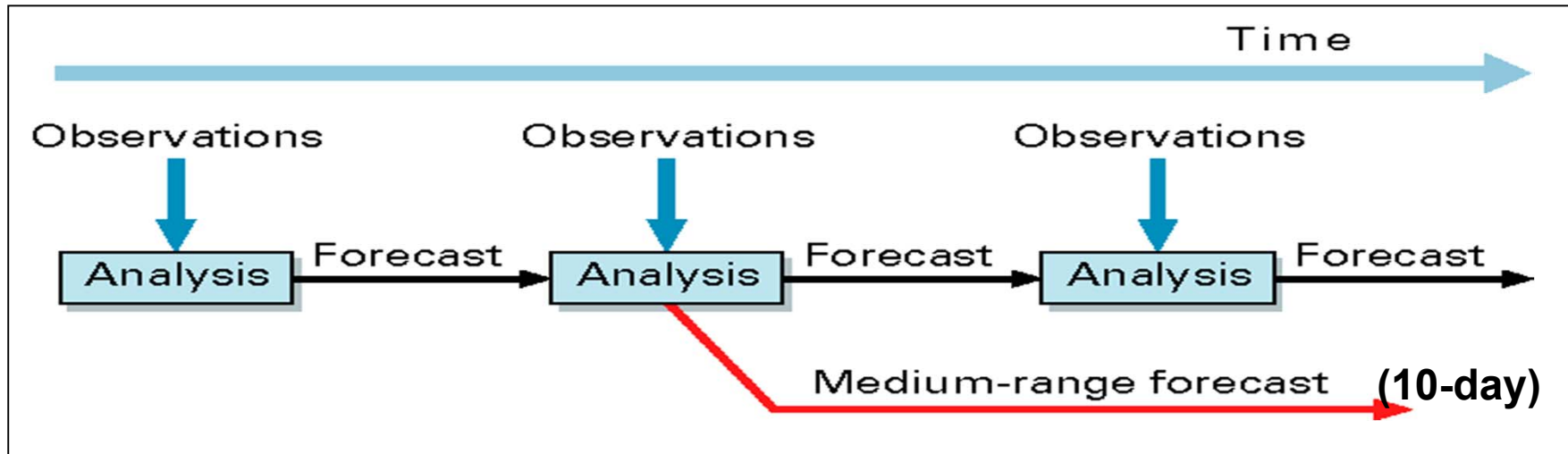


**1st International Surface Working Group
19-20 July 2017, Monterey, CA**

Assimilation of land surface satellite data for Numerical Weather Prediction at ECMWF

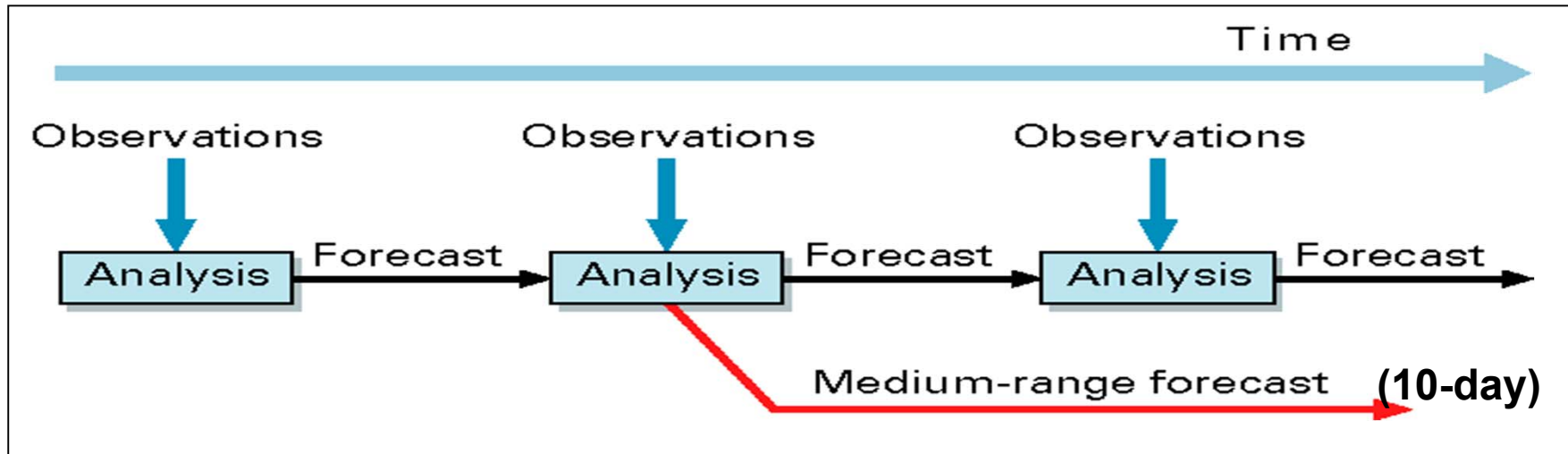
**P. de Rosnay, G. Balsamo, C. Albergel, S. Boussetta, S. English, D.
Fairbairn, H. Hersbach, L. Isaksen,
J. Muñoz Sabater, N. Rodríguez-Fernández,**

ECMWF Integrated Forecasting System (IFS)



- **Forecast Model:** GCM including the H-TESSSEL land surface model (coupled)
- **Data Assimilation** → initial conditions of the forecast model prognostic variables
 - 4D-Var for atmosphere ; 3D-Var for ocean (for ensemble and seasonal)
 - Land Data Assimilation System (LDAS)

ECMWF Integrated Forecasting System (IFS)



- **Forecast Model:** GCM including the H-TESSSEL land surface model (coupled)
- **Data Assimilation** → initial conditions of the forecast model prognostic variables
 - 4D-Var for atmosphere ; 3D-Var for ocean (for ensemble and seasonal)
 - Land Data Assimilation System (LDAS)

Land assimilation in ECMWF systems:

- **NWP:** IFS (with 4D-Var, **LDAS**), 9km, 43r1
- **ERA-Interim:** IFS (with 4D-Var, **LDAS**), 79km, 31r1 (2006)
- **ERA5:** IFS (with 4D-Var, **LDAS**), 31km, 41r2
- **ERA-Interim-Land:** H-TESSSEL forced by ERA → LSM model only: no DA
- **CERA-20C** IFS (with 4D-Var, NEMOVAR), 130km, 41r2 (no LDAS)
- **CERA-SAT** IFS (with 4D-Var, NEMOVAR and **LDAS**), 62km, 42r1
- **ENS** IFS (with 4D-Var, NEMOVAR and **LDAS**), 16/32km, 43r1

ECMWF Land Data Assimilation System (LDAS)

Snow depth

Methods: Cressman for ERA-Interim, 2D Optimal Interpolation (OI) for NWP & for ERA5

Conventional observations: *in situ* snow depth

Satellite data: [NOAA/NESDIS IMS Snow Cover Extent \(daily product\)](#).

Soil moisture (SM)

Methods: - 1D Optimal Interpolation in ERA-Interim (also used at Météo-France, ECCO)
- Simplified Extended Kalman Filter for NWP, ERA5 (also at UKMO)

Conventional observations: Analysed SYNOP 2m air relative humidity and air temp.

