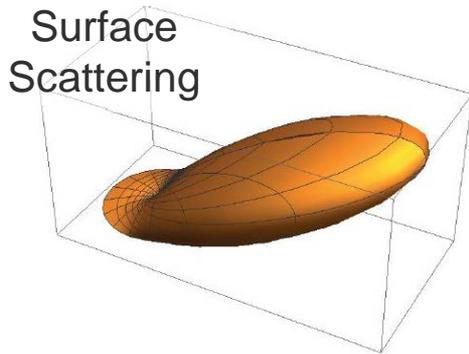
- Radiative transfer theory
- Modelling of bi-static scattering
  - Mono-static backscatter as a special (simple) case
- Generalised phase functions for modelling surface-volume interactions



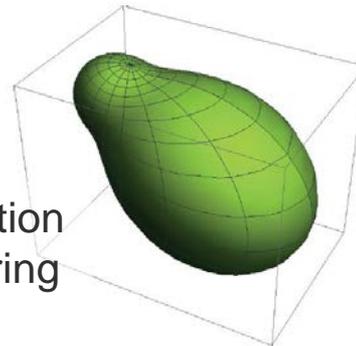
## Exemplary Phase Functions

for

Surface  
Scattering

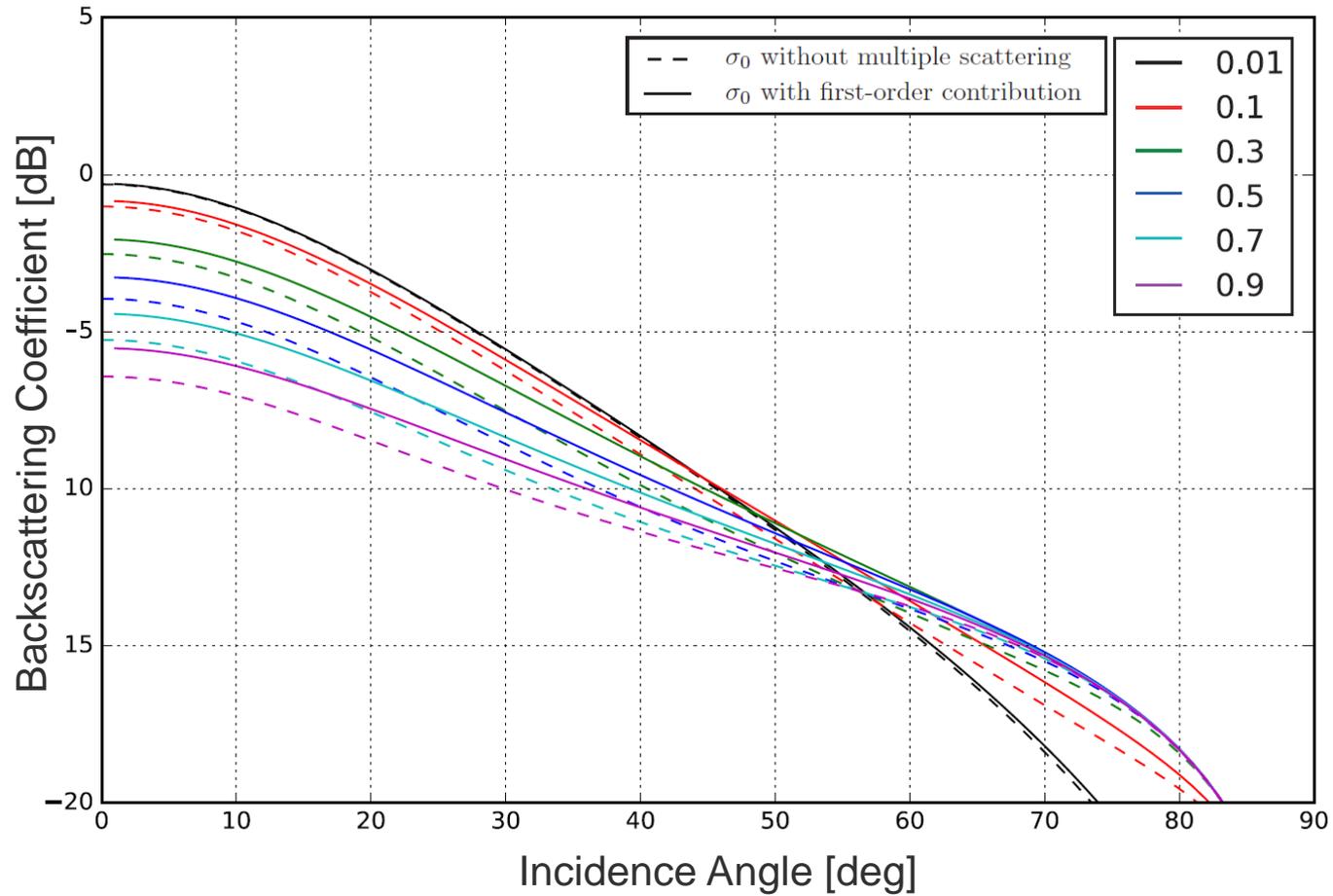


Vegetation  
Scattering



Quast, R., W. Wagner (2016) An analytical solution for first-order scattering in bistatic radiative transfer interaction problems of layered media, Applied Optics, 55(20), 5379-5386.

# Model Simulations



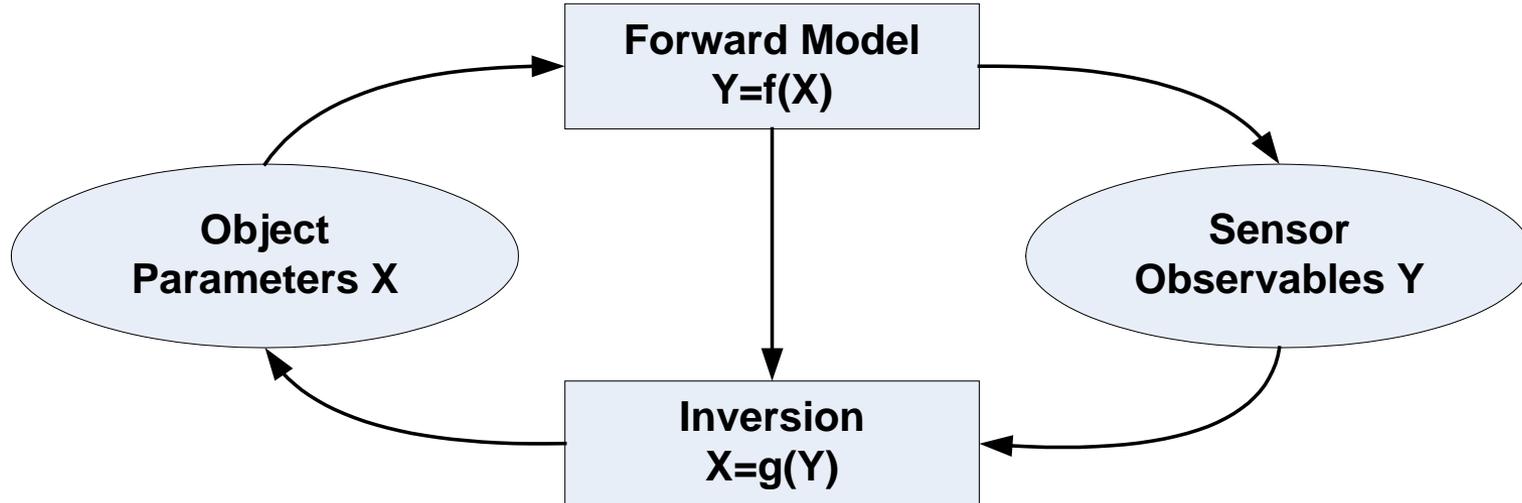
↓  
Vegetation  
Optical Depth  
increases

# Soil Moisture Retrieval

Empirical models

Semi-empirical models

Theoretical models



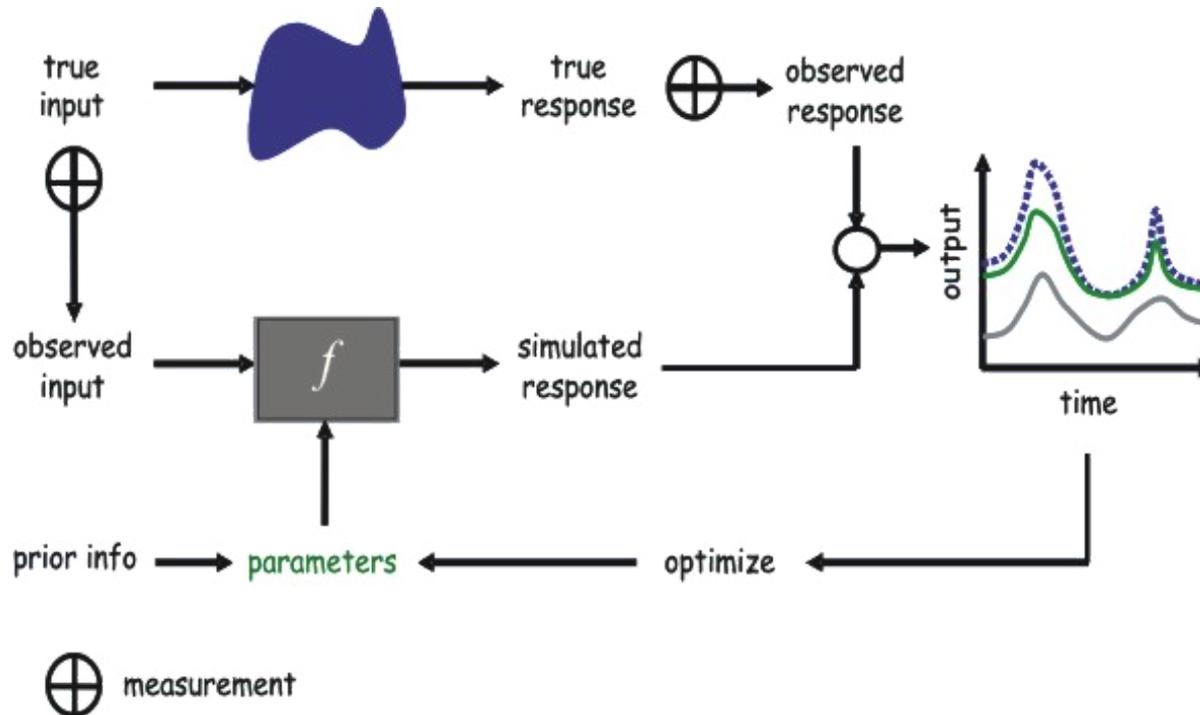
Lookup tables and neural networks

Least-square matching

Direct inversion

# Why Model Calibration is Needed

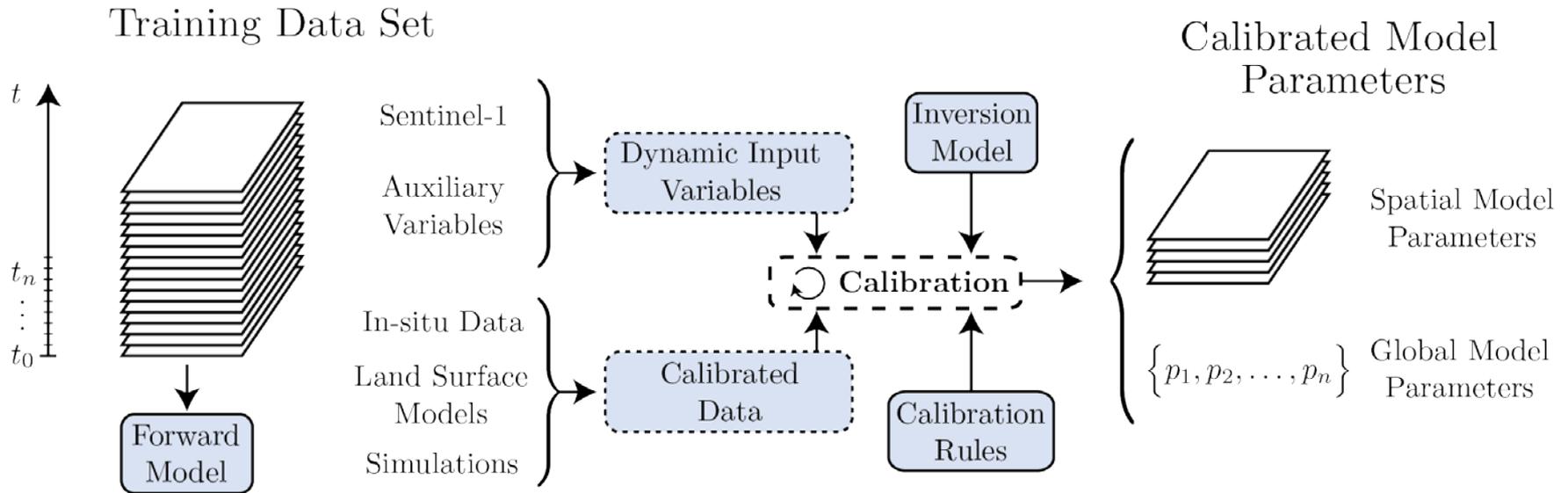
- No model is all-encompassing → Calibration is needed



*“All natural systems models are to some degree lumped, and use effective parameters to characterize these spatial-temporal processes.”*

Jasper Vrugt <http://math.lanl.gov/~vrugt/research/>

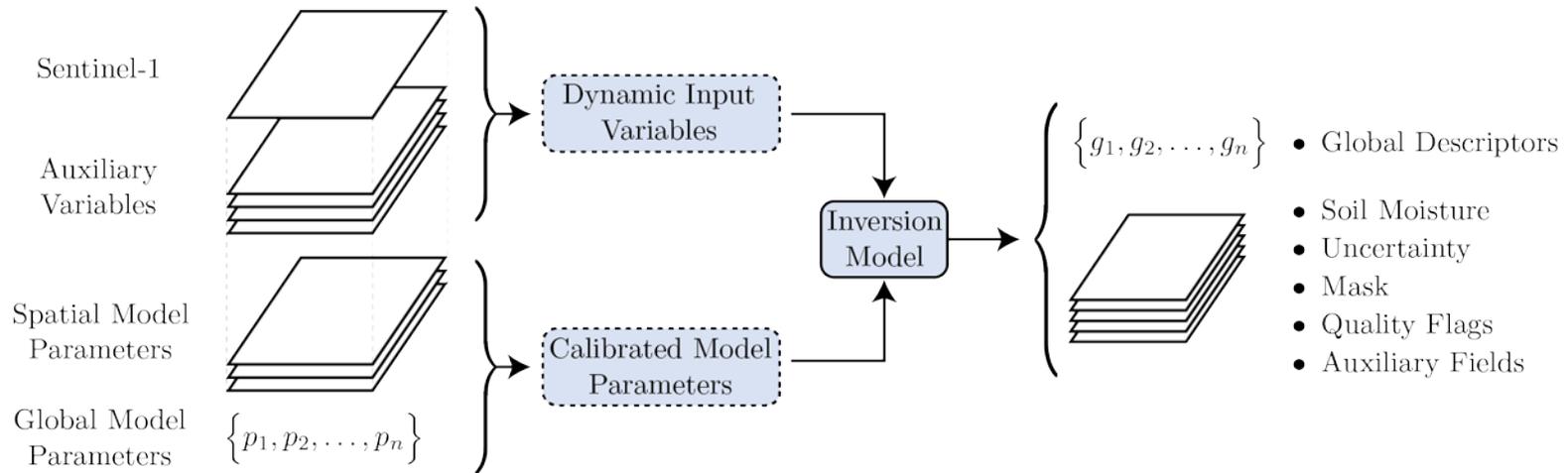
# Calibration Procedure



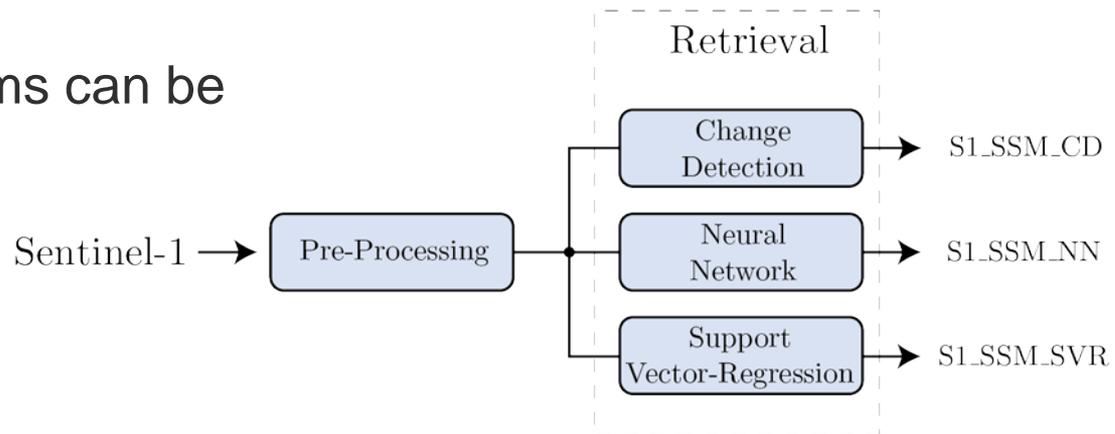
- The TU Wien processing architecture allows for calibration
  - *Per-pixel calibration* is done - as far as possible - just based on historic satellite time series
  - Auxiliary data are used for calibrating *model parameters*

# Retrieval Procedure

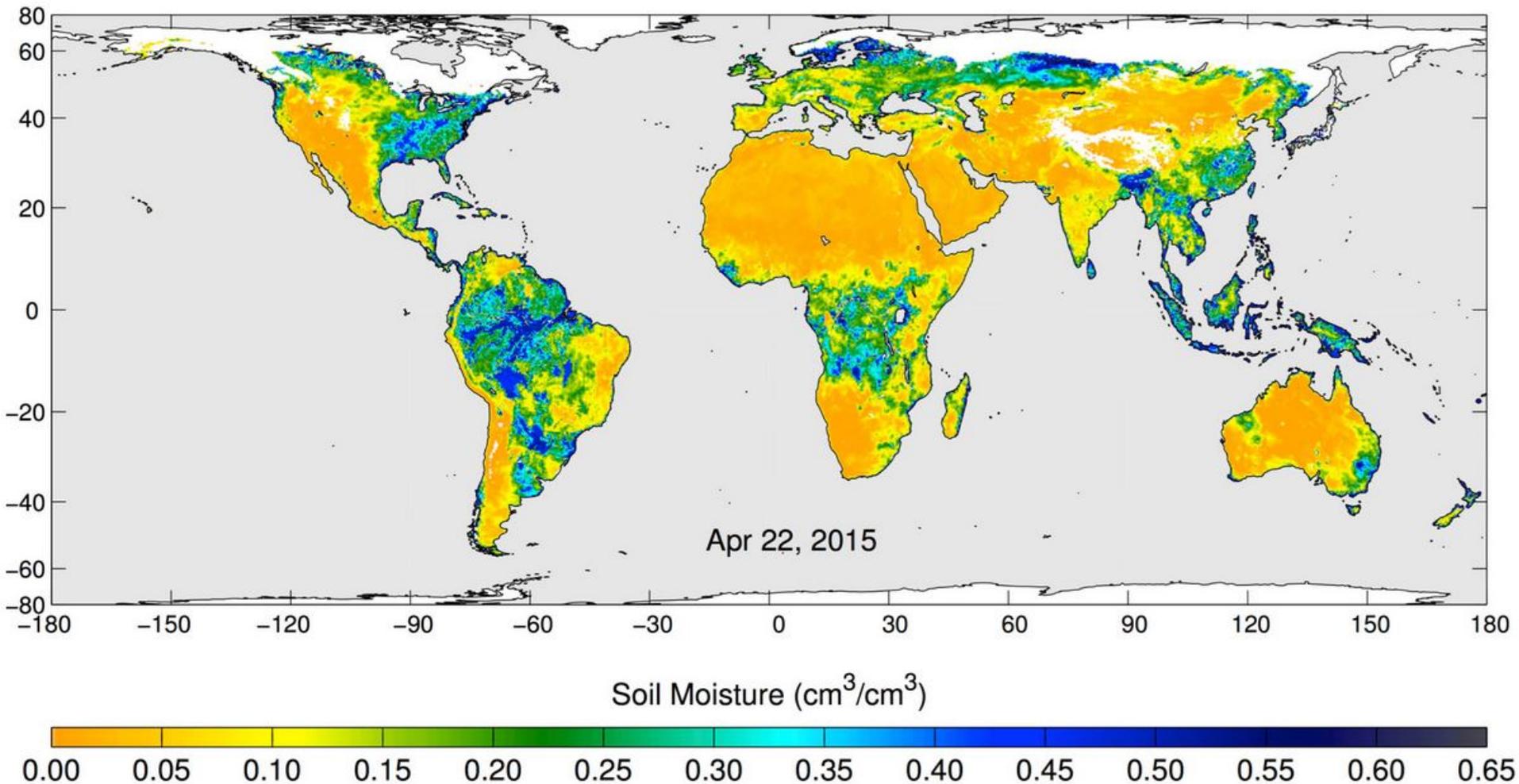
- Retrieval can be performed in near-real-time and off-line



- Several algorithms can be used in parallel



# SMAP Soil Moisture Image



Composite of three days of SMAP radiometer data, centered on April 22, 2015. White areas indicate snow, ice or frozen ground. From <https://smap.jpl.nasa.gov/resources/87/>.