Satellite data: [Scatterometer SM for NWP \(ASCAT\) & for ERA5 \(ERS/SCAT & ASCAT\)](#)

[ESA SMOS brightness temperature in development, research NASA SMAP](#)

Soil Temperature and Snow Temperature

1D-OI using analysed T2m as observation (NWP, ERA-Interim, ERA5)

Snow in the ECMWF IFS for NWP

Snow Model: Component of H-TESEL (Dutra et al., JHM 2010, Balsamo et al JHM 2009)

Single layer snowpack

- Snow water equivalent SWE (m)
- Snow Density ρ_s

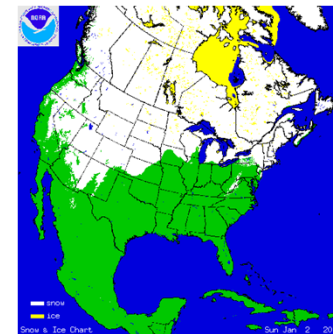
} Prognostic variables

Observations: de Rosnay et al ECMWF Newsletter 2015

- Conventional snow depth data: SYNOP and National networks
- Snow cover extent: NOAA NESDIS/IMS daily product (4km)

Data Assimilation: de Rosnay et al SG 2014

- Optimal Interpolation (OI) in operational IFS
- The result of the data assimilation is the analysis of SWE and snow density



Snow cover observations

Interactive Multisensor Snow and Ice Mapping System (IMS)

- Time sequenced imagery from geostationary satellites
- AVHRR,
- VIIRS,
- SSM/I, etc....
- Station data

Northern Hemisphere product

- Daily
- Polar stereographic projection

Information content: Snow/Snow free

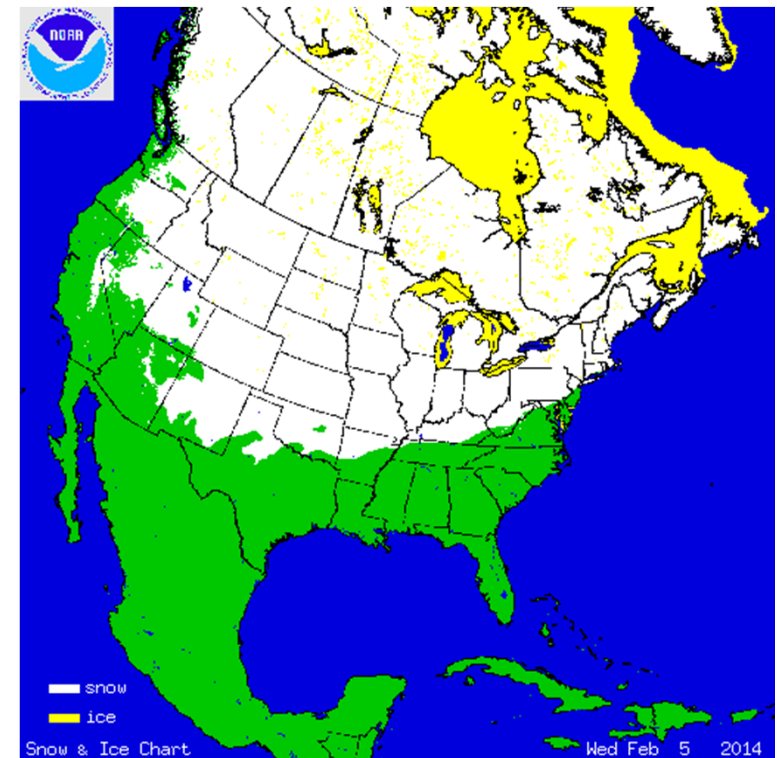
Data used at ECMWF:

- **24km product** (ERA-Interim)
- **4 km product** (NWP, ERA5)

Latency:

Available daily at 23 UTC. Assimilated in the subsequent analysis at 00UTC

NOAA/NESDIS IMS Snow extent data



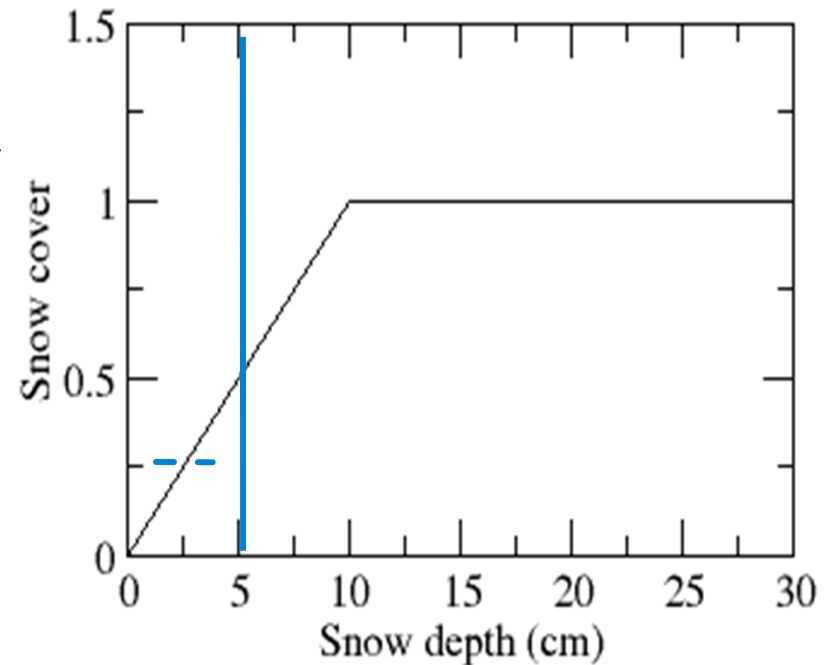
<http://nsidc.org/data/g02156.html>

Revised NESDIS/IMS snow cover DA

- IMS snow cover (SC) means $SC > 50\%$
- But no quantitative information on snow depth
- Relation snow cover (SC)/Snow Depth (SD):
 $SC = 50\%$ corresponds to $SD = 5\text{cm}$
- Previously: direct insertion of 10cm when IMS has snow & model has no snow
- IFS revision: assimilate IMS & account for observation error

IMS has snow ($SC \geq 50\%$)
 → Model snow depth $\geq 5\text{cm}$

Guess NESDIS/IMS	Fst	Snow	No Snow
	Snow	x	DA
No Snow	DA	DA	DA



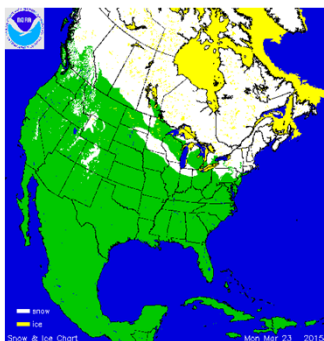
Model relation between Snow Cover (SC) and Snow Depth (SD)

Error specifications:

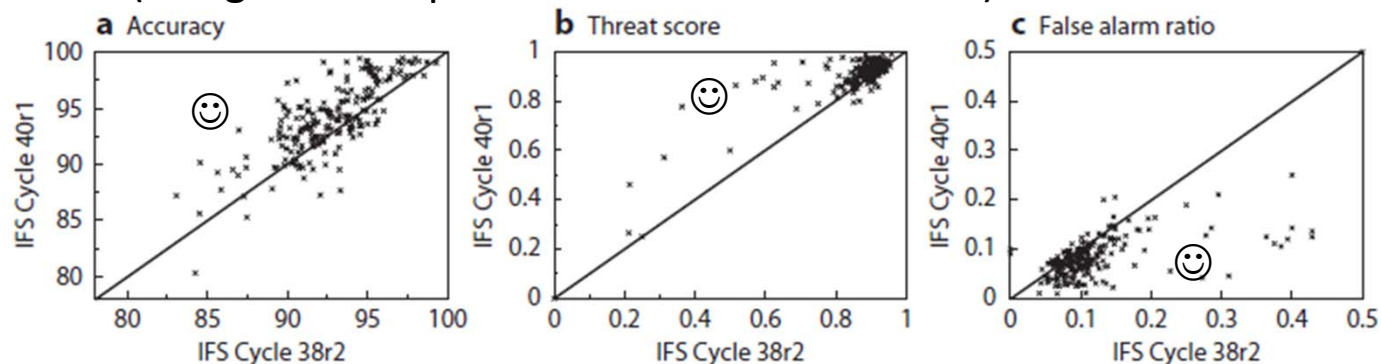
BG:	σ_b	= 3cm
SYNOP	σ_{SYNOP}	= 4cm
IMS	σ_{ims}	= 8cm

Snow analysis: Forecast impact

Revised IMS snow cover data assimilation (2013)



Impact on snow October 2012 to April 2013
(using 251 independent *in situ* observations)



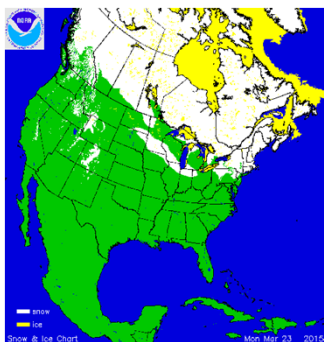
	Snow observed	No snow observed
Snow In analysis	a Hits	b False alarm
No snow In analysis	c Misses	d Correct no snow

The following scores are used for the evaluation:

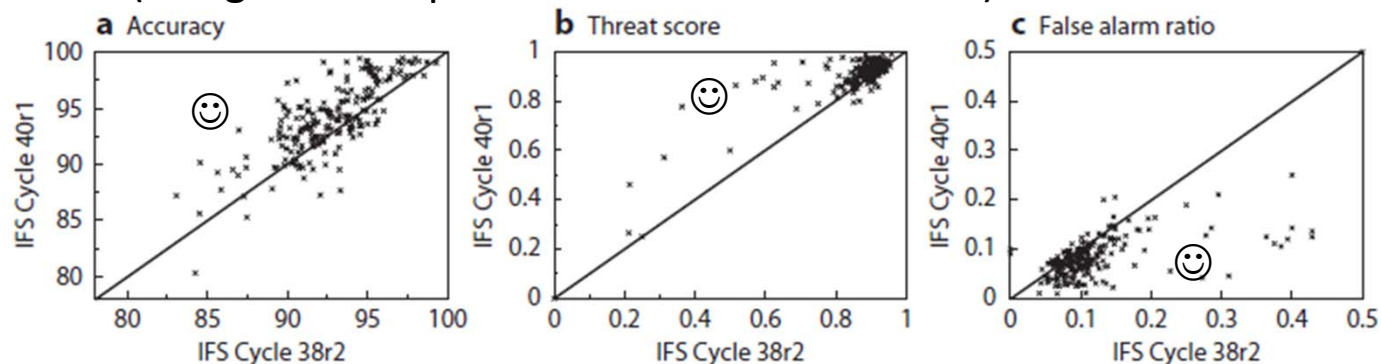
- Accuracy = $a + d / (a + b + c + d)$
- False alarm ratio = $b / (a + b)$
- Threat score = $a / (a + b + c)$

Snow analysis: Forecast impact

Revised IMS snow cover data assimilation (2013)

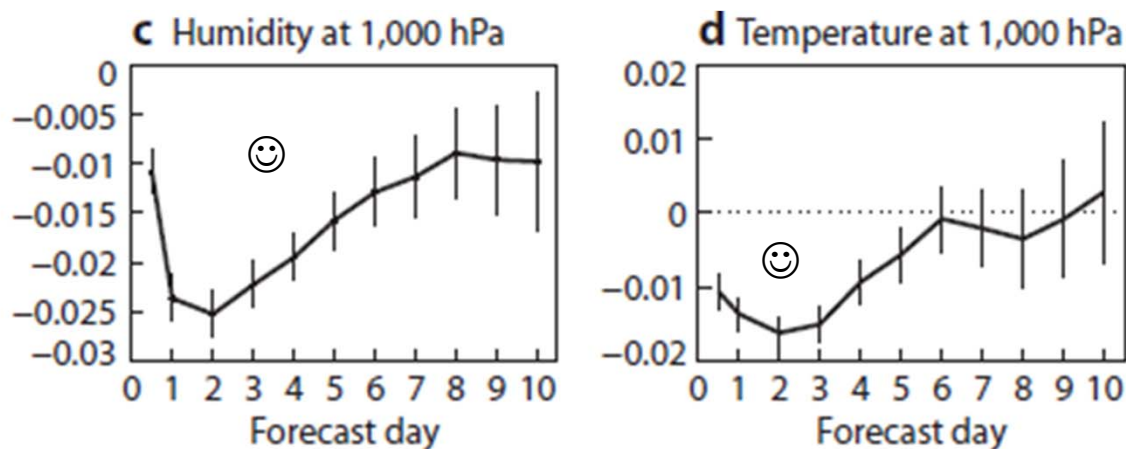


Impact on snow October 2012 to April 2013
(using 251 independent *in situ* observations)



Impact on atmospheric forecasts

October 2012 to April 2013 (RMSE new-old)



de Rosnay et al., ECMWF
NL 143, Spring 2015

Simplified EKF soil moisture analysis

For each grid point, analysed soil moisture state vector x_a :

$$x_a = x_b + K(y - H[x_b])$$

x background soil moisture state vector,
 H non linear observation operator

y observation vector

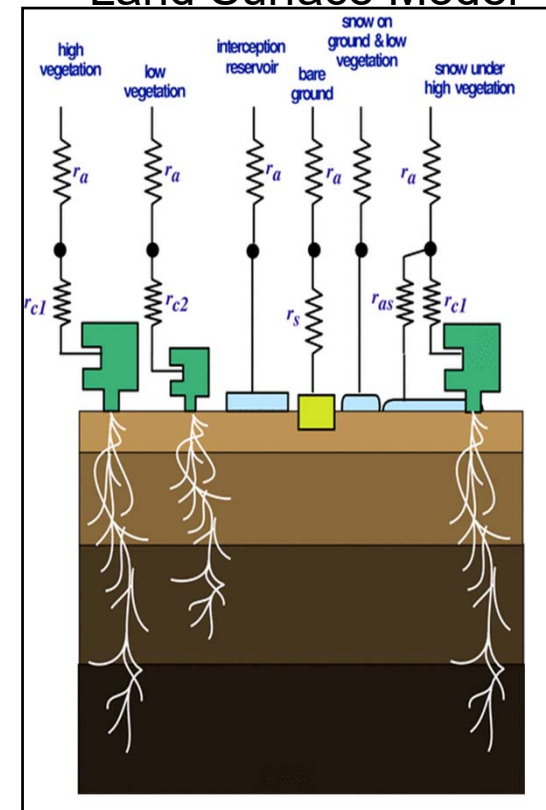
K Kalman gain matrix, fn of
 H (linearisation of H), P and R (covariance matrices
of background and observation errors).