# Limitations & Caveats

- Soil moisture retrieval is not possible over
  - Urban areas, concrete and rock
  - Water bodies and inundation
  - Frozen or snow covered soil
  - Under forests and dense shrubs
- Soil moisture data quality varies in space and time because of
  - Vegetation water content and structure
  - Sub-surface scattering in dry areas
  - Topographic effects
  - Temperature dependency (for passive only)
- Data quality described using uncertainty estimates (from error propagation) and advisory flags

# Information Content



# Signal versus Noise

- The information content of soil moisture is in our view best characterised by the **signal-to-noise ratio (SNR)**
  - Key criterion in data assimilation
- **Signal** is tied to a certain scale
  - **Noise** refers to random instrument noise as well as representativeness errors
  - SNR is scale dependent
- Soil moisture scaling approaches
  - Highly non-linear hydrological processes are assumed to linearize at coarse satellite scales
  - Standard error model

$$\hat{\Theta} = \alpha + \beta(\Theta + \varepsilon)$$

$\hat{\Theta}$  ... Satellite retrieval or model soil moisture

$\Theta$  ... "true" soil moisture state

$\alpha, \beta$  ... linear parameters

$\varepsilon$  ... residual error

# Triple Collocation

- Originally proposed to estimate **random error variances**
  - Covariance-formulation

Assumptions:

$$\hat{\Theta}_X = \alpha_X + \beta_X (\Theta + \varepsilon_X)$$

$$\hat{\Theta}_Y = \alpha_Y + \beta_Y (\Theta + \varepsilon_Y)$$

$$\hat{\Theta}_Z = \alpha_Z + \beta_Z (\Theta + \varepsilon_Z)$$

$$\text{Cov}(\Theta, \varepsilon_i) = 0$$

$$\text{Cov}(\varepsilon_i, \varepsilon_j) = 0$$

$$i, j \in \{X, Y, Z\}$$

$$i \neq j$$

$$\text{Var}(\hat{\Theta}_i) = \beta_i^2 \text{Var}(\Theta) + \beta_i^2 \text{Var}(\varepsilon_i)$$

$$\text{Cov}(\hat{\Theta}_i, \hat{\Theta}_j) = \beta_i \beta_j \text{Var}(\Theta)$$

Error variances:

$$\beta_X \text{Var}(\varepsilon_X) = \text{Var}(\hat{\Theta}_X) - \frac{\text{Cov}(\hat{\Theta}_X, \hat{\Theta}_Y) \text{Cov}(\hat{\Theta}_X, \hat{\Theta}_Z)}{\text{Cov}(\hat{\Theta}_Y, \hat{\Theta}_Z)}$$

$$\beta_Y \text{Var}(\varepsilon_Y) = \text{Var}(\hat{\Theta}_Y) - \frac{\text{Cov}(\hat{\Theta}_Y, \hat{\Theta}_X) \text{Cov}(\hat{\Theta}_Y, \hat{\Theta}_Z)}{\text{Cov}(\hat{\Theta}_X, \hat{\Theta}_Z)}$$

$$\beta_Z \text{Var}(\varepsilon_Z) = \text{Var}(\hat{\Theta}_Z) - \frac{\text{Cov}(\hat{\Theta}_Z, \hat{\Theta}_X) \text{Cov}(\hat{\Theta}_Z, \hat{\Theta}_Y)}{\text{Cov}(\hat{\Theta}_X, \hat{\Theta}_Y)}$$

Scaling coefficients:

$$\beta_X = 1$$

$$\beta_Y^X = \frac{\text{Cov}(\hat{\Theta}_X, \hat{\Theta}_Z)}{\text{Cov}(\hat{\Theta}_Y, \hat{\Theta}_Z)}$$

$$\beta_Z^X = \frac{\text{Cov}(\hat{\Theta}_X, \hat{\Theta}_Y)}{\text{Cov}(\hat{\Theta}_Z, \hat{\Theta}_Y)}$$

Stoffelen, A. (1998). Toward the true near-surface wind speed: Error modeling and calibration using triple collocation. *Journal of Geophysical Research: Oceans* (1978–2012), 103(C4), 7755-7766.

# Triple Collocation

- Recently extended to estimate the **signal-to-noise ratio**

$$\text{SNR}_X = \frac{\text{Var}(\Theta)}{\text{Var}(\varepsilon_i)} = \frac{1}{\frac{\text{Var}(\hat{\Theta}_X) \text{Cov}(\hat{\Theta}_Y, \hat{\Theta}_Z)}{\text{Cov}(\hat{\Theta}_X, \hat{\Theta}_Y) \text{Cov}(\hat{\Theta}_X, \hat{\Theta}_Z)} - 1} \quad \begin{array}{l} i, j, k \in \{X, Y, Z\} \\ i \neq j \neq k \end{array}$$

Draper, C., Reichle, R., de Jeu, R., Naeimi, V., Parinussa, R., & Wagner, W. (2013). Estimating root mean square errors in remotely sensed soil moisture over continental scale domains. *Remote Sensing of Environment*, 137, 288-298.

McColl, K. A., Vogelzang, J., Konings, A. G., Entekhabi, D., Piles, M., & Stoffelen, A. (2014). Extended triple collocation: Estimating errors and correlation coefficients with respect to an unknown target. *Geophysical Research Letters*.

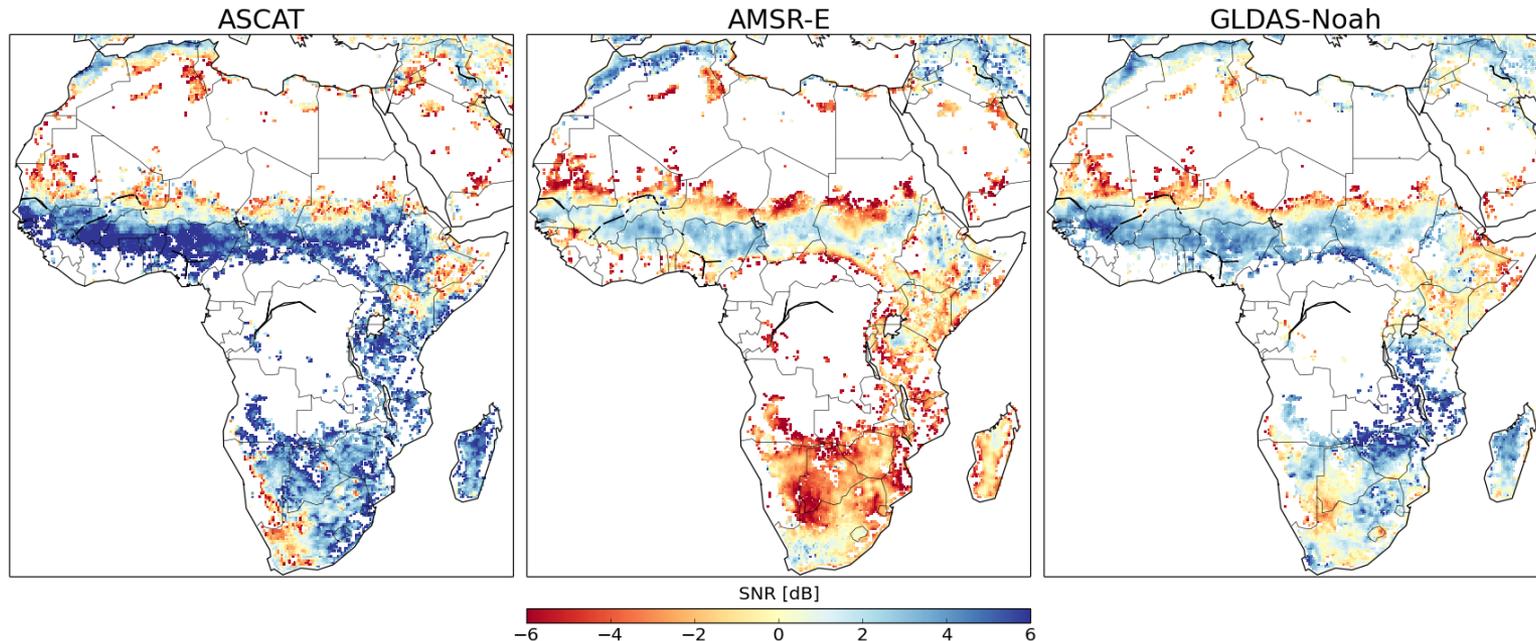
# Signal to Noise Ratio

- More easy interpretability when expressed in **decibel** units

$$\text{SNR}_i[\text{dB}] = 10 \log \frac{\text{Var}(\Theta)}{\text{Var}(\varepsilon_i)}$$

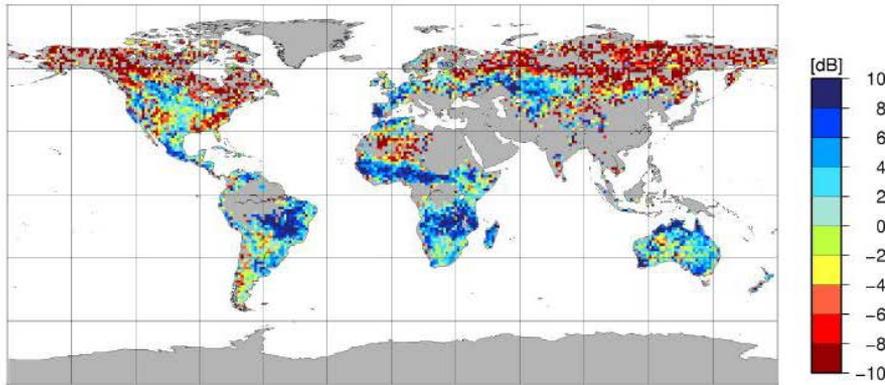
0 dB: signal variance = noise variance

+/- 3 dB: signal variance = double / half noise variance

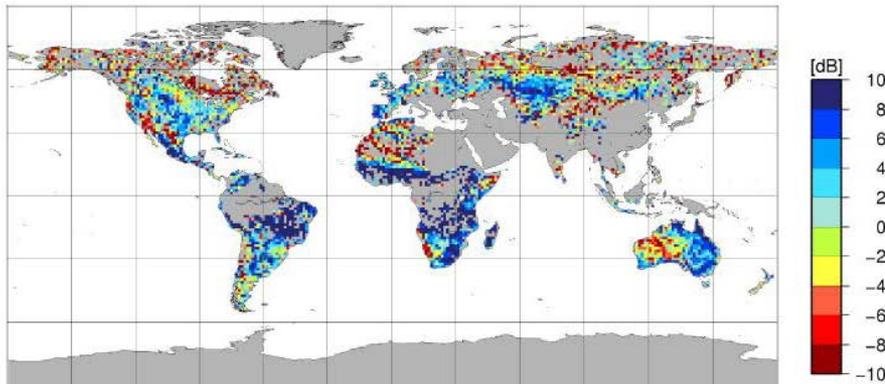


Gruber, A., C. H. Su, S. Zwieback, W. Crow, W. Dorigo, W. Wagner (2016) Recent advances in (soil moisture) triple collocation analysis, International Journal of Applied Earth Observation and Geoinformation, 45, 200-211.

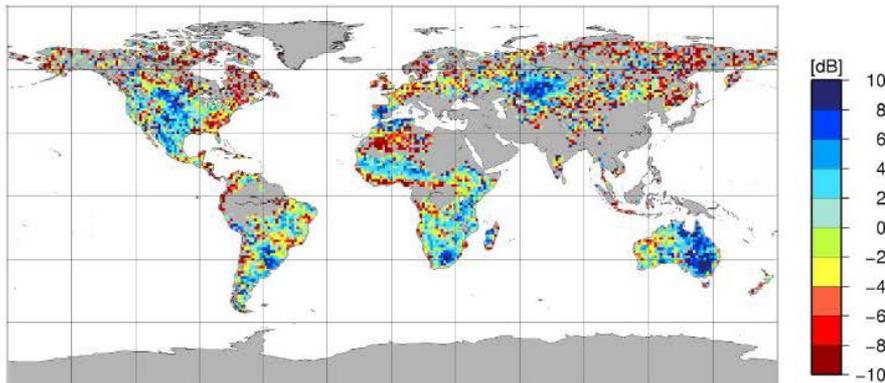
SNR of ERA-Interim



SNR of ASCAT

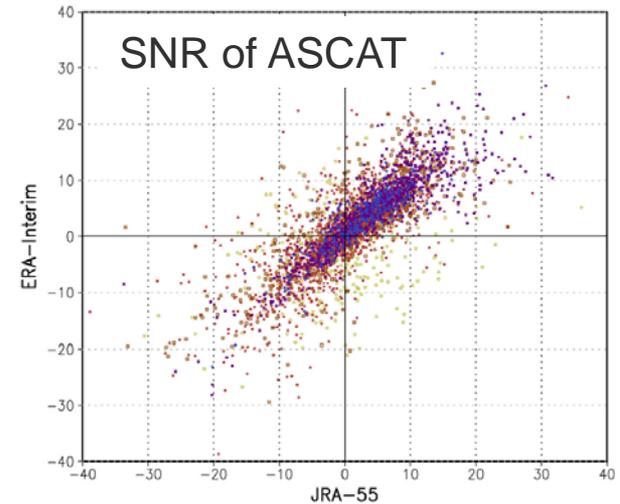


SNR of SMOS



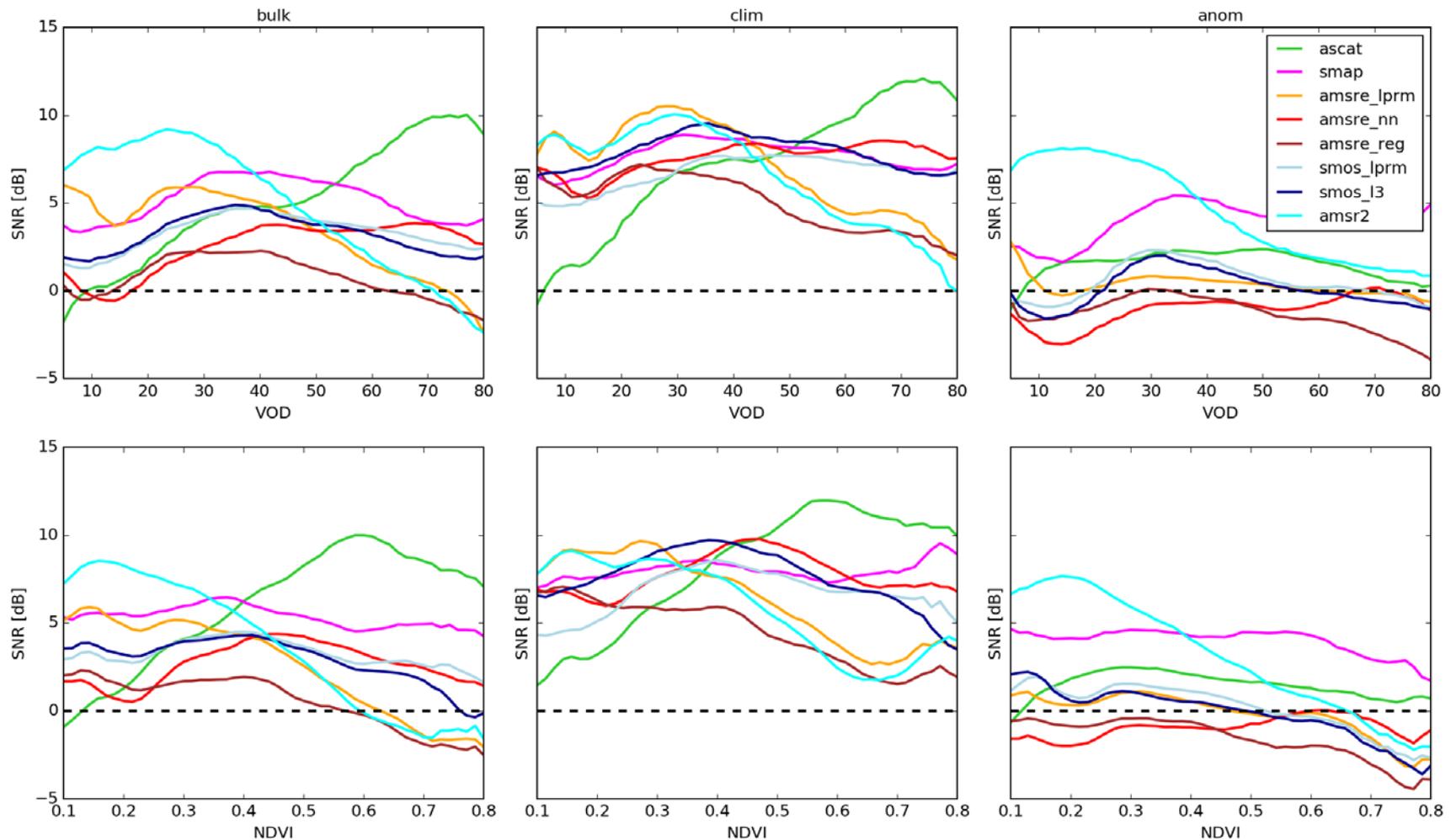
## SNR of ASCAT & SMOS

- SNR can be estimated with a large number of triplets
- Results are robust against exchange of model reference



Miyaoka et al. (2017) Triple collocation analysis of soil moisture from Metop-A ASCAT and SMOS against JRA-55 and ERA-Interim. IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, in press.