Used at ECMWF (operations and ERA5), DWD, UKMO

Observations used at ECMWF:

- Conventional SYNOP pseudo observations (analysed T2m, RH2m)
- Satellite MetOp-A/B ASCAT soil moisture
- SMOS TB Data at 30, 40, 50 degrees

The simplified EKF is used to corrects the soil moisture trajectory of the Land Surface Model



Drusch et al., GRL, 2009

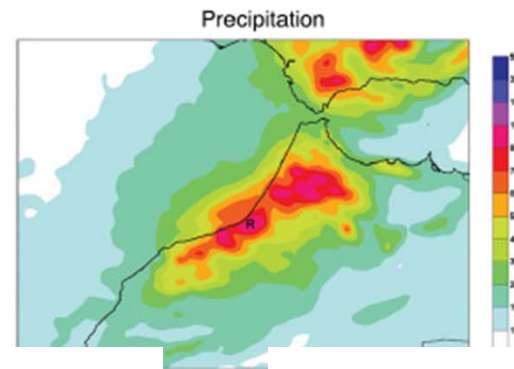
de Rosnay et al., ECMWF News Letter 127, 2011

de Rosnay et al., QJRMS, 2013

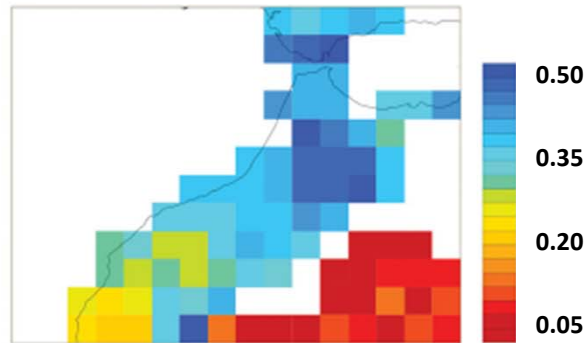
SMOS and ASCAT monitoring

Case study that illustrates the relevance of SMOS and ASCAT to monitor soil moisture in extreme conditions

Flash flood in
Morocco 23
February 2017

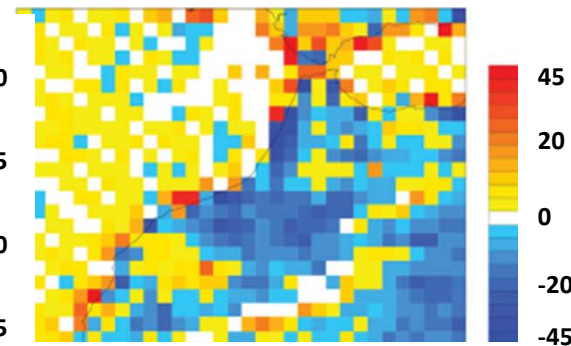


ASCAT Soil Moisture



ASCAT Soil Moisture in m^3/m^3
(red is dry / blue is wet)
→ Saturated

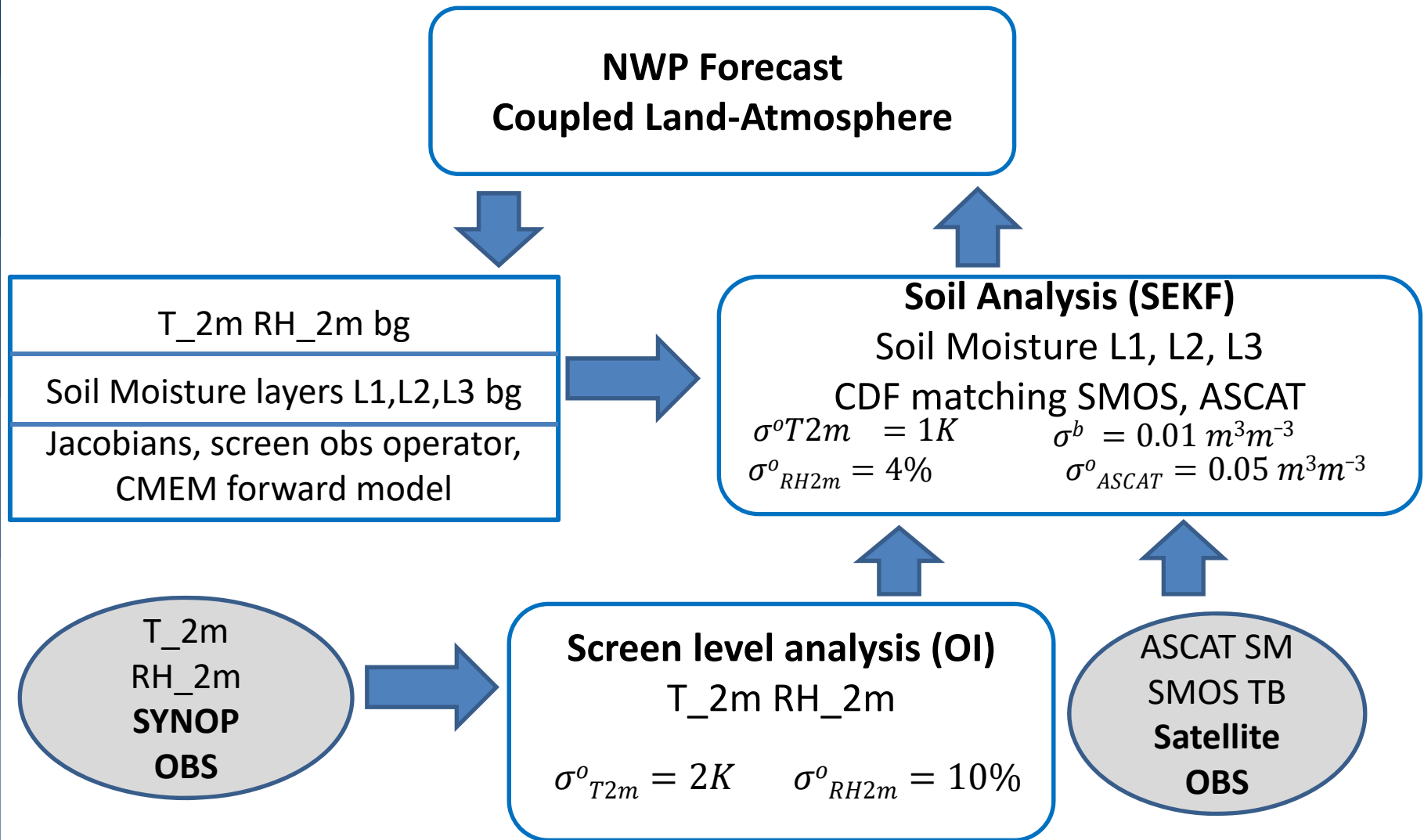
SMOS Brightness Temp



FG depar in K
(Blue means SMOS is wetter than ECMWF)

SMOS colder/wetter than FG
-> ECMWF FG drier

Soil analysis in the IFS

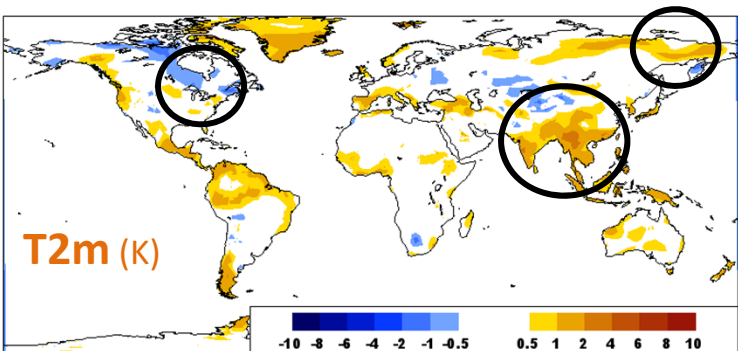
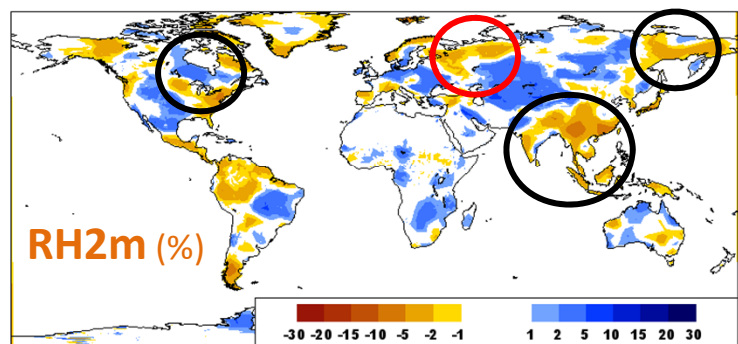
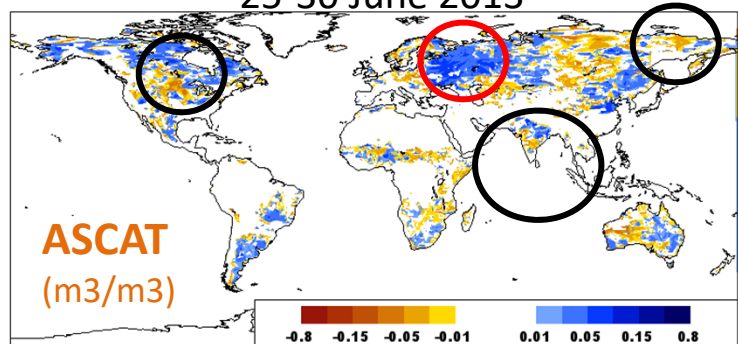


→ Operational soil moisture data assimilation: combines SYNOP and satellite data

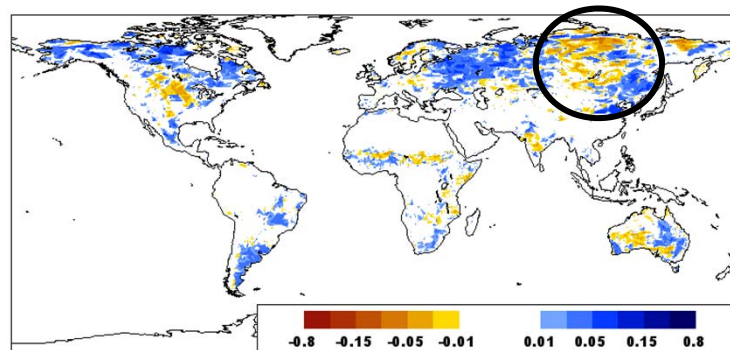
ASCAT Soil Moisture data assimilation

Innovation (Obs- model)

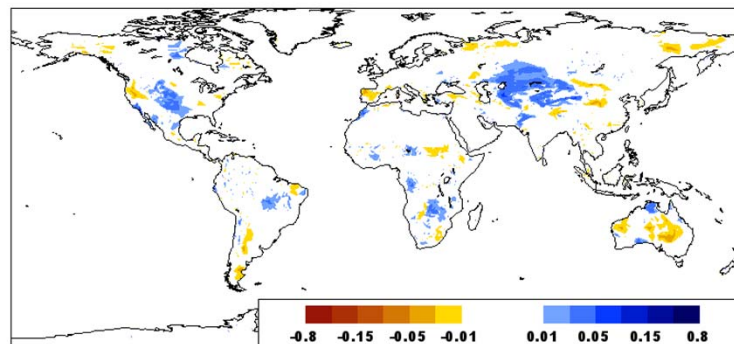
25-30 June 2013



Accumulated Increments (m³/m³)
in top soil layer (0-7cm)

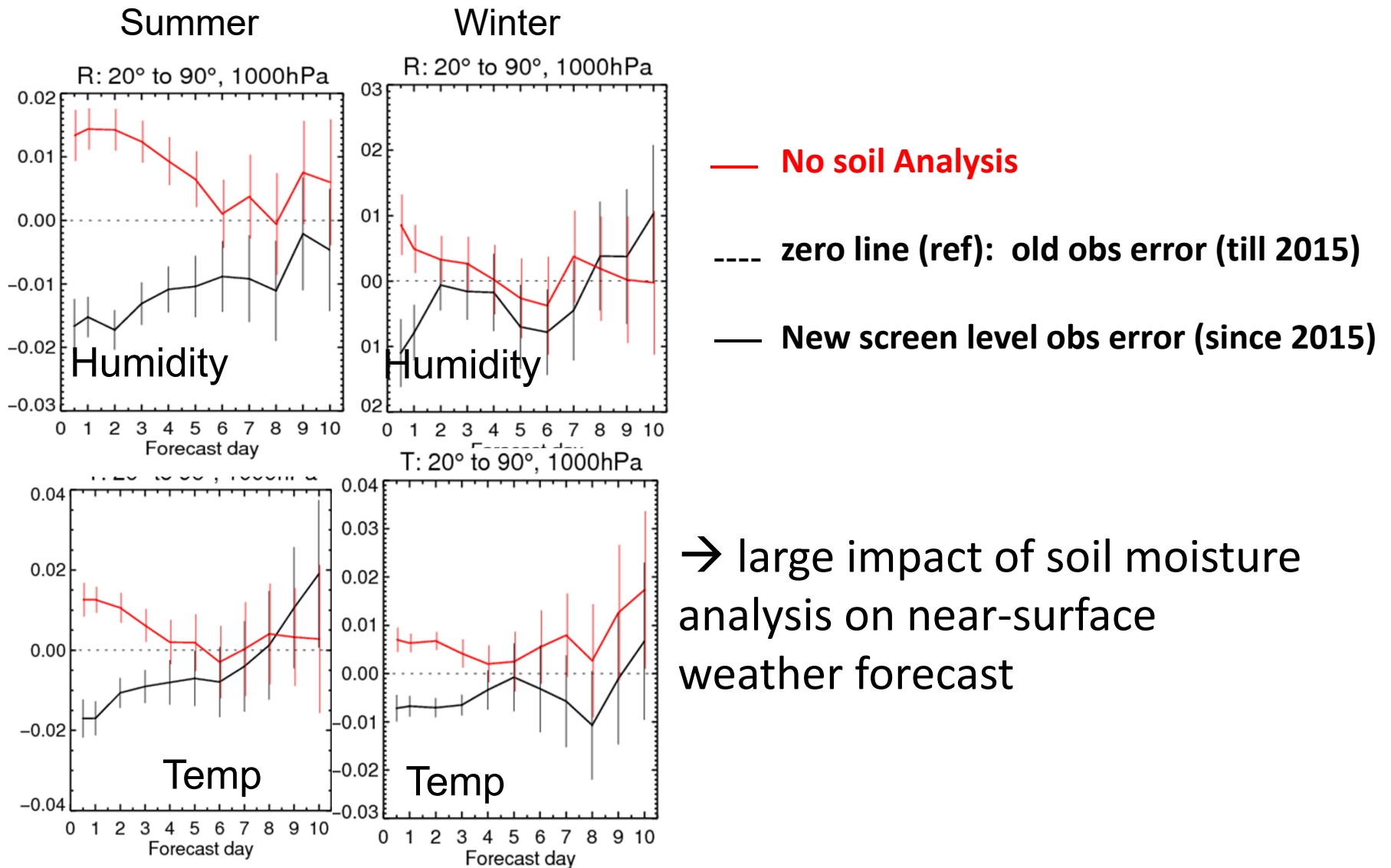


Due to ASCAT



Due to SYNOP T2m and RH2m

Soil Analysis for NWP: Impact on the forecast ?