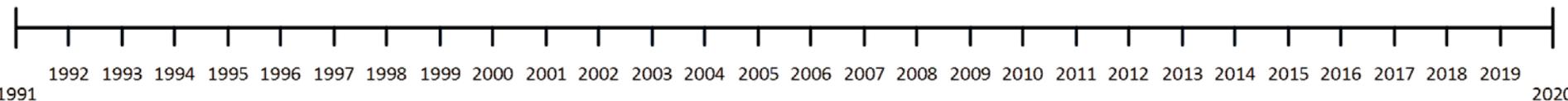
# SNR as a Function of Vegetation



Comparison of SNR for original soil moisture data sets (left), their climatology (middle) and anomalies (right). Unpublished preliminary results prepared by Alexander Gruber.

# ASCAT Surface Soil Moisture

# Scatterometer Soil Moisture Research



1991-2000: ERS-1 SCAT

1995-2011: ERS-2 SCAT

2006 up to present: METOP-A ASCAT

2012 up to present: METOP-B ASCAT

TU Wien  
Research Emphases

1994-1998: First Soil Moisture Studies

1999: TU Wien Algorithm published

2000-2003: Roll-out on Global Scale

2004-2008: Algorithm Adaption to ASCAT

2009-2016: Advanced Error Characterisation Techniques

2012-2020: Model Calibration

2002: First Global Soil Moisture Data Set Published

2008: First Near-Real-Time Soil Moisture Data Service

2012: First Multi-Satellite Soil Moisture Climate Data Record

Partners & Users  
Research Highlights

2004: Comparison to Global Soil Moisture Wetness Project

2009-2012: Satellite Inter-comparison Studies

2014: Retrieval Method Inter-comparison

2006: Validation Study by Meteo-France

Since 2008: Data Assimilation Studies at Met Services

Since 2008: Triple Colocation Applied to Soil Moisture

Since 2010: Hydrology Studies by External Teams



# NRT 25 km ASCAT SM Service



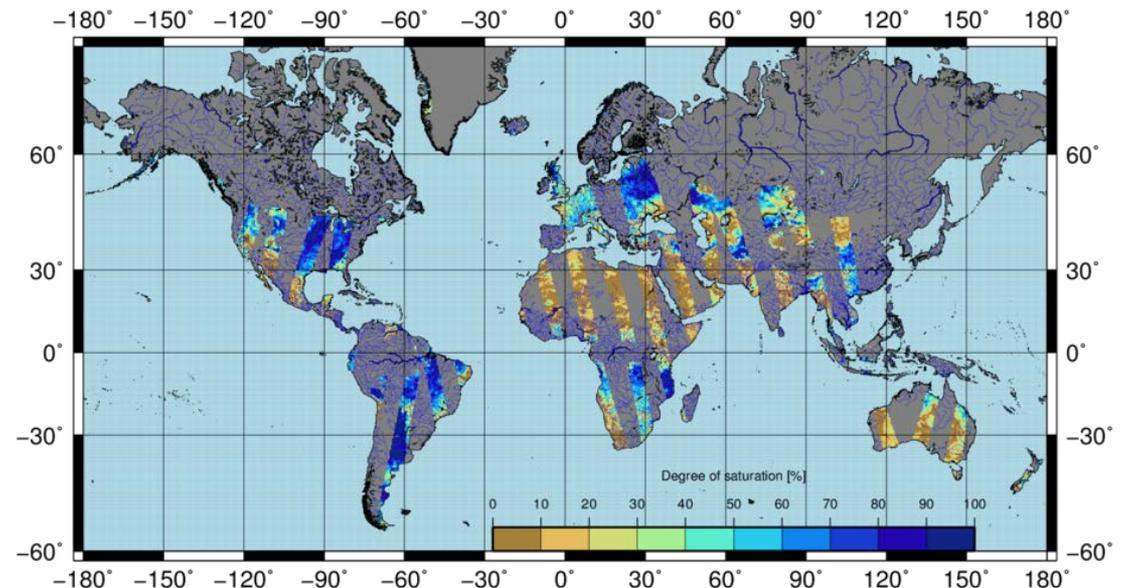
## ■ Evolution

- Was initiated as Day 2 product outside H-SAF on recommendation by Met Office and ECMWF to implement it more quickly, starting operations in 2008
- Was brought into H-SAF in 2012 (start of CDOP2)

## ■ Roles

- EUMETSAT
  - NRT operations
- TU Wien
  - R&D
  - Delivery of model parameters for NRT processor
- ZAMG
  - NRT testing chain

ASCAT soil moisture 20170404\_0210, Metop-A, 125



# H SAF Surface Soil Moisture Products

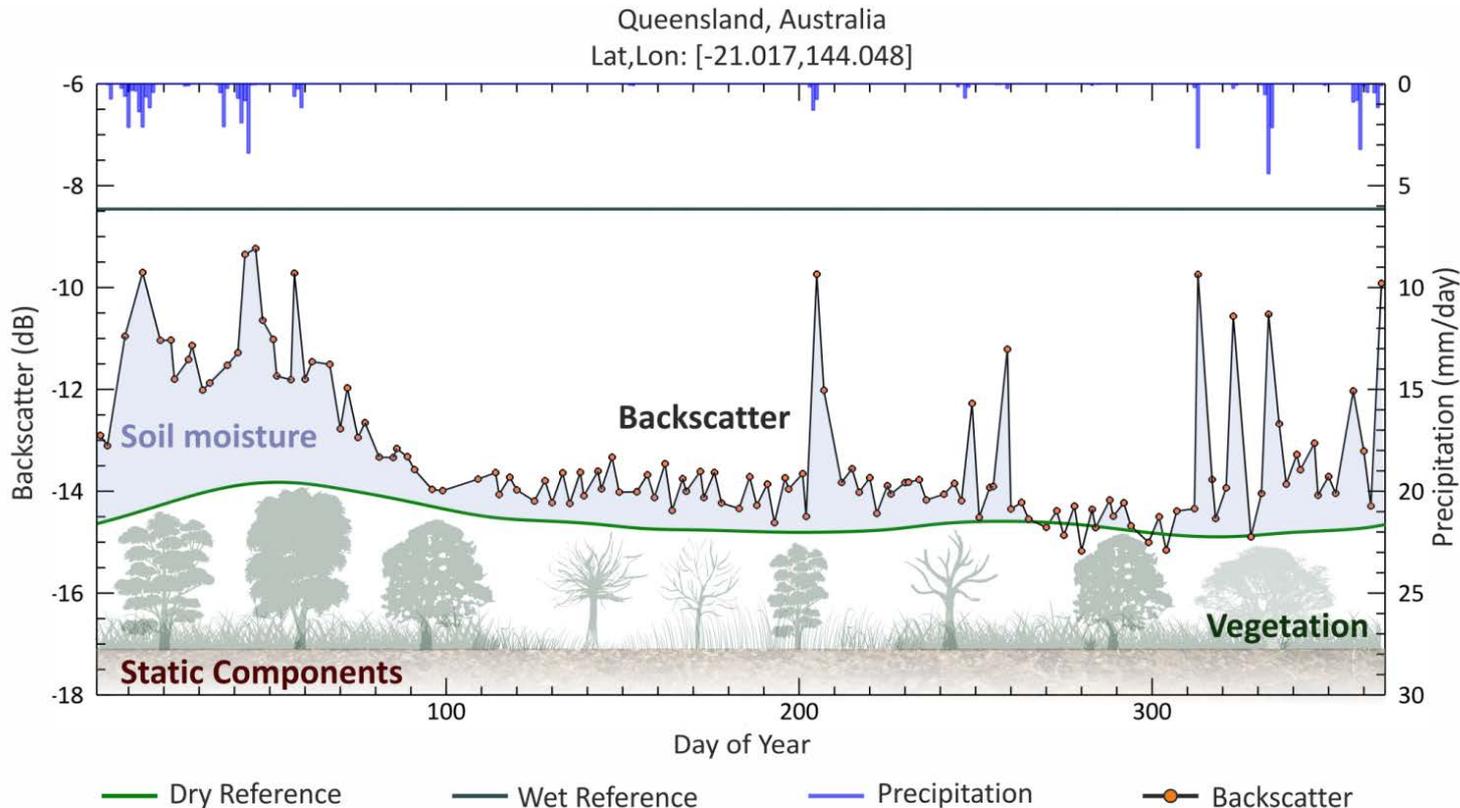
- Near Real-time (NRT) products
  - H16, H101-H103 are official H SAF soil moisture products **produced by EUMETSAT** (and re-distributed by H SAF under a different file name via FTP)
    - H101: Metop-A ASCAT soil moisture at 12.5 km sampling
    - H102: Metop-A ASCAT soil moisture at 25 km sampling
    - H16: Metop-B ASCAT soil moisture at 12.5 km sampling
    - H103: Metop-B ASCAT soil moisture at 25 km sampling
  - H08 Disaggregated Metop ASCAT NRT SSM at 1 km – Pre-operational
  
- Data records (DR)
  - H25: Metop ASCAT DR2015 SSM time series 12.5 km sampling – Released
  - H109: Metop ASCAT DR2016 SSM time series 12.5 km sampling – Released
  - H111: Metop ASCAT DR2017 SSM time series 12.5 km sampling – Under review
  - H113: Metop ASCAT DR2018 SSM time series 12.5 km sampling – Processed in Jan 2018
  
- Offline products (regular extensions to data records)
  - H108: Metop ASCAT DR2015 EXT SSM time series 12.5 km sampling – Operational
  - H110: Metop ASCAT DR2016 EXT SSM time series 12.5 km sampling – Under review

# TU Wien Change Detection Approach

- Formulated in 1996-98 out of the need to circumvent the lack of adequate backscatter models

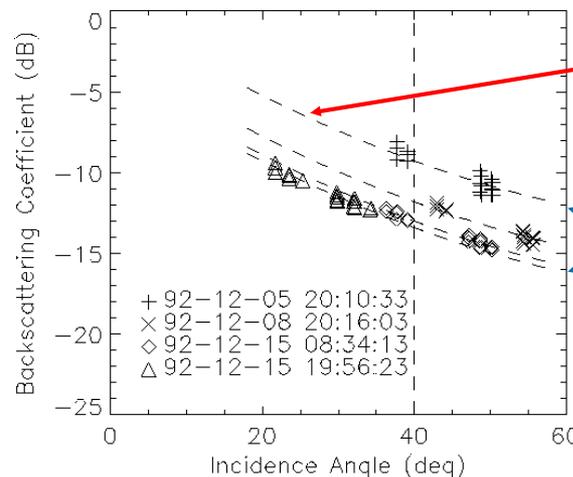
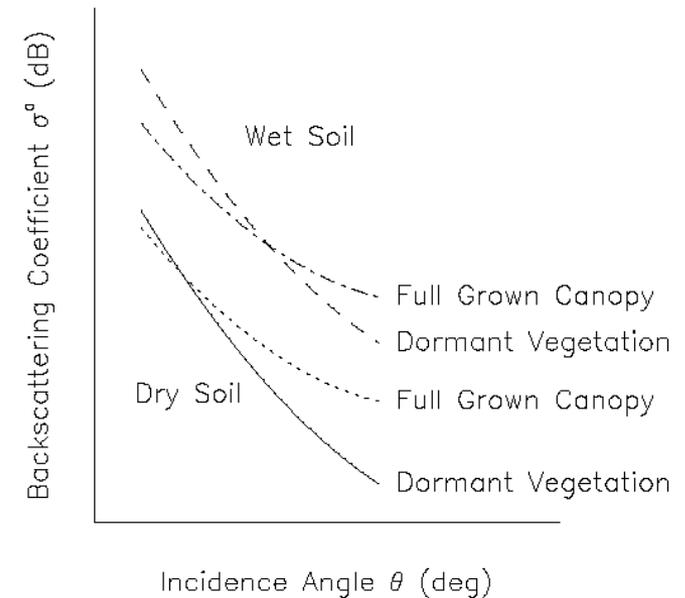
- Accounts indirectly for surface roughness and land cover

$$m_s(t) = \frac{\sigma^0(t) - \sigma_{dry}^0(t)}{\sigma_{wet}^0(t) - \sigma_{dry}^0(t)}$$



# TU Wien Backscatter Model

- Motivated by physical models and empirical evidence
  - Formulated in decibels (dB) domain
  - Linear relationship between backscatter (in dB) and soil moisture
  - Empirical description of incidence angle behaviour
  - Seasonal vegetation effects cancel each other out at the "cross-over angles"
    - dependent on soil moisture



Incidence angle behaviour is determined by vegetation and roughness roughness

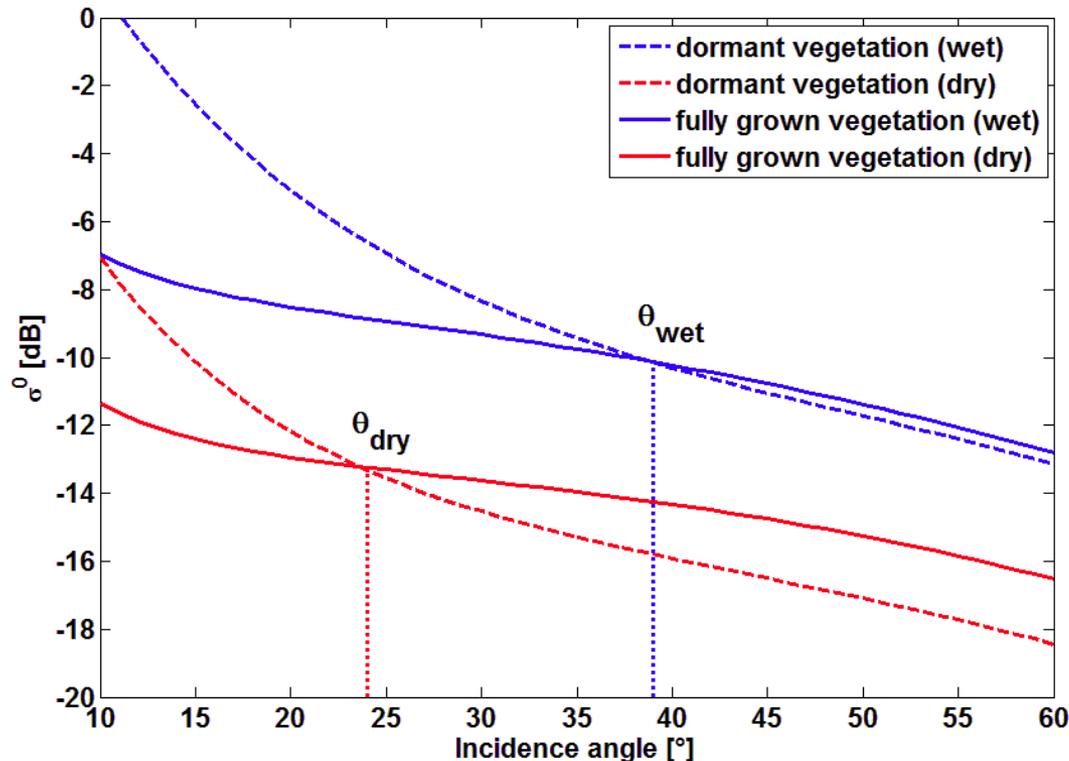
Changes due to soil moisture variations

ERS Scatterometer measurements

# Functional Behaviour

- The TU Wien backscatter model mimics a semi-empirical backscatter model with a strong surface-volume interaction term

$$\sigma^0 = (1 - f_{nt}) \left[ \frac{\omega_{tr} \cos \theta}{2} \left( 1 - e^{-\frac{2\tau_{tr}}{\cos \theta}} \right) + \sigma_s^0(\theta) e^{-\frac{2\tau_{tr}}{\cos \theta}} + 2\chi R_0 \omega_{tr} \tau_{tr} e^{-\frac{2\tau_{tr}}{\cos \theta}} \right] + f_{nt} \frac{\omega_{nt} \cos \theta}{2}$$

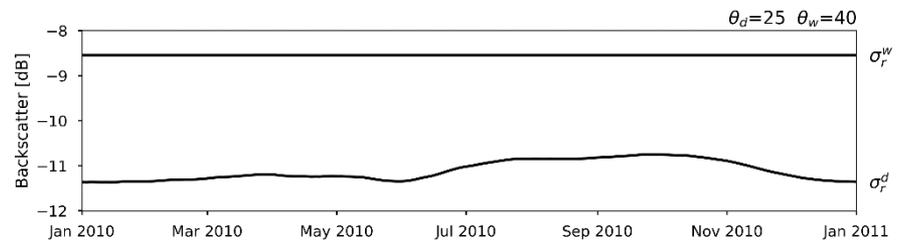
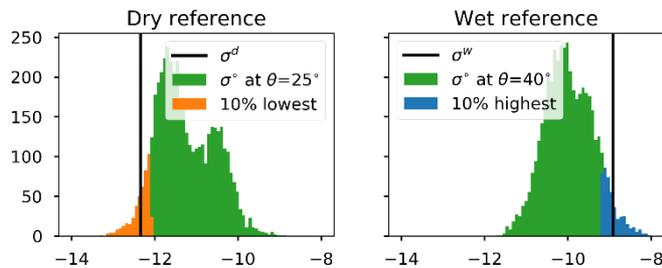
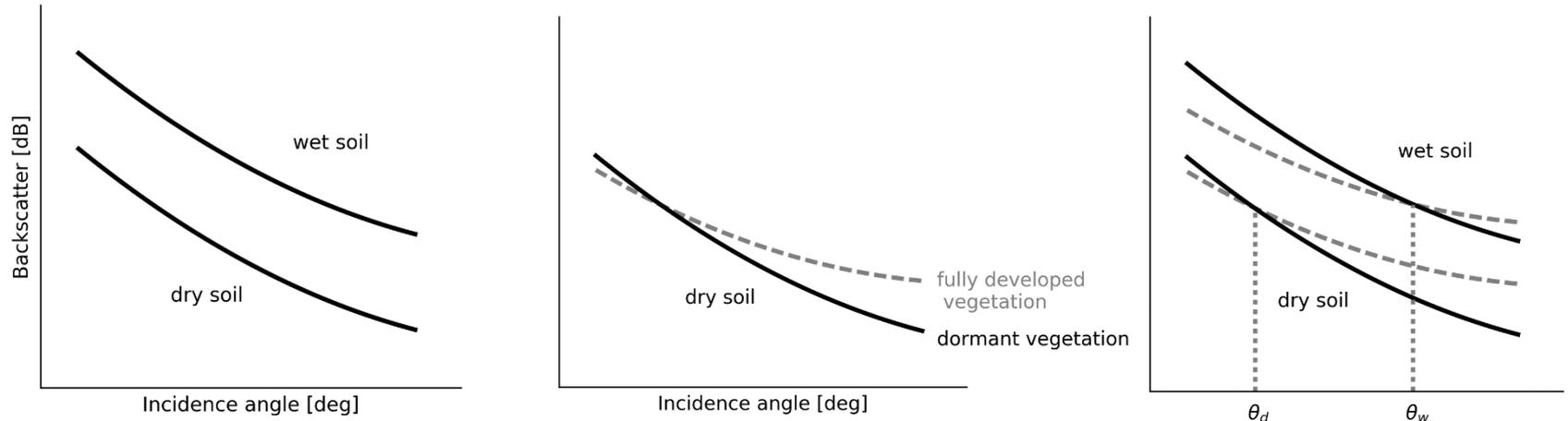


Mixing model with fraction of non-transparent ( $nt$ ) and transparent ( $tr$ ) vegetation

Bare soil scattering  $\sigma_s^0(\theta)$  modelled with Improved Integral Equation Method I<sup>2</sup>EM

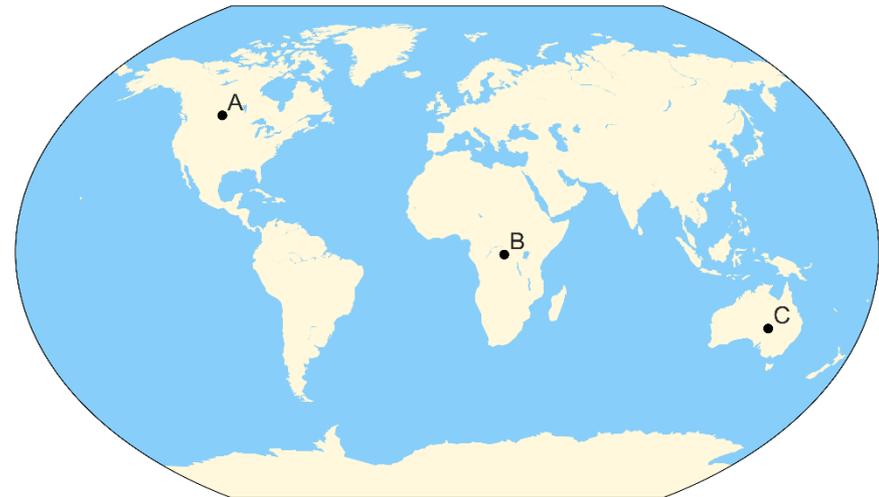
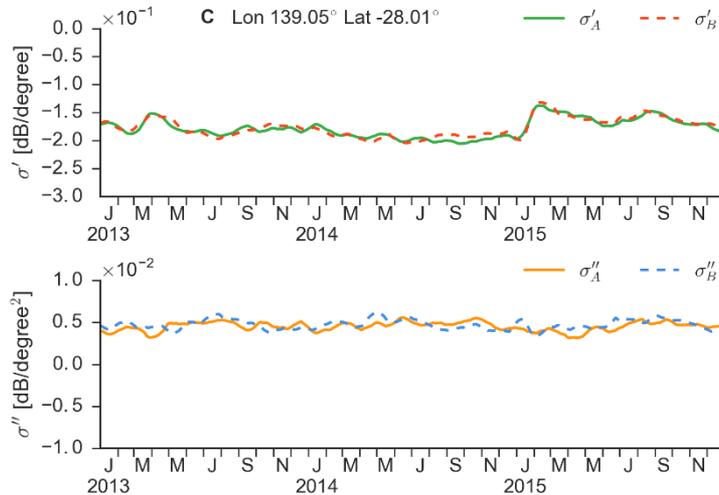
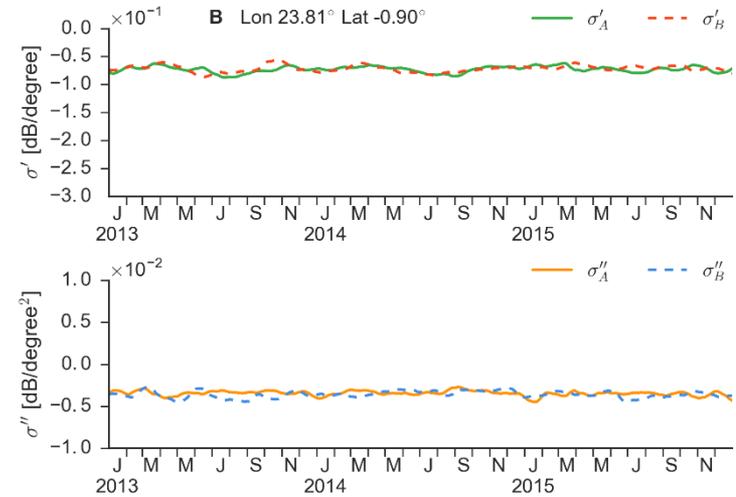
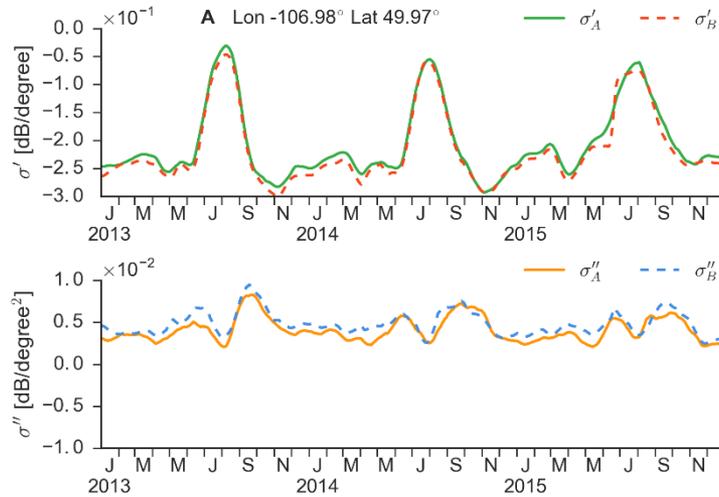
Interaction term enhances soil moisture contributions

# Cross-over angle concept

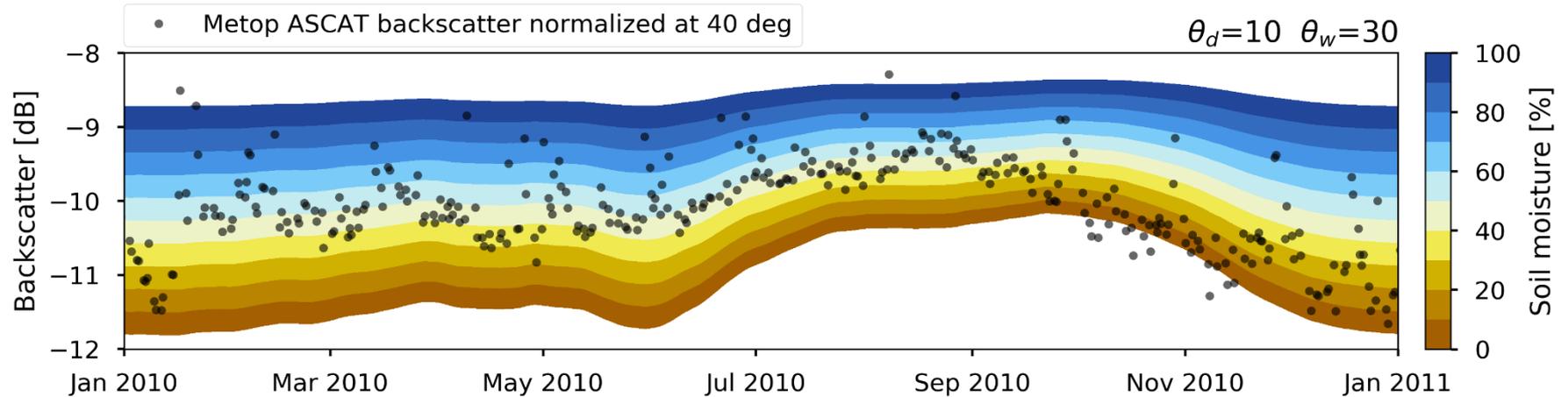
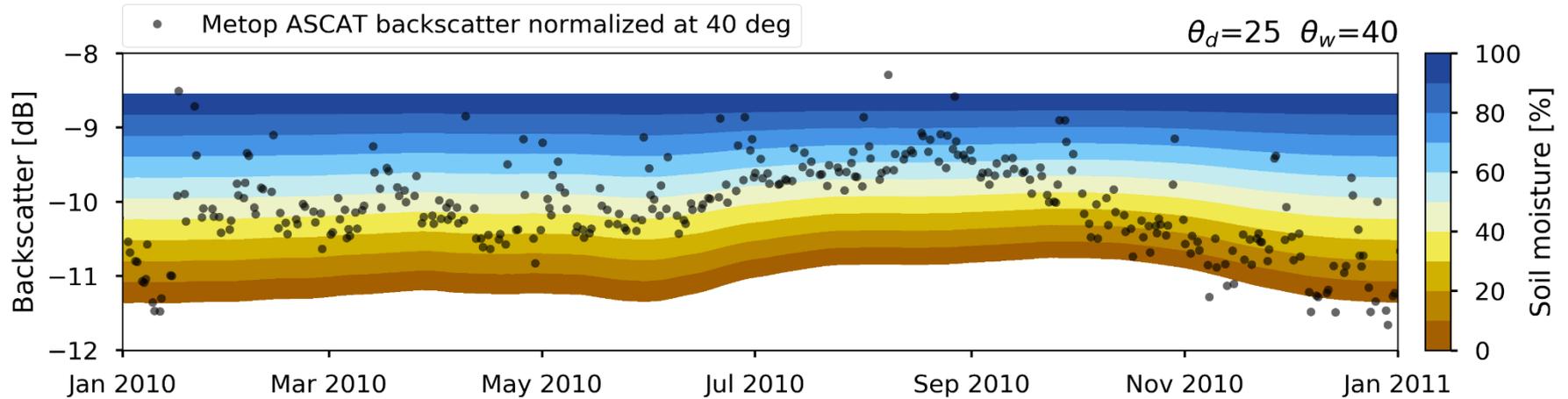


- Calibration of cross-over angles

# Slope and Curvature: Metop-A vs Metop-B

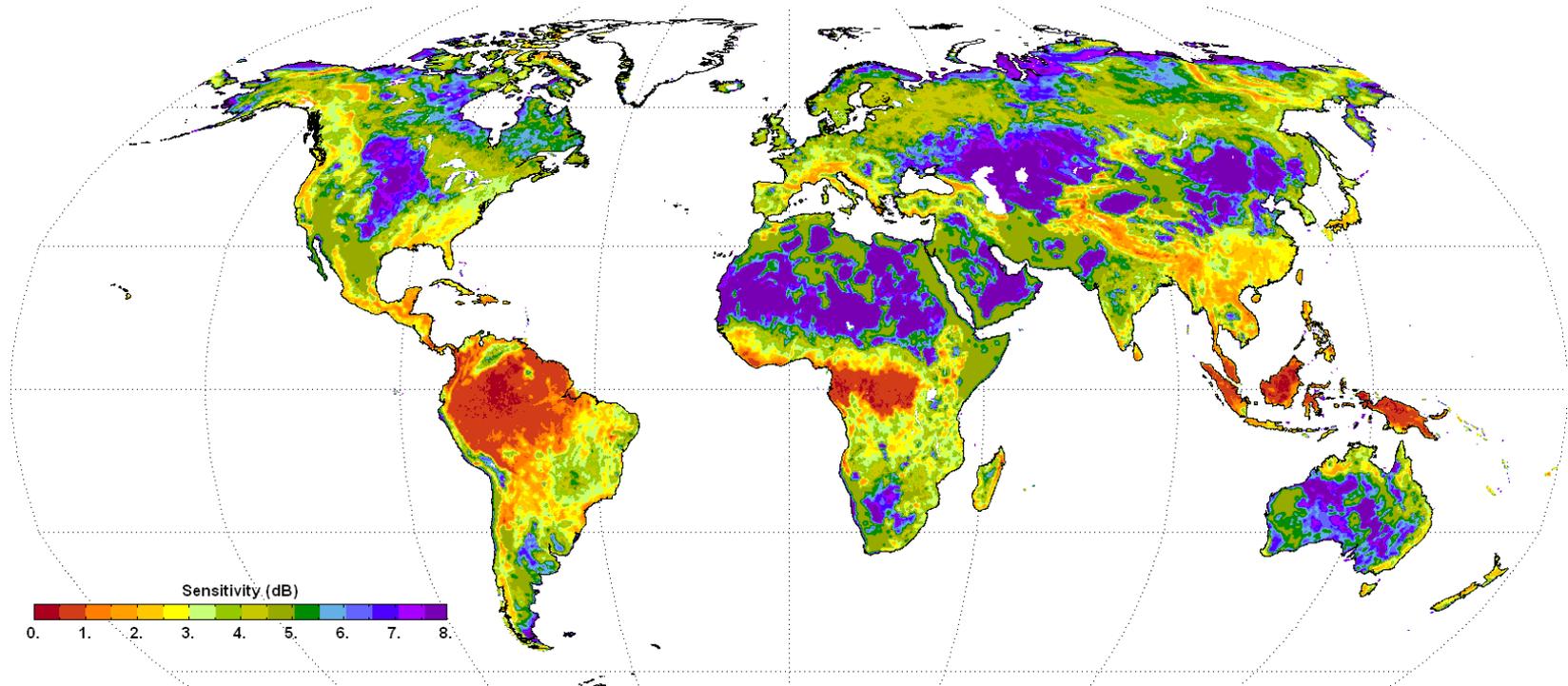


# Cross-over angle calibration & Dry/wet reference



# Model Parameters: Sensitivity

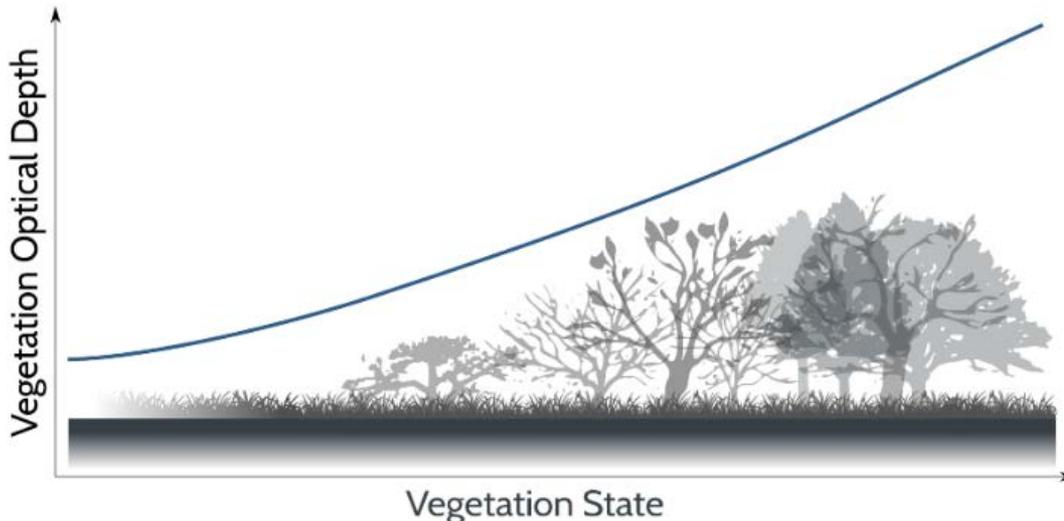
- The sensitivity is an output of calibration procedures to estimate backscatter at (completely) dry and wet (saturated) conditions respectively
  - describes the signal response to soil moisture changes
  - depends strongly on land cover



# Vegetation Optical Depth

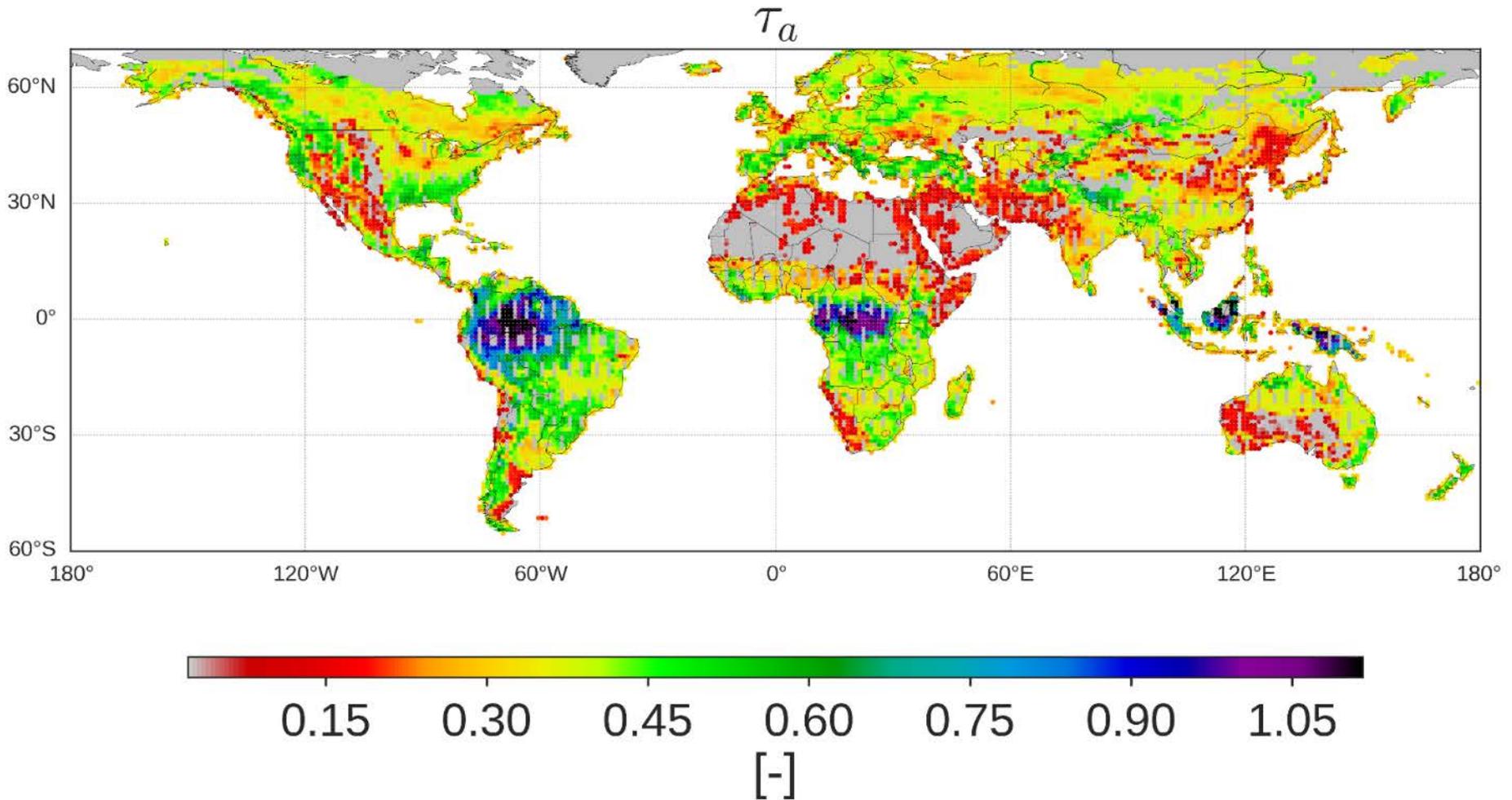
- Using the [Water Cloud](#) model we can now retrieve VOD from the TU Wien backscatter model formulation as well
- VOD is a measure of how much the soil moisture signal is taken away by the vegetation layer

$$\tau = \frac{\cos\theta}{2} \ln \frac{\Delta\sigma^0 \text{ for bare soil}}{\Delta\sigma^0 \text{ for vegetation covered soil}}$$