ECMWF new Re-analysis (ERA5)

Assimilation of Scatterometer soil moisture data ERS/SCAT and MetOpA/B ASCAT

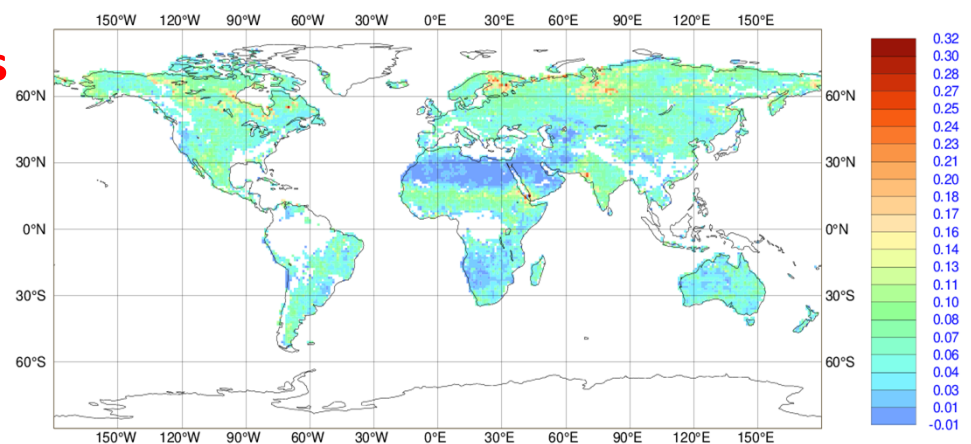
Use of EUMETSAT ASCAT-A reprocessed data (25km sampling)

	FG departure Mean m^3m^{-3}	FG departure StDev m^3m^{-3}	(FMA 2010)
Using NRT ASCAT	0.013	0.05	
Using Reproc ASCAT	0.006	0.044	

→ Reprocessed ASCAT has reduced background departure statistics both in mean and Stdev

ERA5 on-going production (Copernicus C3S)

**ASCAT surface soil moisture
STDev First guess departure (Obs-
Model) in m^3/m^3
for JJAS 2014**



EUMETSAT H-SAF soil moisture

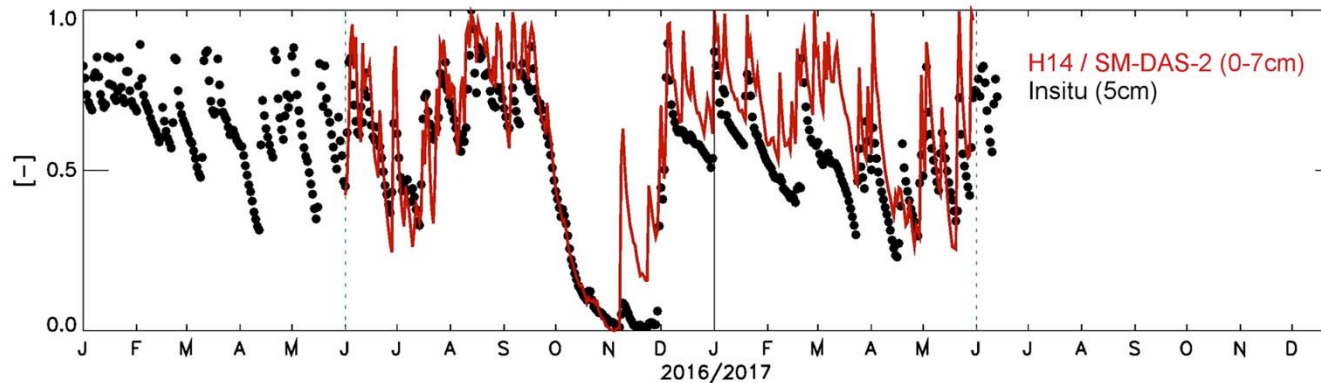
Scatterometer root zone soil moisture based on data assimilation

Evaluation of SM-DAS-2/H14

Fairbairn, Albergel et al.

Surface and root zone liquid soil moisture content

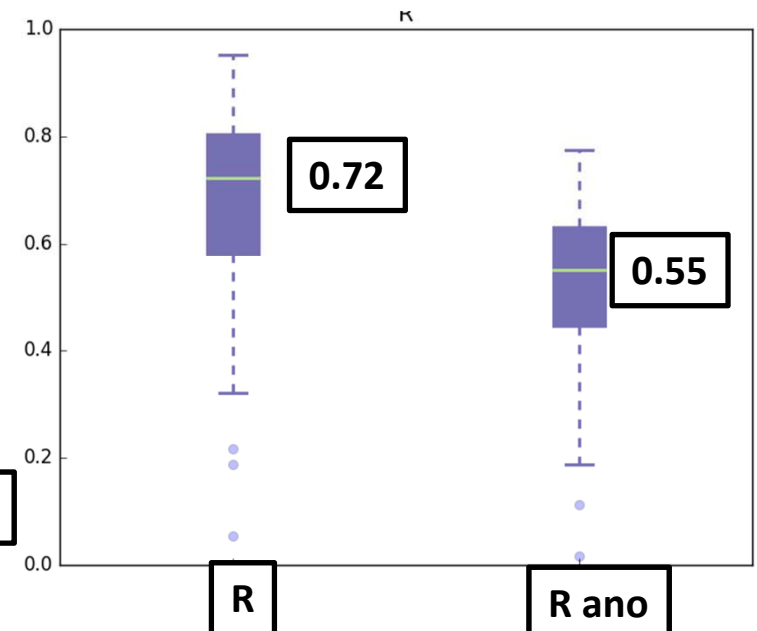
USCRN network, LaFayette, LA, USA, 30.0900N 91.8700W



Observation (5cm)
H14/SM-DAS-2 (0-7cm)

Accuracy requirements for product SM-DAS-2 [R]