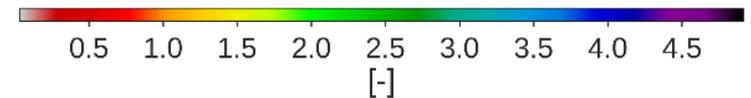
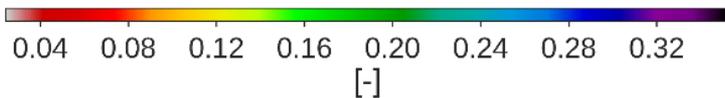
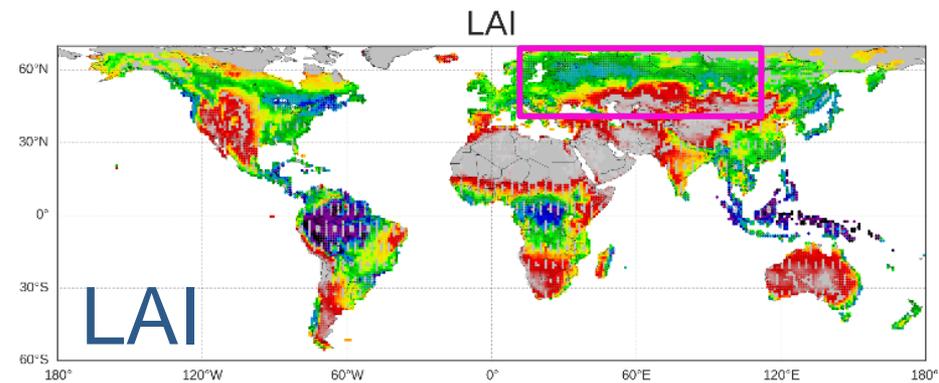
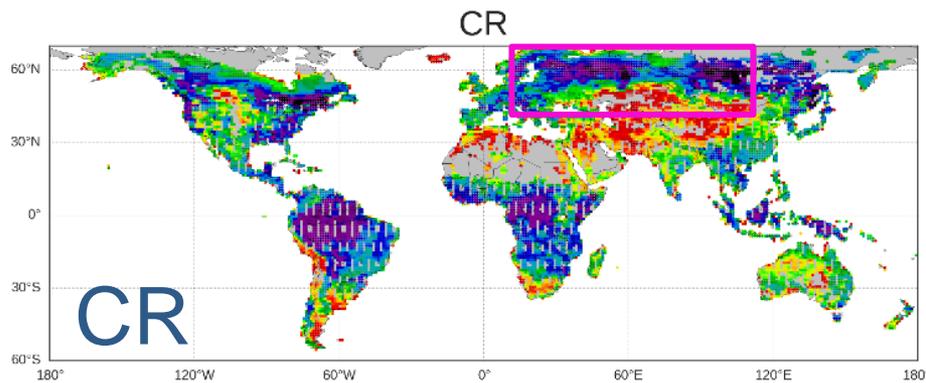
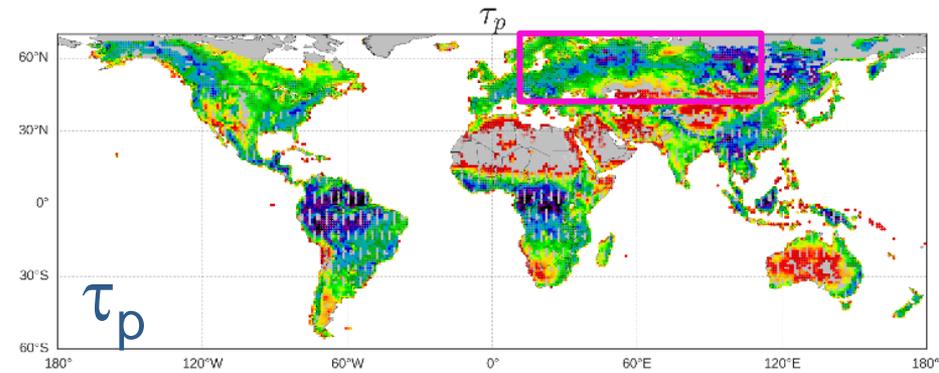
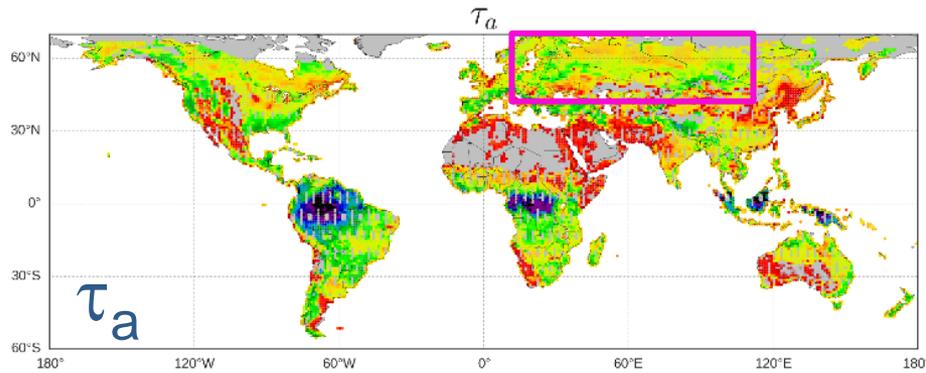


Vreugdenhil et al. (2016) Analysing the Vegetation Parameterisation in the TU-Wien ASCAT Soil Moisture Retrieval, *IEEE Transactions on Geoscience and Remote Sensing*, 54(6), 3513-3531.

# ASCAT Vegetation Optical Depth



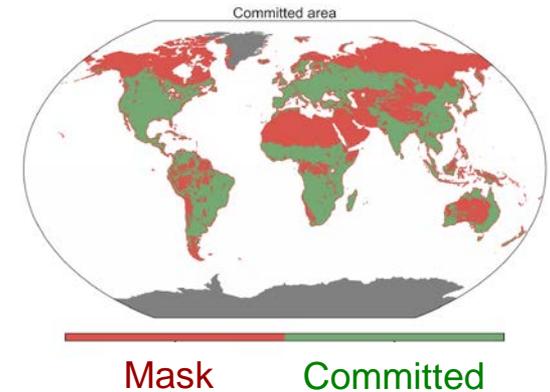
# Global Vegetation Patterns



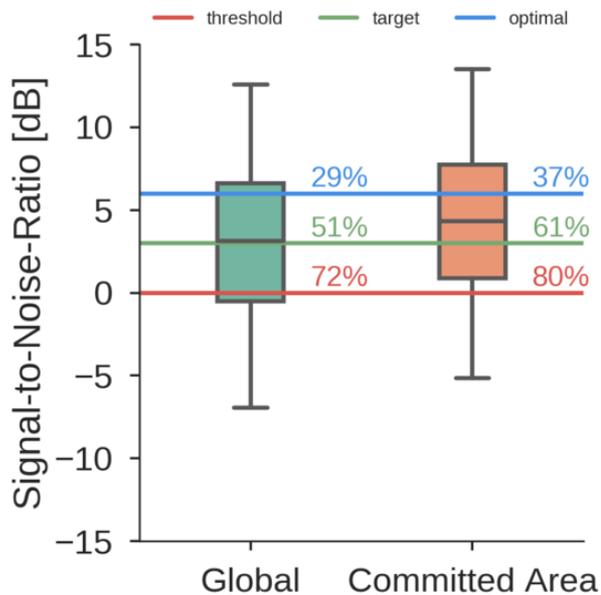
Mean values for  $\tau_a$ ,  $\tau_p$ , CR and LAI

# ASCAT Validation Metrics

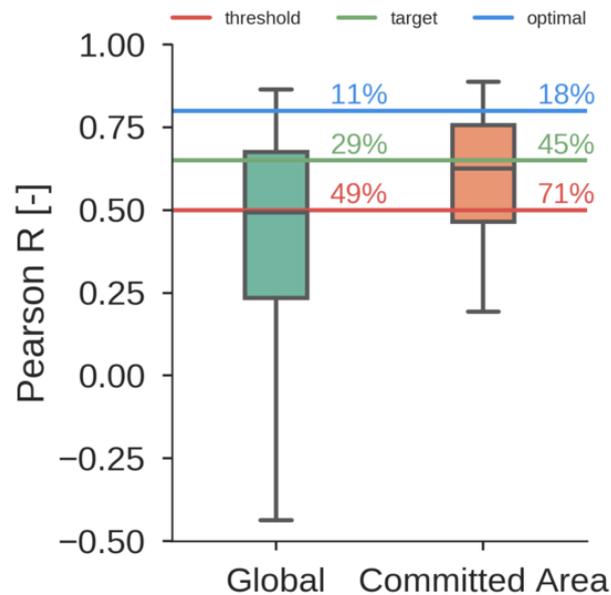
- Until 2012: RMSE as for SMOS and SMAP
- 2012-2016: Correlation to external model data set
- From 2016: SNR applied to committed areas only



Criteria since 2016

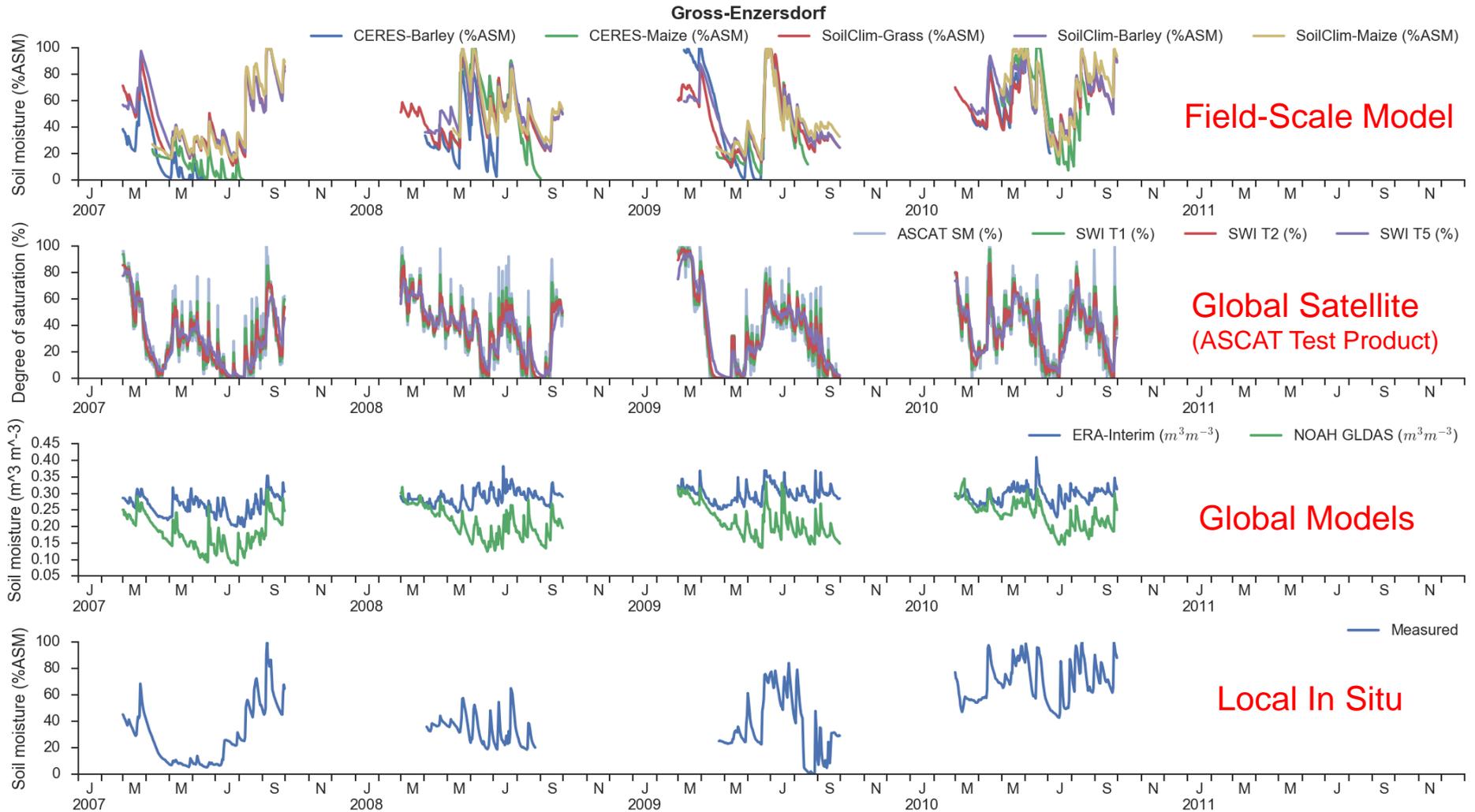


Criteria 2012-2016



<http://rs.geo.tuwien.ac.at/h-saf/H111/report.html>

# Soil Moisture from Models, In Situ and Satellites

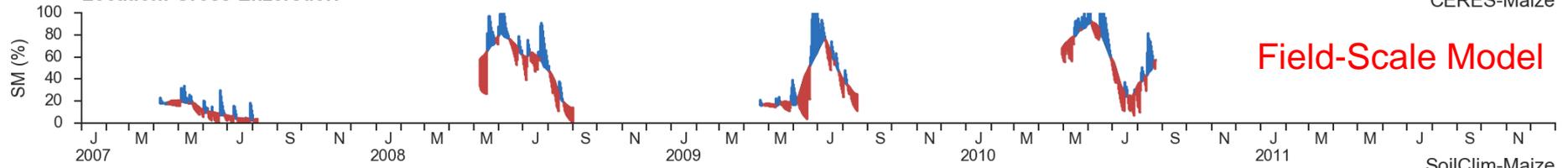


# Comparison of Short-Term Anomalies

Location: Gross-Enzersdorf

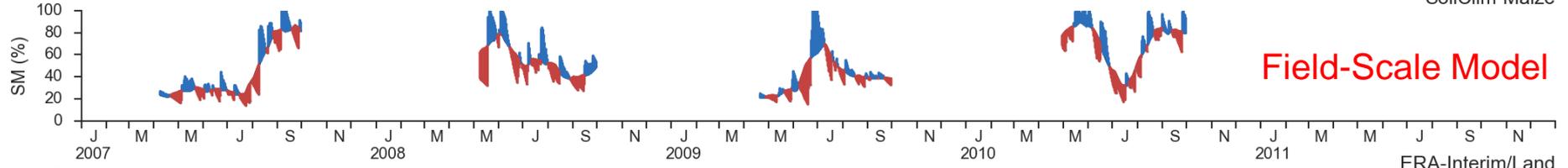
CERES-Maize

Field-Scale Model



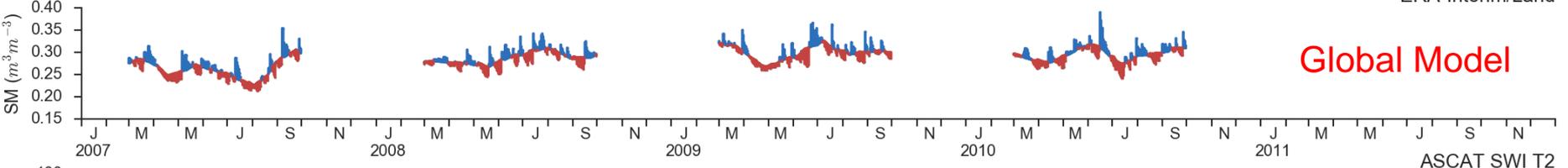
SoilClim-Maize

Field-Scale Model



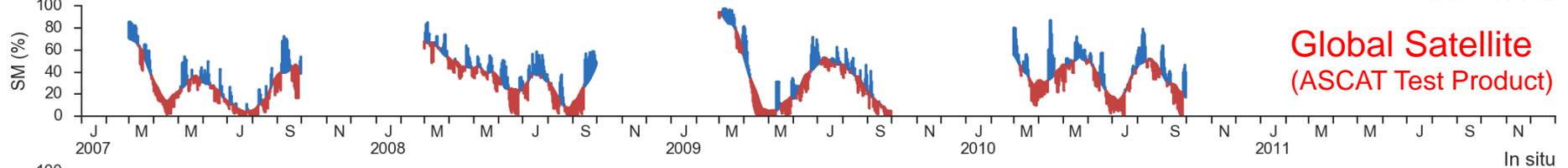
ERA-Interim/Land

Global Model



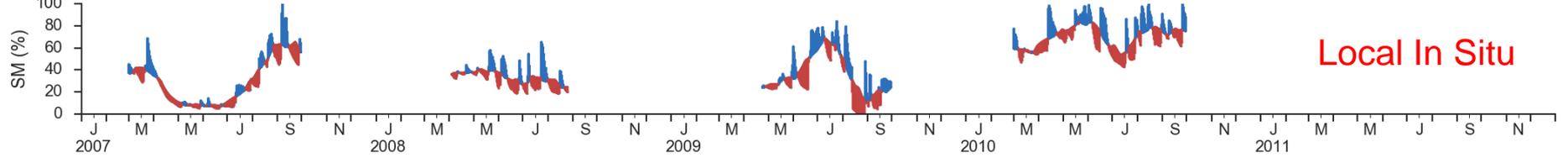
ASCAT SWI T2

Global Satellite  
(ASCAT Test Product)



In situ

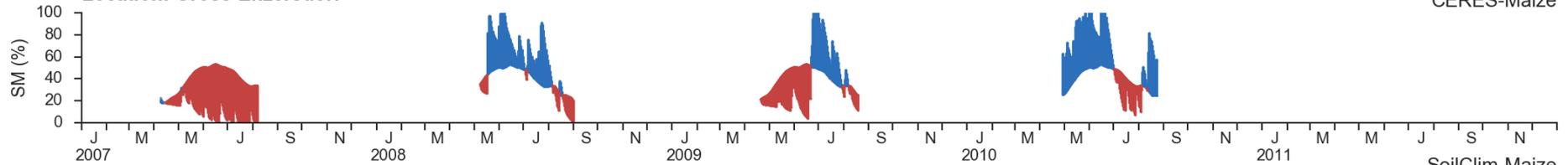
Local In Situ



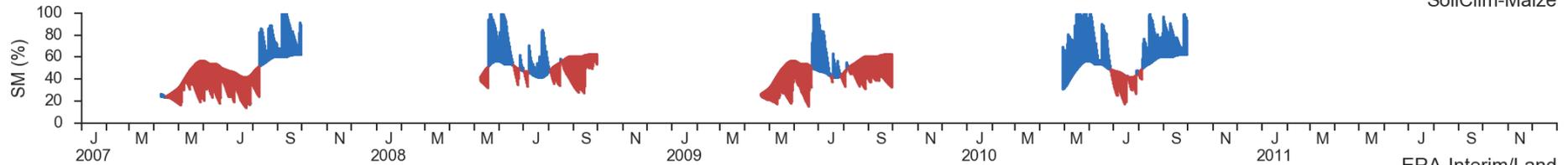
# Comparison Against Mean Seasonal Signals

Location: Gross-Enzersdorf

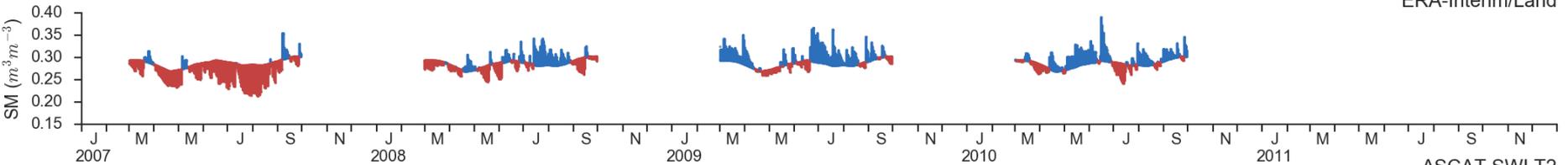
CERES-Maize



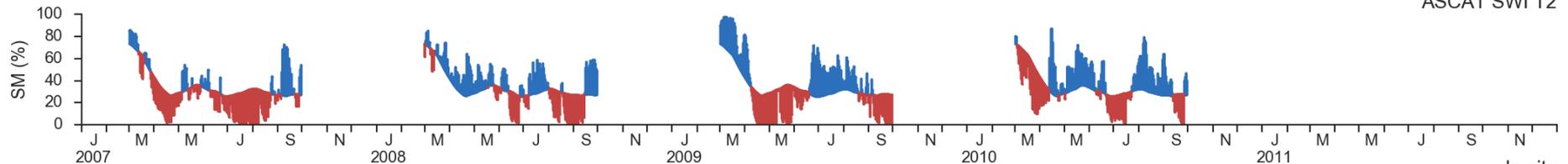
SoilClim-Maize



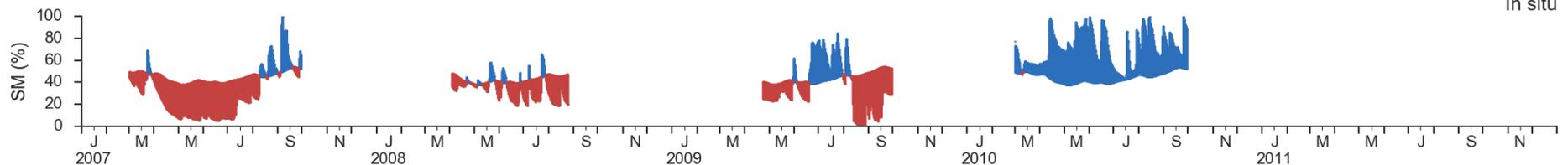
ERA-Interim/Land



ASCAT SWI T2

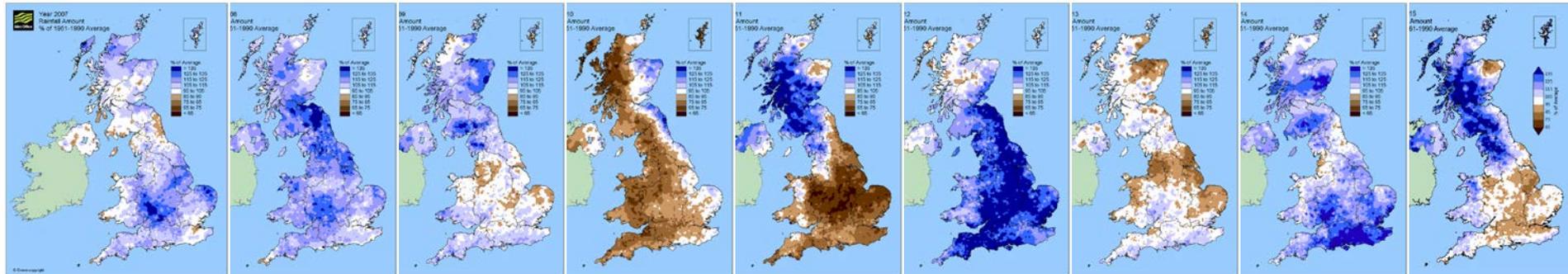


In situ

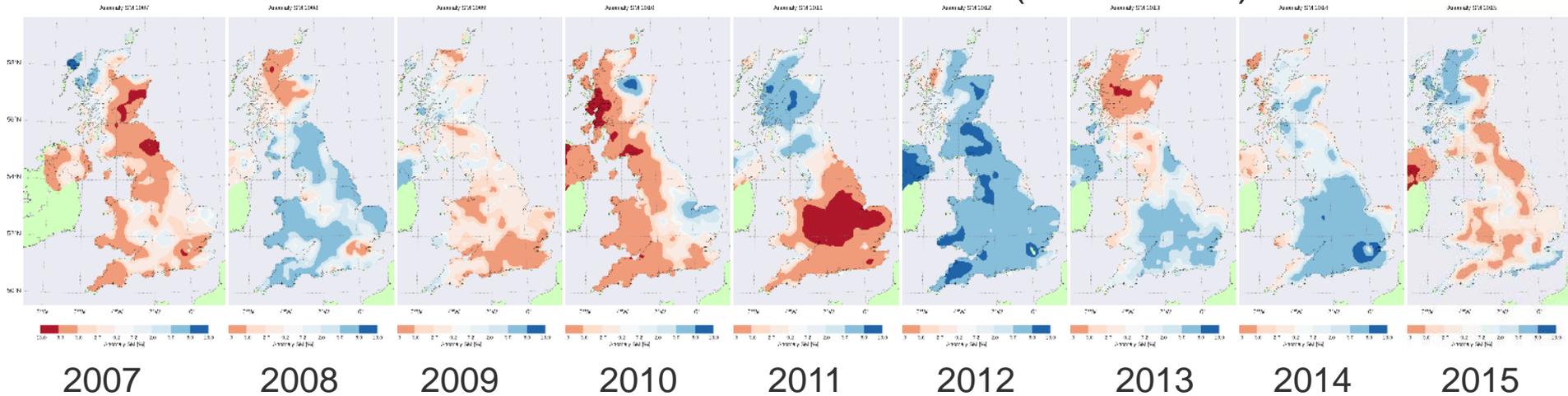


# ASCAT Soil Moisture

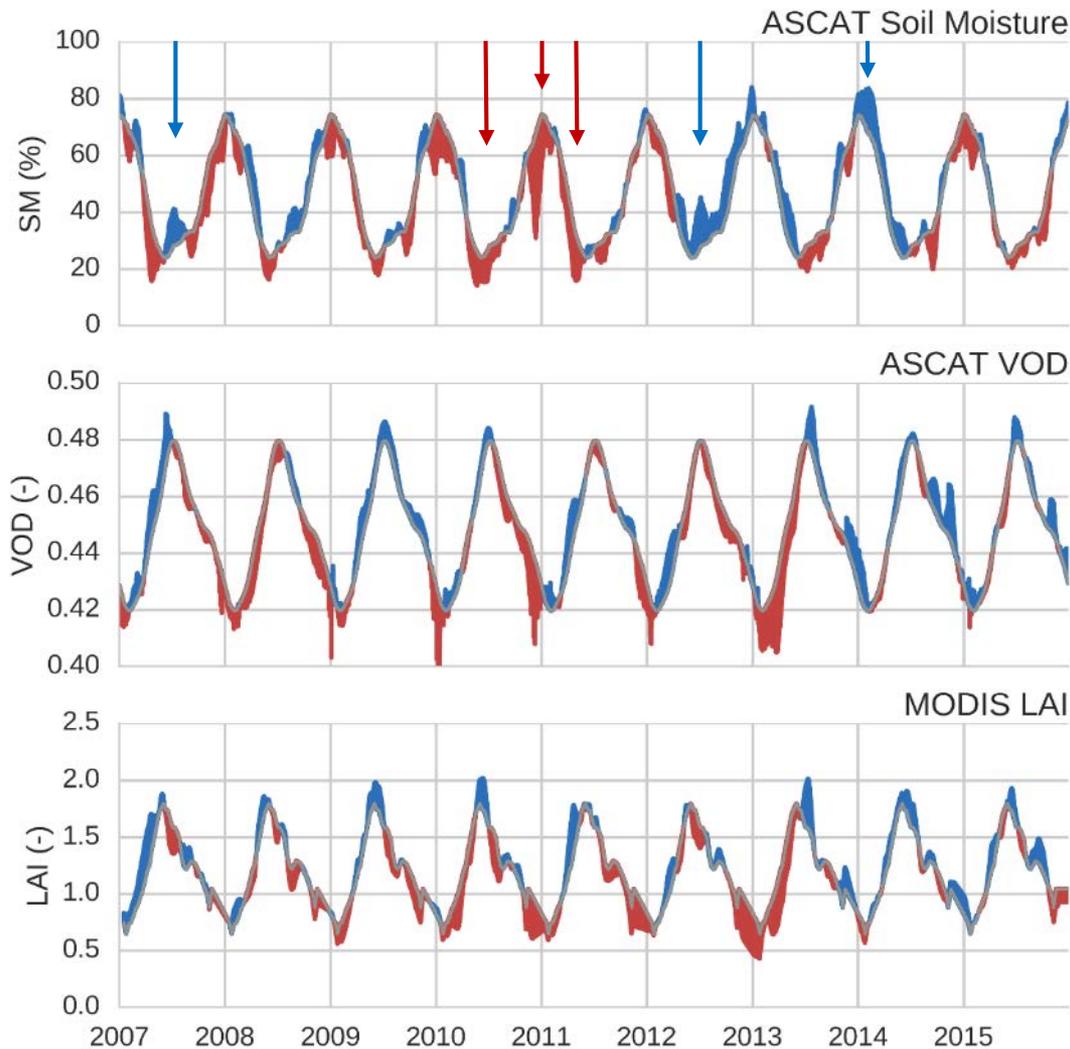
## MetOffice Precipitation anomalies (1961-1990)



## ASCAT Soil Moisture anomalies (2007-2015)



# Soil Moisture, VOD and LAI Anomalies

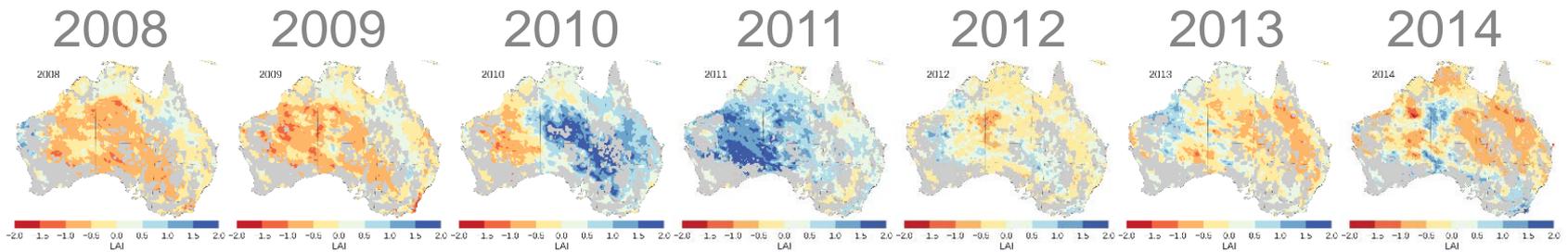


MetOffice weather summaries:

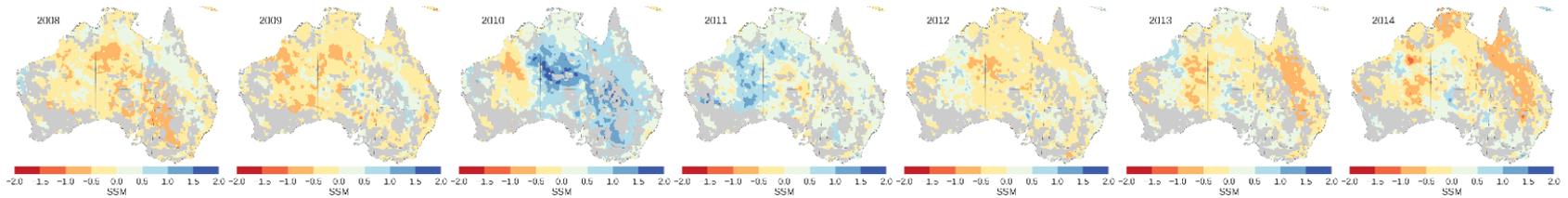
- **2007** very wet summer
- **2010** – Very dry April, May June
- **2010/2011** – Winter was dry and cold – lot of snow
- **2011** – Spring rainfall below normal over whole UK. Less than 1/3 of normal rainfall over southern and eastern England
- **2012** – Summer Floods
- **2013/2014** – Winter was wettest recorded since 1910

# Inter-annual variability – $\tau_a$

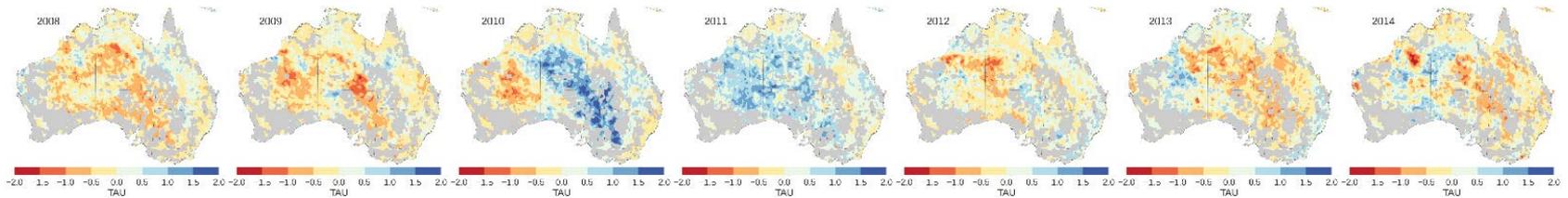
LAI



SM



$\tau_a$



Normalized yearly anomalies

# Value-Added Soil Moisture Products

# ASCAT Soil Moisture Services

## ■ Hydrology SAF

- Cooperation with EUMETSAT, ZAMG and ECMWF to deliver
  - 25 km ASCAT surface soil moisture data in near-real-time
  - Disaggregated 1 km ASCAT/ASAR soil moisture maps
  - Assimilated ASCAT soil moisture profile



## ■ Copernicus Global Land

- Cooperation with ZAMG and VITO to deliver
  - Daily 25 km Soil Water Index (SWI) product based on H-SAF soil moisture data
  - Evolution activity to produce 1km ASCAT/Sentinel-1 SWI data



## ■ CCI → C3S

- Cooperation with Vandersat, EODC and others to deliver
  - Long-term (1978 up to present) 0.25° merged active/passive microwave soil moisture product



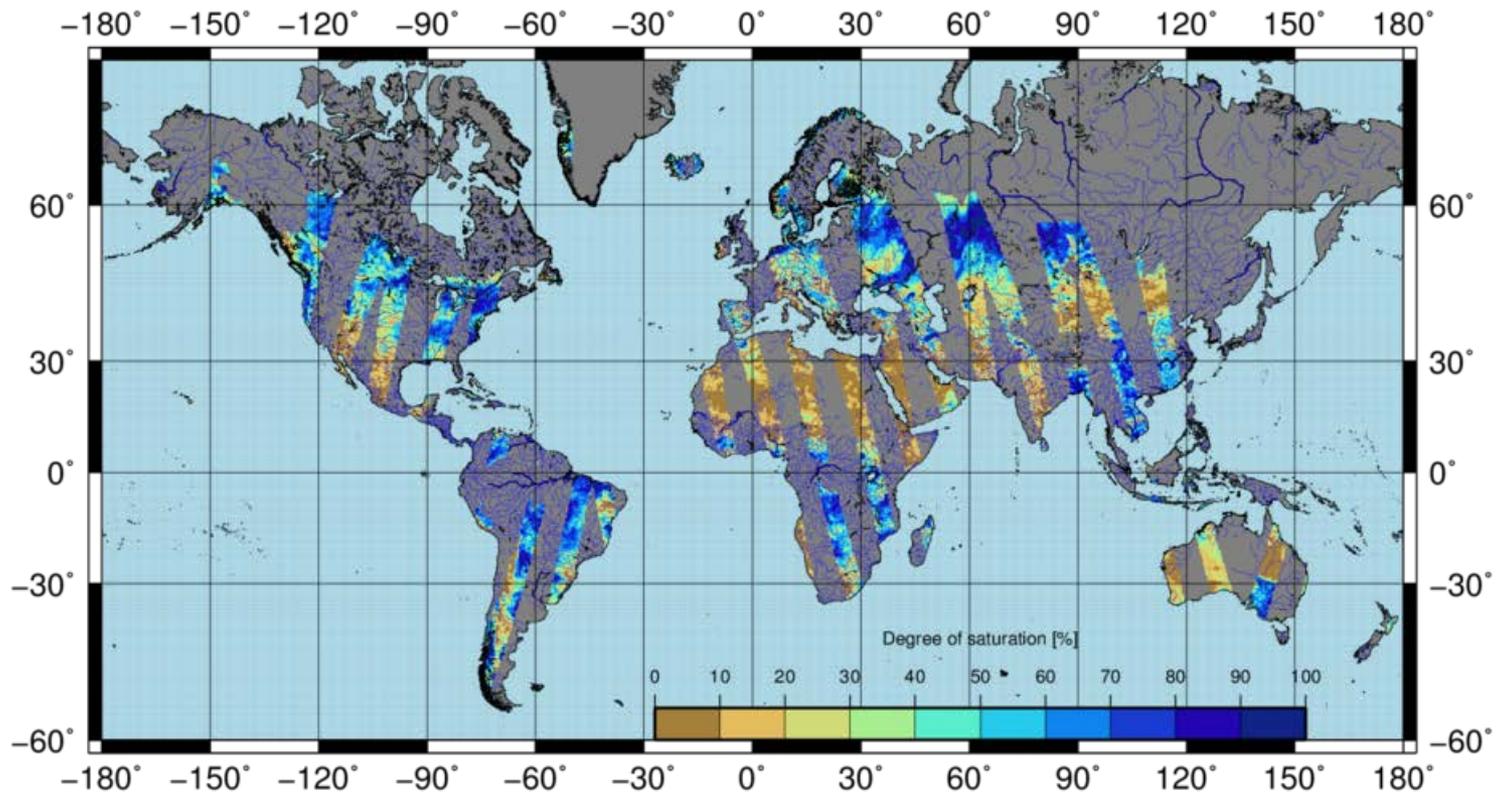
## ■ International Soil Moisture Network