Unit	Threshold	Target	Optimal
Dimensionless	0.50	0.65	0.80



USCRN, 102 stations used

The EUMETSAT
Network of
Satellite Application
Facilities



EUROPEAN CENTRE FOR MEDIUM-RANGE WEATHER FORECASTS

© ECMWF

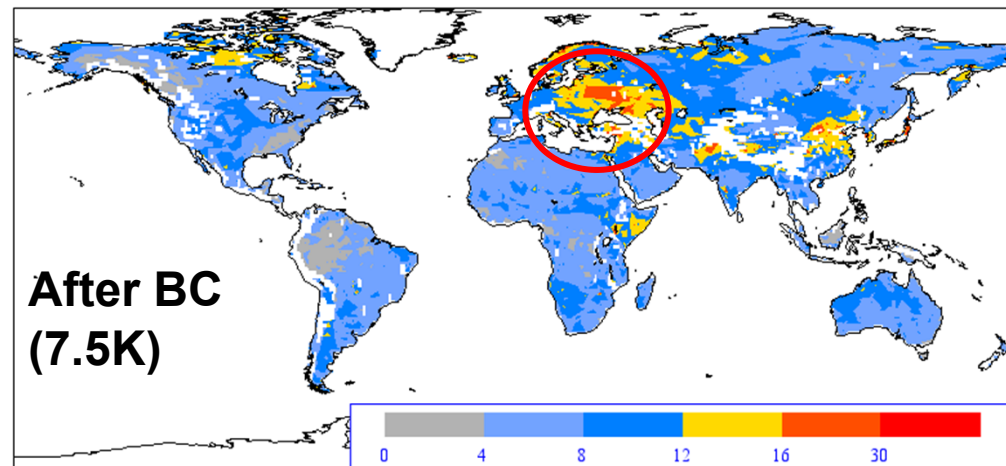
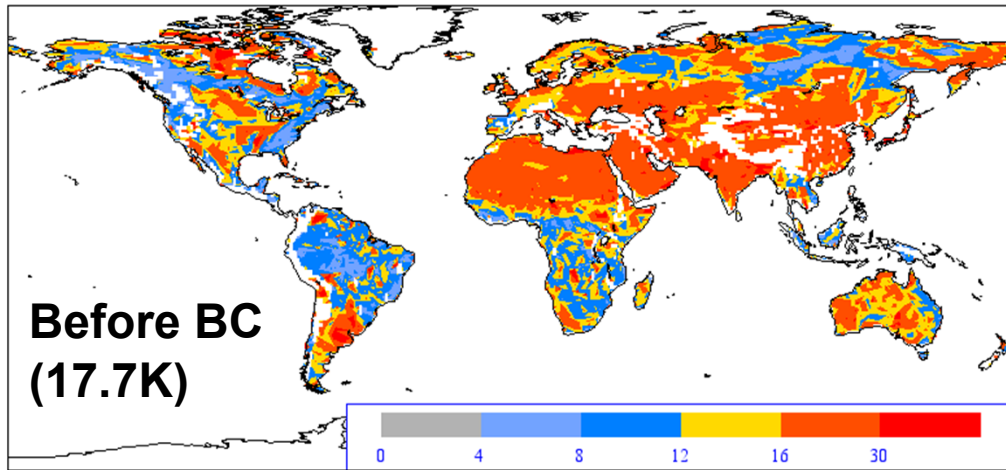
ECMWF L-band TB Bias correction

RMSD between SMOS Obs and ECMWF

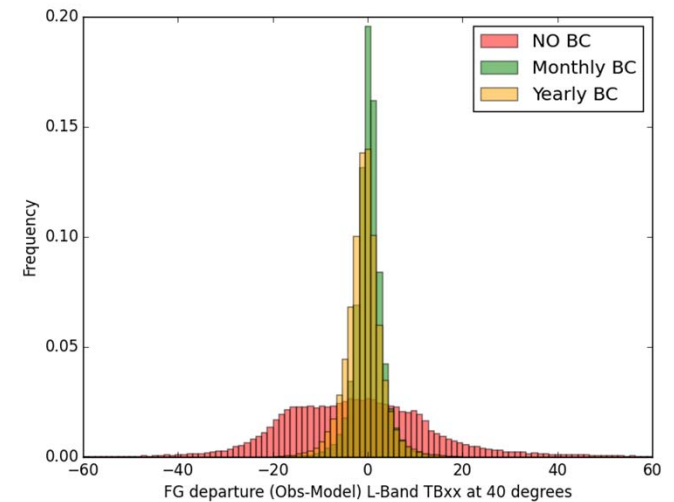
2012

RMSD (K)