- Global data hosting facility for in situ soil moisture data



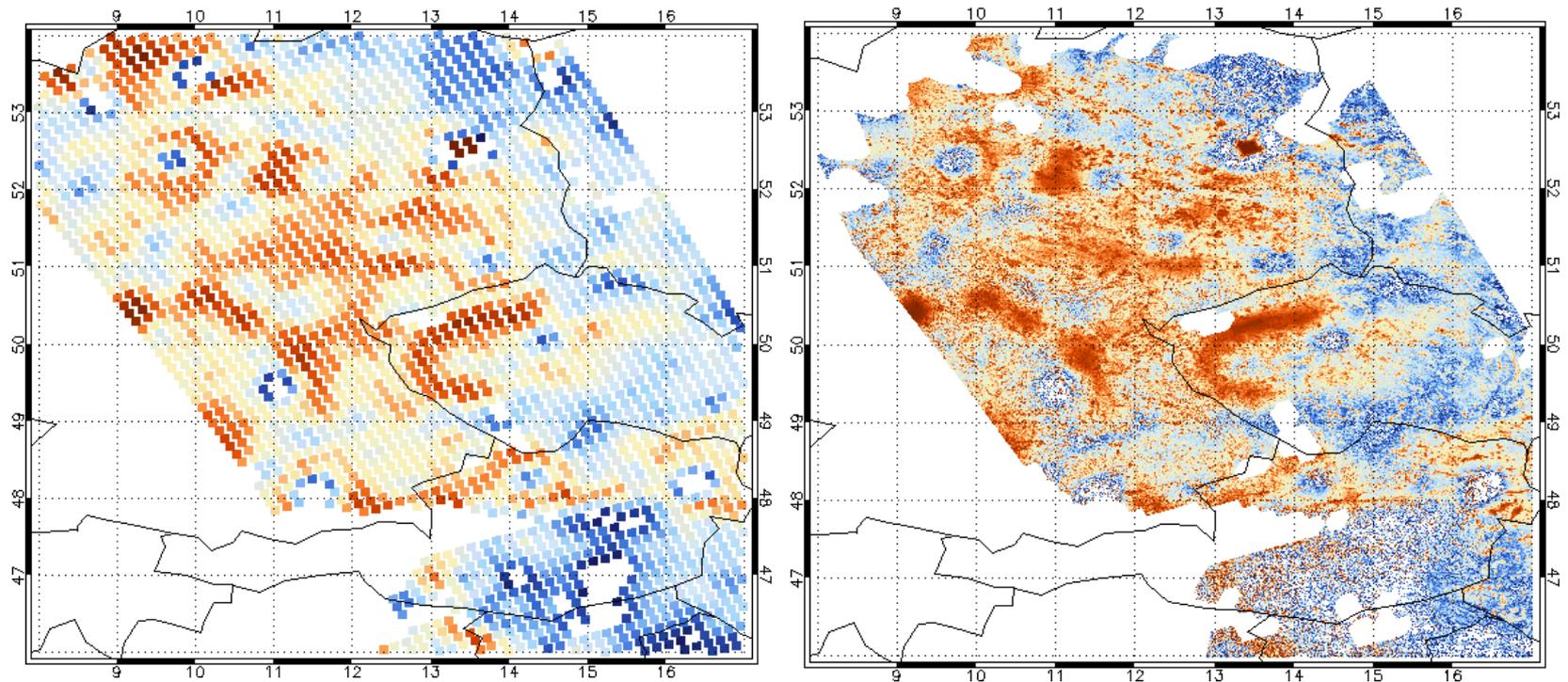
# ASCAT Surface Soil Moisture

ASCAT soil moisture 20170425\_0210, Metop-A, 125



# Disaggregated 1 km ASCAT Surface Soil Moisture

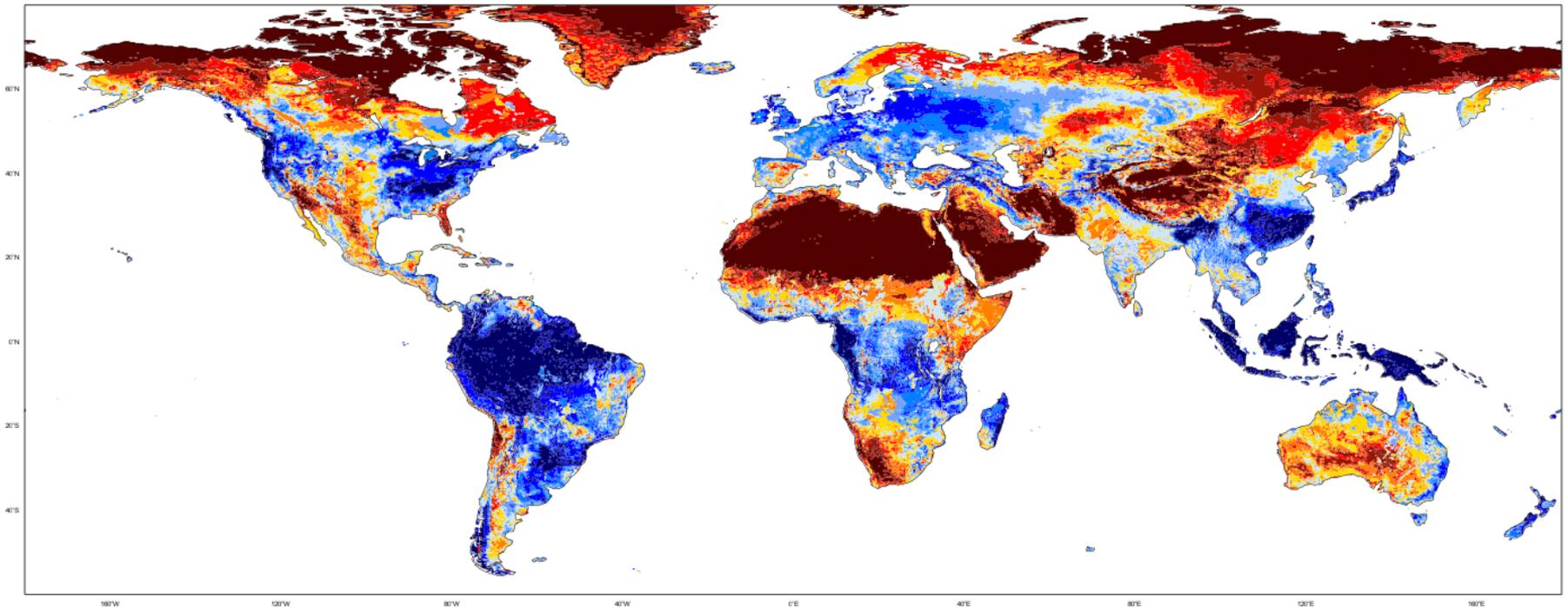
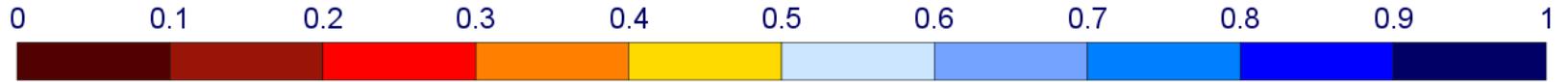
- Resampling of 25 km data using a static downscaling method based on scaling parameters derived from SAR time series



Left: 25 km ASCAT, right: 1 km downsampled surface soil moisture (DSSM).  
No-data values are masked and given a quality flag information.

# Assimilated ASCAT Soil Moisture

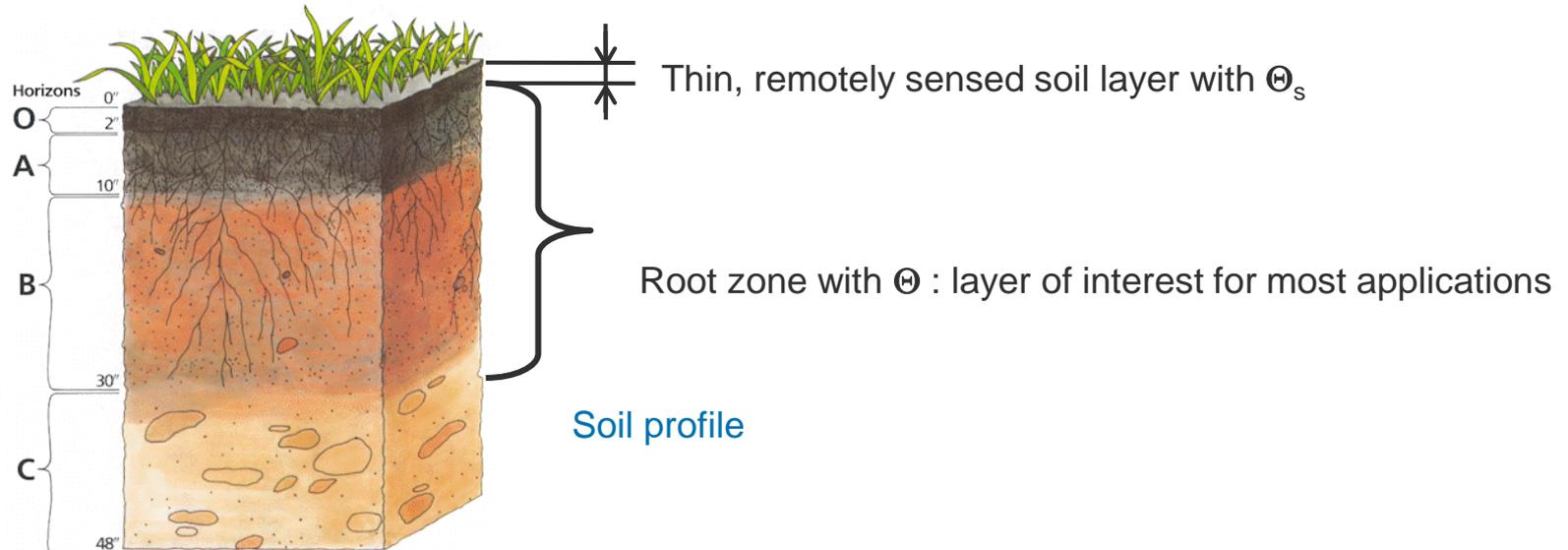
H14 Layer 3 (28-100cm) H-SAF CDOP - Copyright © Eumetsat20170425



# Soil Water Index

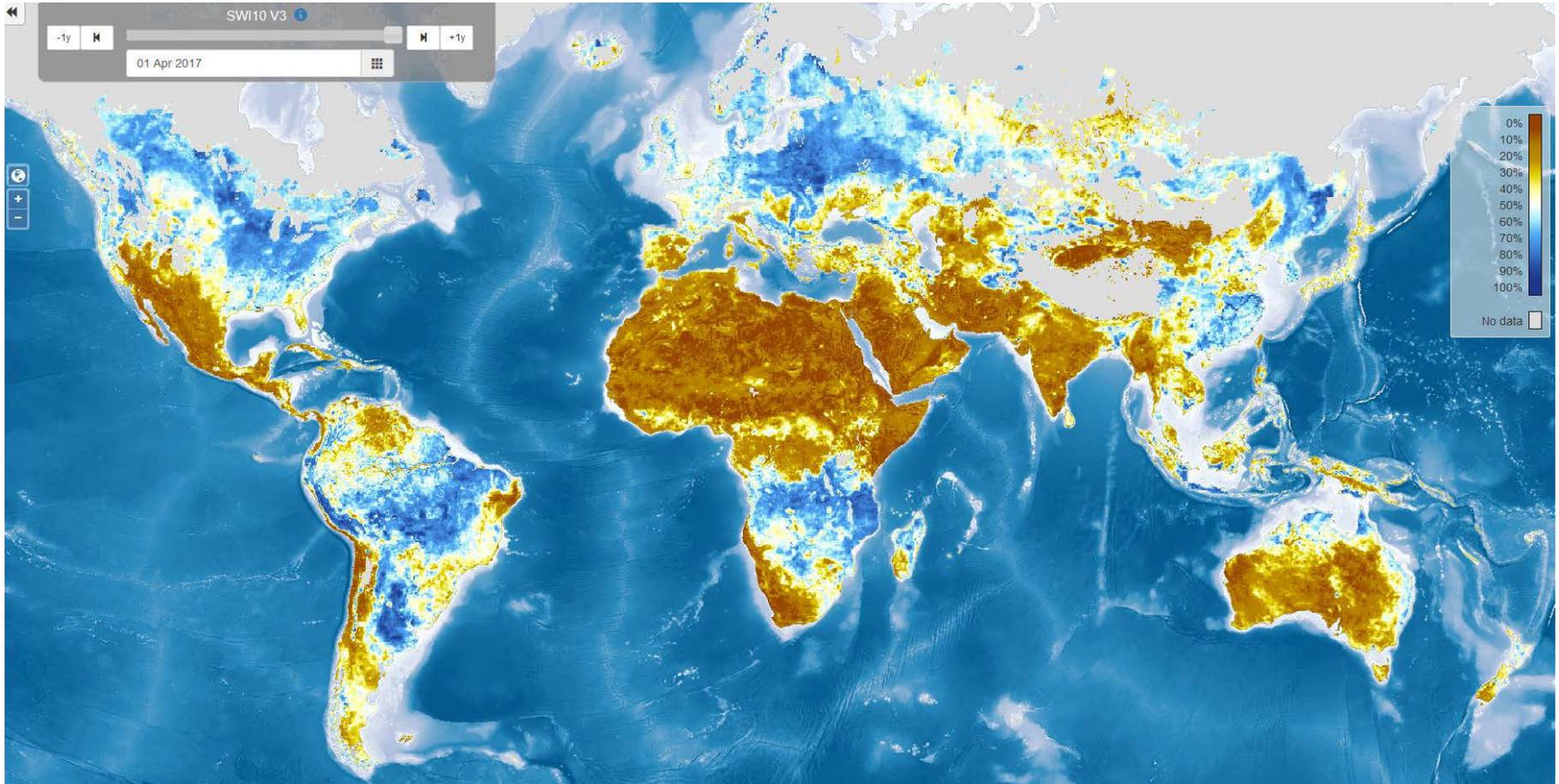
- The SWI is an indicator of the profile soil moisture content
- The method rests upon simple differential model for describing the exchange of soil moisture between surface layer ( $\Theta_s$ ) and the “reservoir” ( $\Theta$ )
  - T ... characteristic time

$$\frac{d\Theta}{dt} = \frac{1}{T} (\Theta_s - \Theta) \quad \Rightarrow \quad \Theta(t) = \frac{1}{T} \int_{-\infty}^t \Theta_s(t') \exp\left[-\frac{t-t'}{T}\right] dt'$$

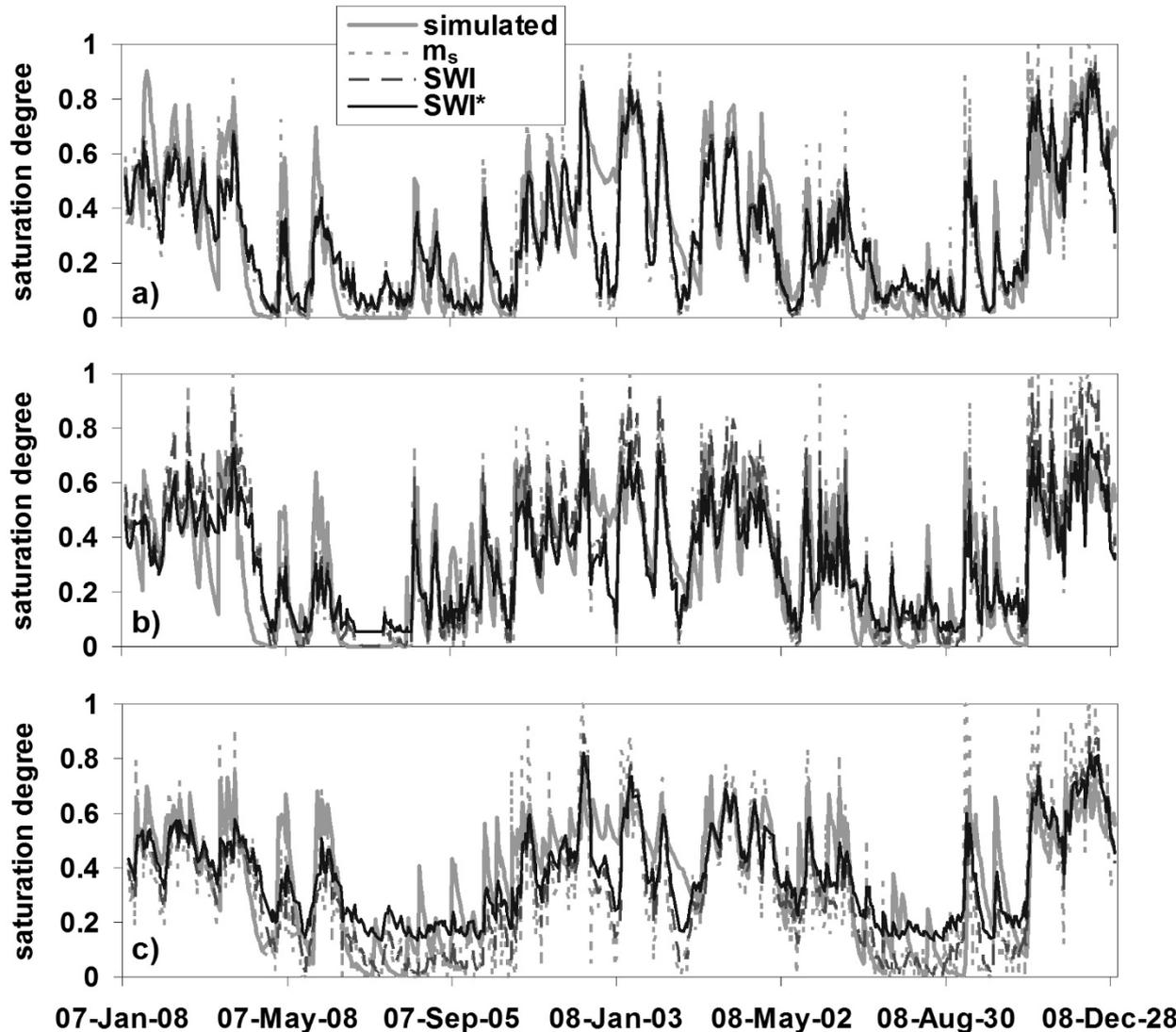


Wagner, W., G. Lemoine, H. Rott (1999) A Method for Estimating Soil Moisture from ERS Scatterometer and Soil Data, Remote Sensing of Environment, 70, 191-207.

# ASCAT Soil Water Index

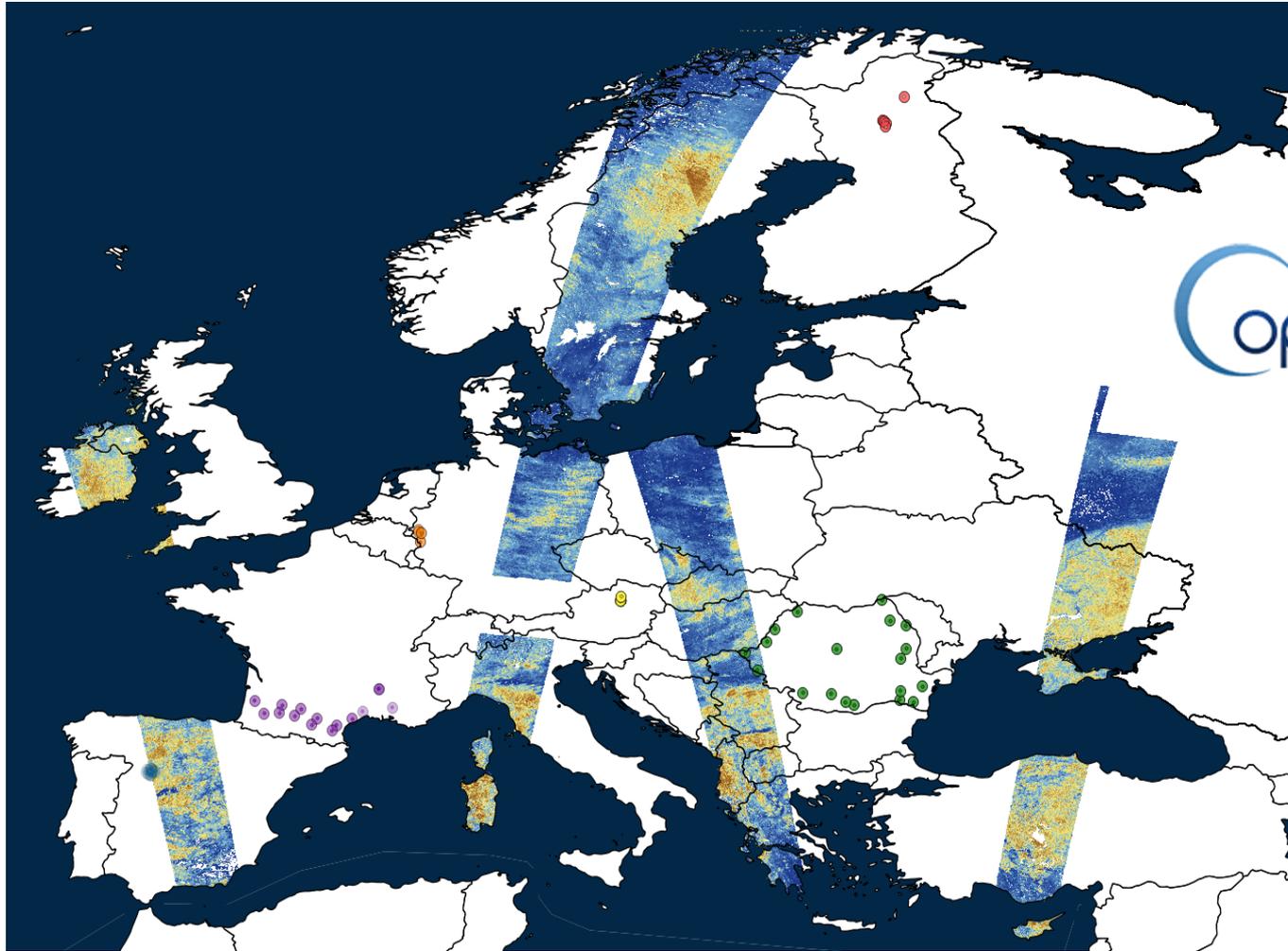


# ASCAT versus Model



ASCAT versus 3 cm simulated degree of saturation for products,  $m_s$ , SWI, and SWI\* and investigated sites: a) Vallaccia, b) Cerbara, and c) Spoleto.

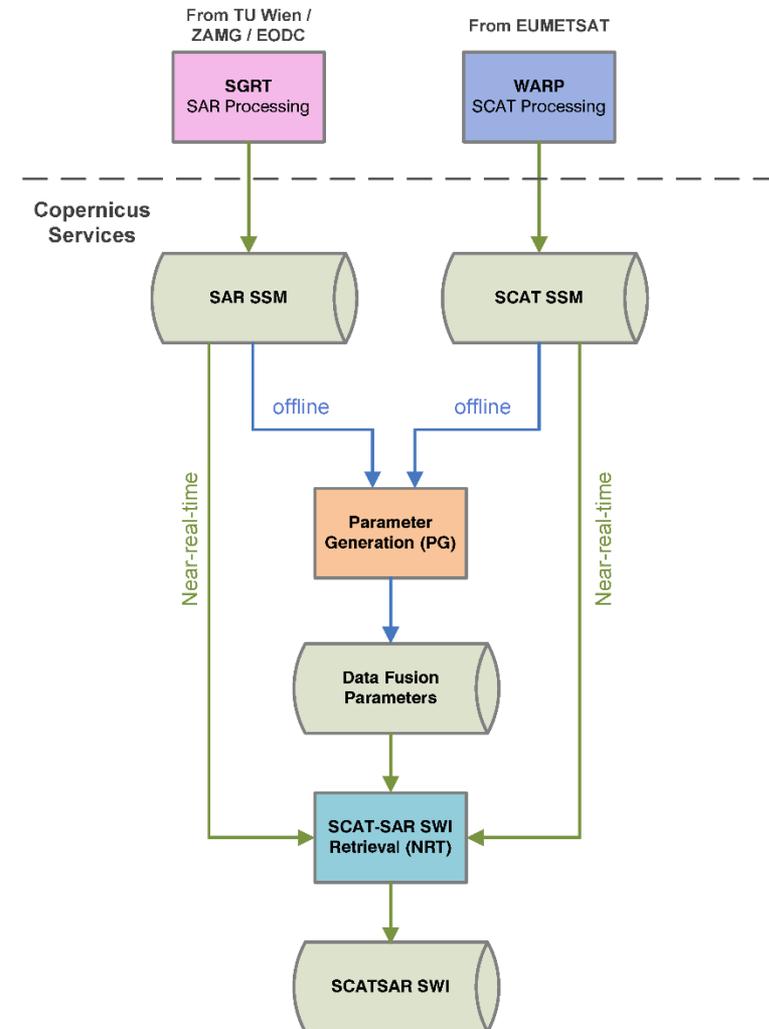
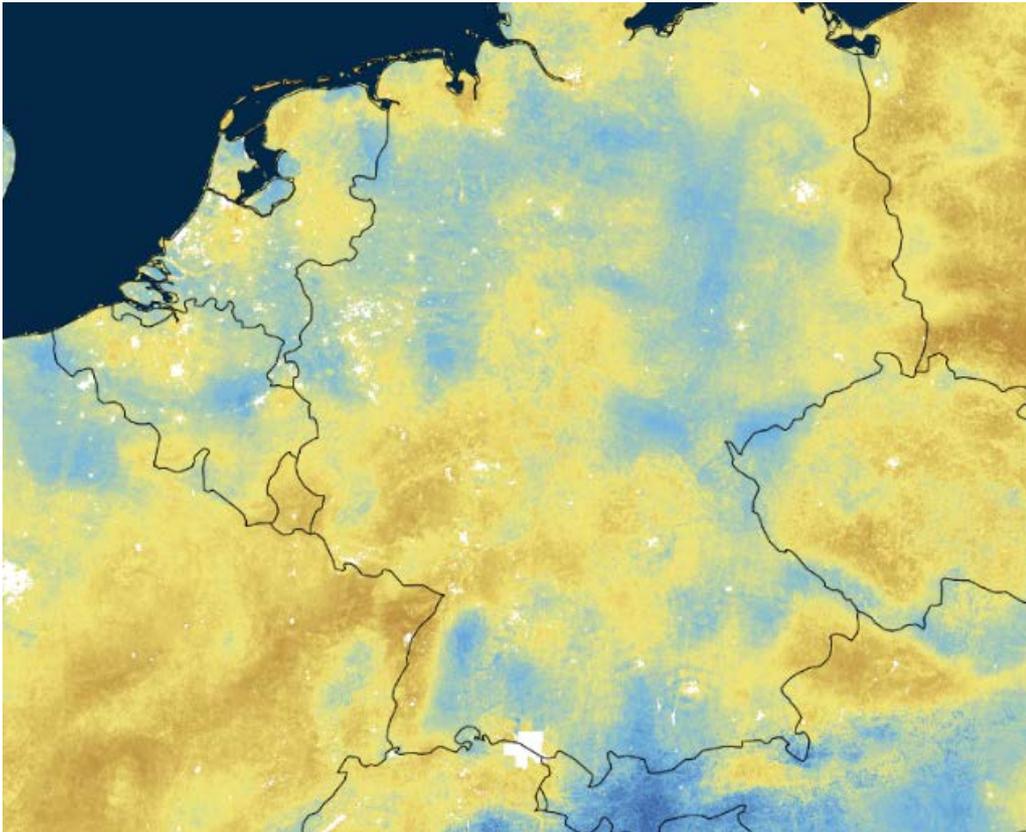
# 1 km Sentinel-1 SM Data



SSM1km on 2015-09-06 with ISMN stations used for validation

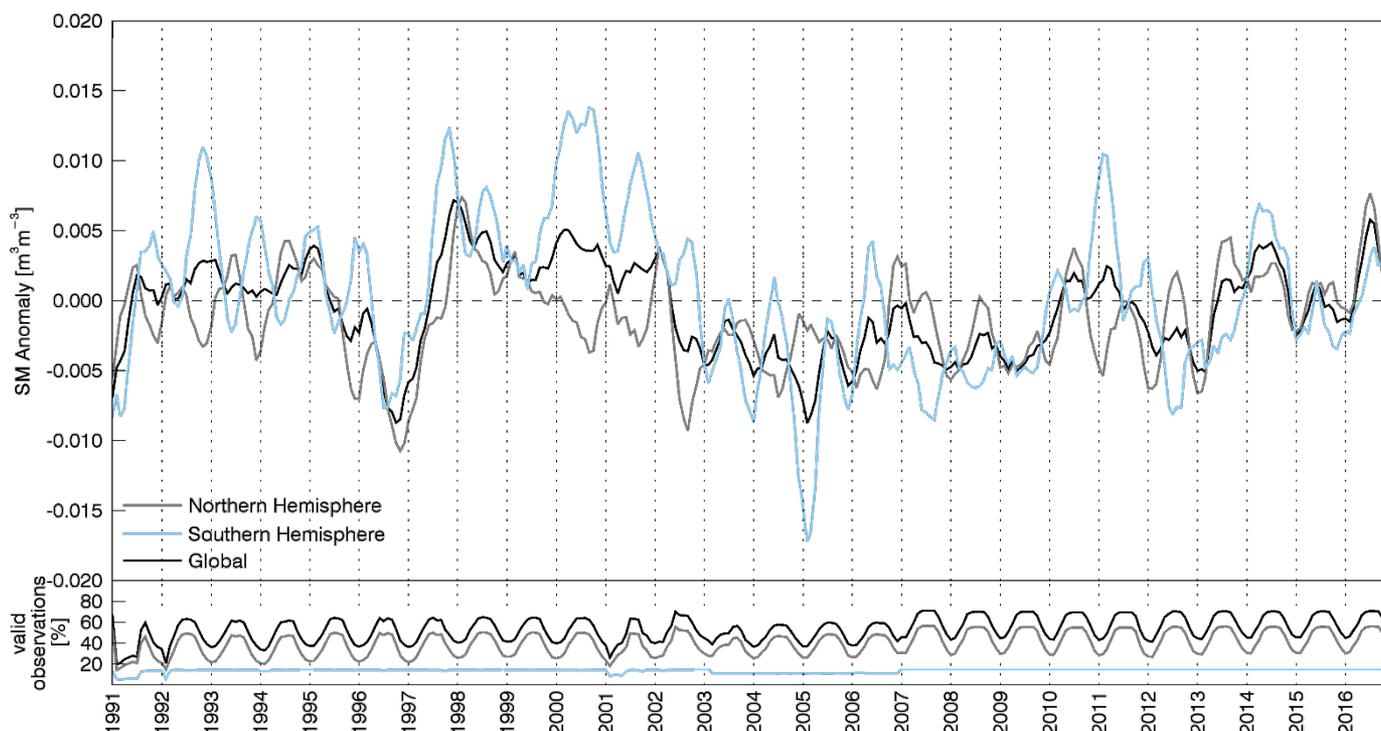
# ASCAT - Sentinel-1 Fusion Scheme

- Sentinel-1 surface soil moisture
  - Temporally sparse
- Fusion with ASCAT to produce daily profile Soil Water Index



# Soil Moisture Climate Data Record

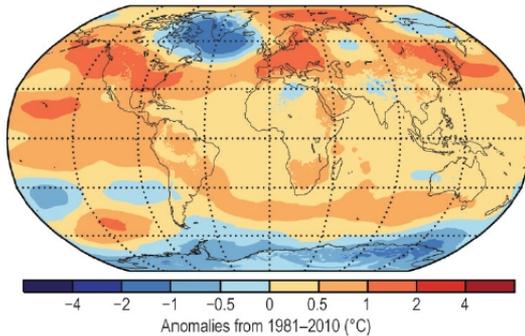
- ESA Climate Change Initiative (CCI)
  - > 3200 registered users (status 25.4.2017)
  - Latest release: v03.2 on 21.2.2017
    - **3 datasets**: Merged active, merged passive, and combined active-passive
- Transfer to Copernicus Climate Change Service (C3S)



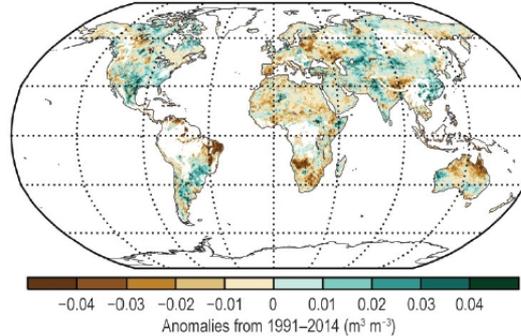
Prepared for BAMS  
State of the Climate  
2016

# BAMS State of the Climate in 2015

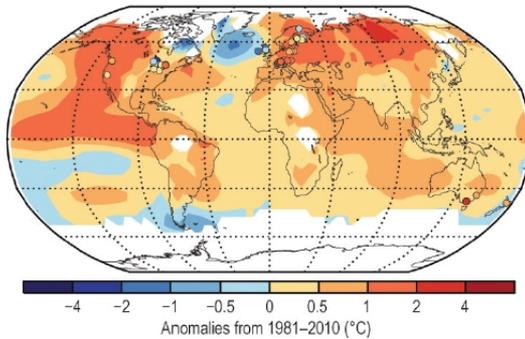
(b) Lower Tropospheric Temperature



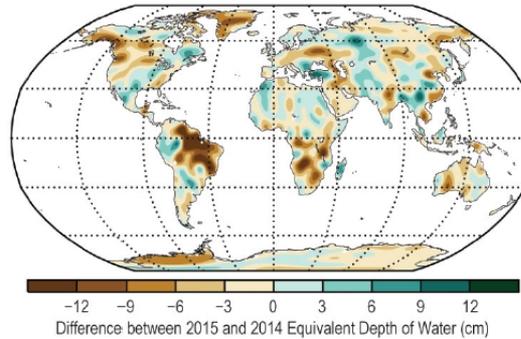
(f) Soil Moisture



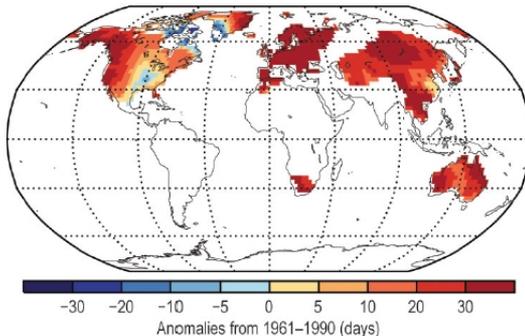
(c) Surface Temperature



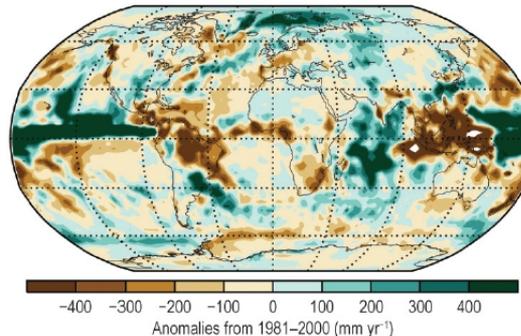
(g) Terrestrial Water Storage



(d) Warm Days



(h) Precipitation

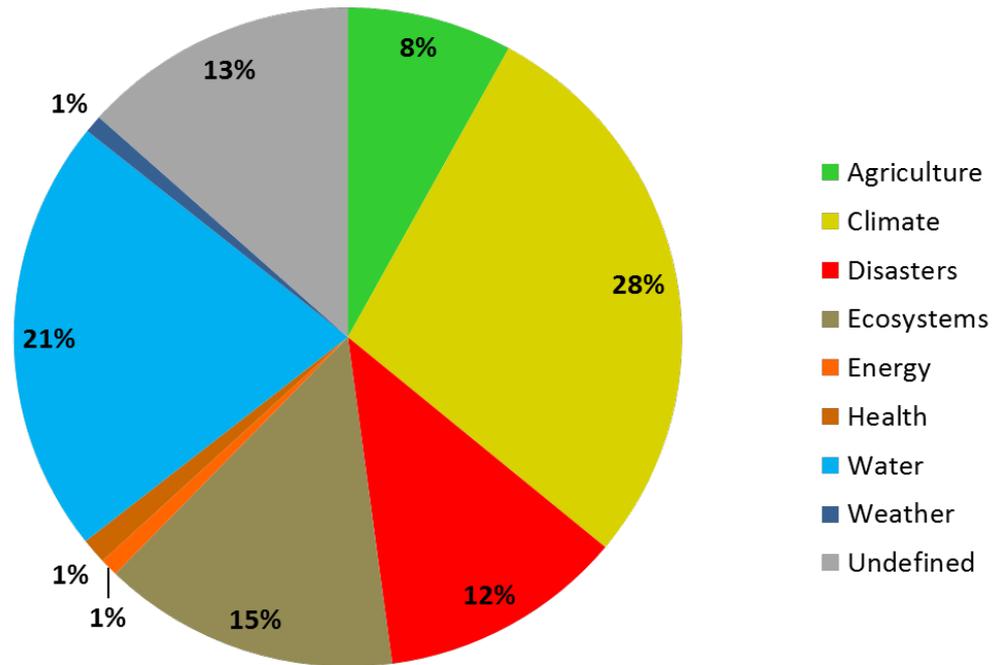


“Drier-than-average conditions were also evident over the global landmass. Soil moisture was below average for the entire year, and terrestrial groundwater storage was lower than at any other time during the record, which began in 2002. Areas in “severe” drought greatly increased, from 8% at the end of 2014 to 14% by the end of 2015.”

Yearly anomalies for selected variables in 2015. Extract of Plate 2.1 of BAMS State of the Climate 2015 report. Figure f shows soil moisture anomalies derived from ESA CCI soil moisture data set.

# CCI Soil Moisture Data Users

- Already over 3200 users (status April 2017)
- Scientific users dominate, but already 20 % of all users come from public and commercial sector



## Application Domains

Agriculture has grown by 2% in the past years

# Conclusions

- Large number and high diversity of ASCAT soil moisture data users
- There is no one data product/service that can serve all user requirements
- Users must familiarise them with service specifications & data product characteristics
- ASCAT particularly interesting for operational users thanks to METOP-SG and EUMETSAT's long term vision on the SAFs

## Acknowledgements

EUMETSAT: H-SAF CDOP2 and CDOP3

EU: Grant agreements no 603608 "Earth2Observe" and 606971 "AdvancedSAR"

ESA: ESRIN Contract No. 4000112226/14/I-NG "Phase 2 of CCI Soil Moisture"

Austrian Space Application Programme: 854030 "EOP-Danube"

Austrian Science Fund: Vienna Doctoral Programme on Water Resources Systems