TBxx, 40 degrees

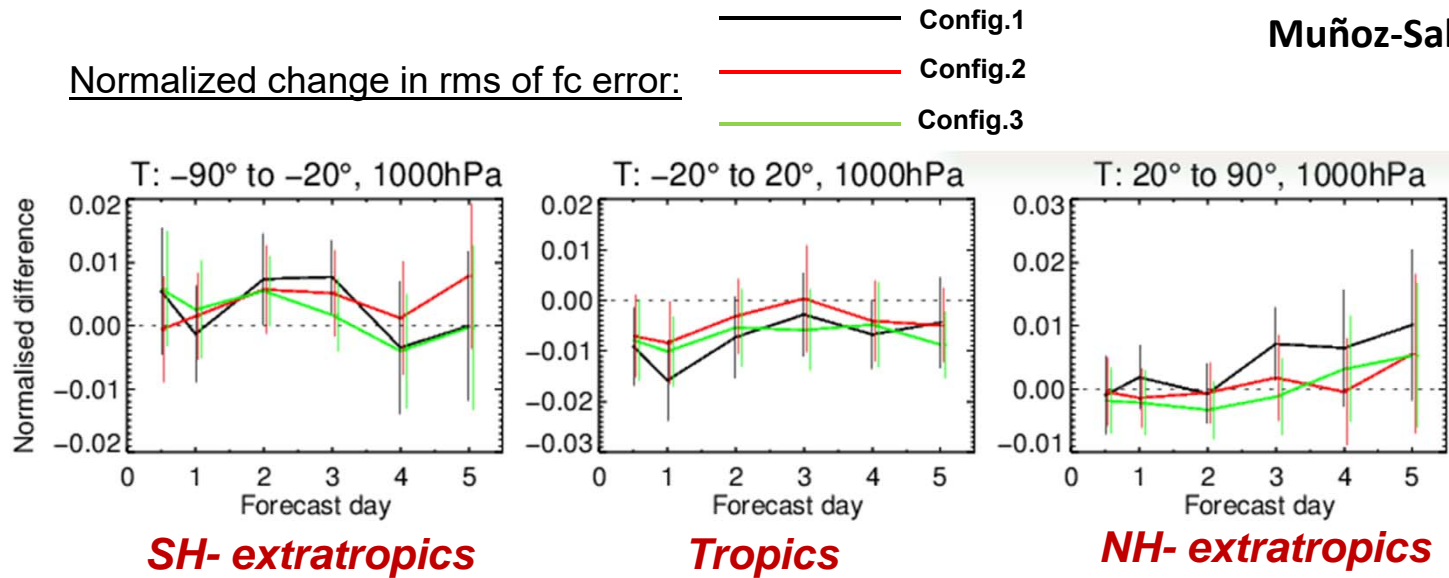


Low residual RMSD, except in RFI areas



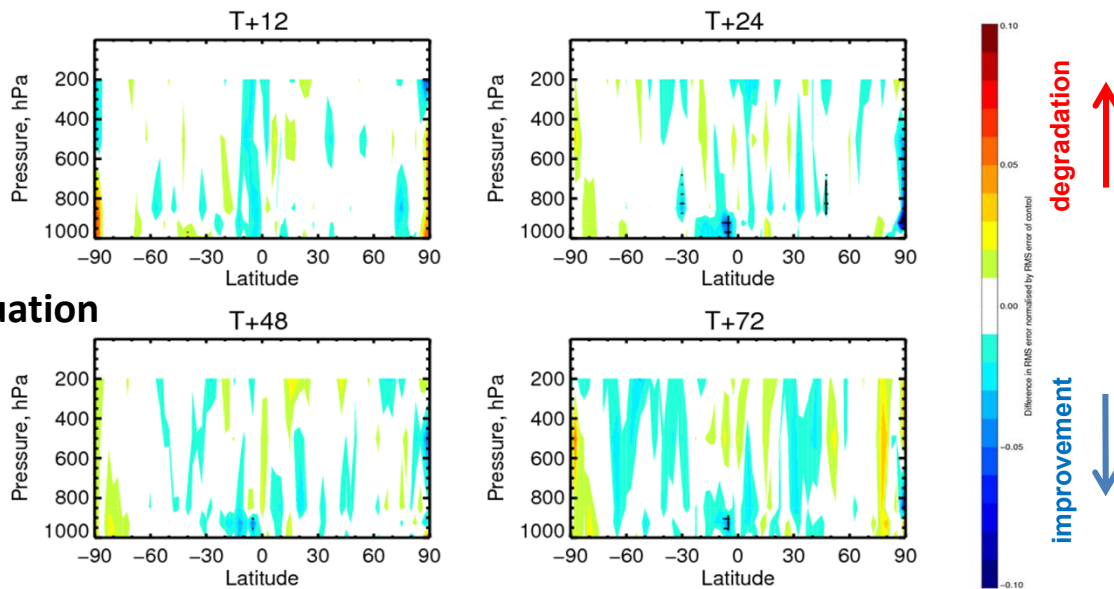
SMOS data assimilation impact on atmospheric scores

Muñoz-Sabater et al.



Configuration 3

Based on short experiments
Longer experiment under evaluation



EUROPEAN CENTRE FOR MEDIUM-RANGE WEATHER FORECASTS

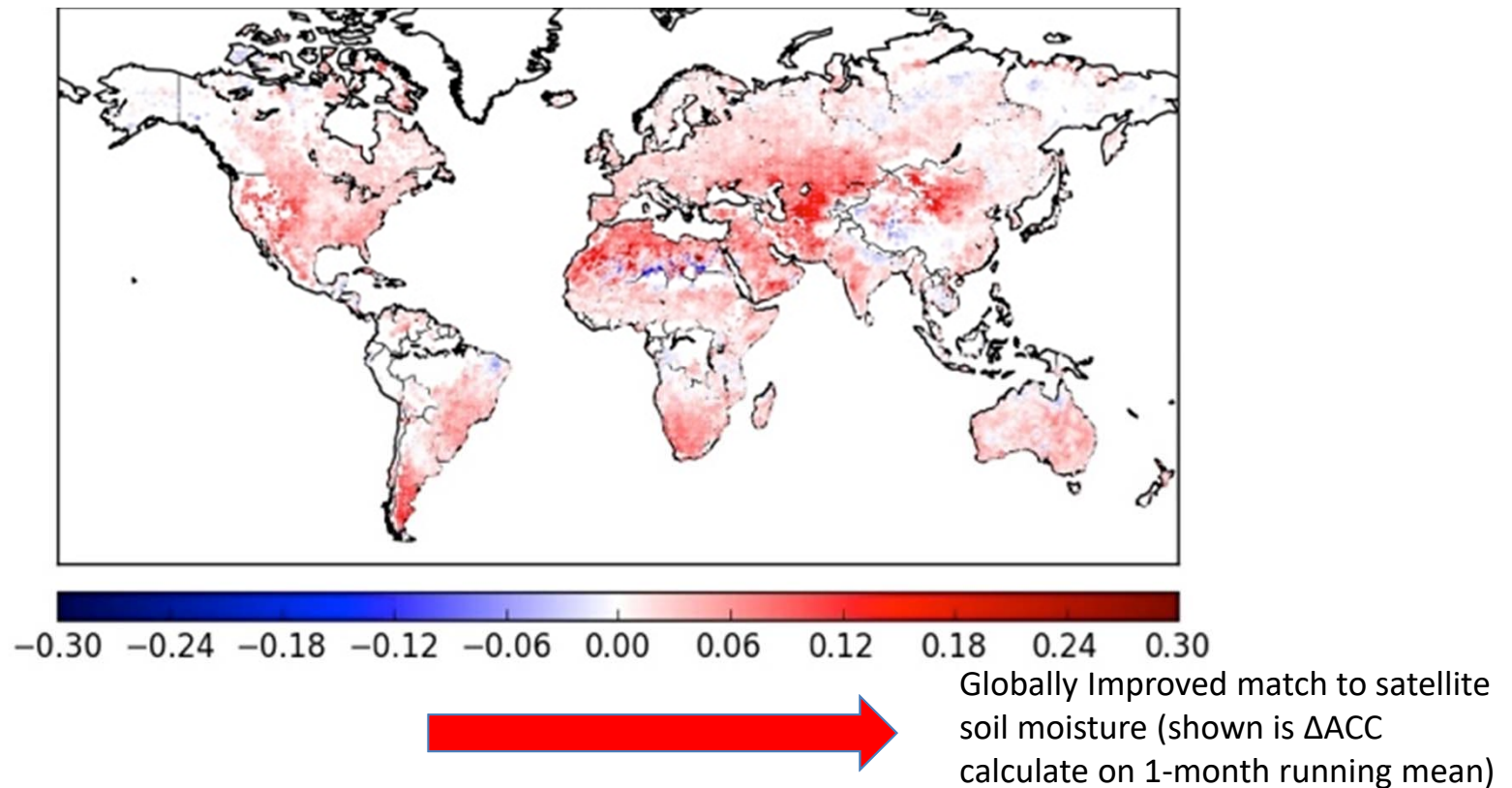
© ECMWF

Impact of soil vertical resolution for satellite soil moisture

Tests with H-TESEL soil resolution increased: top layer 0-7cm replaced by 3 layers 0-1cm, 1-3cm, 3-7cm

Impact on Anomaly Correlation with ESA-CCI satellite soil moisture

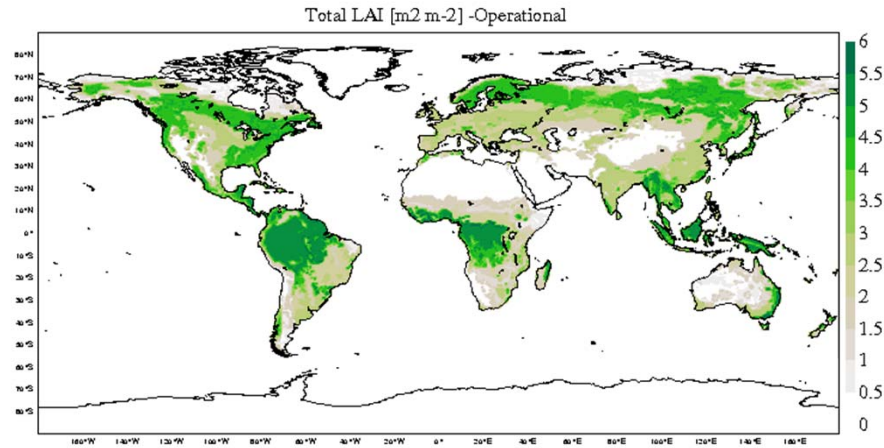
(Albergel, Balsamo)



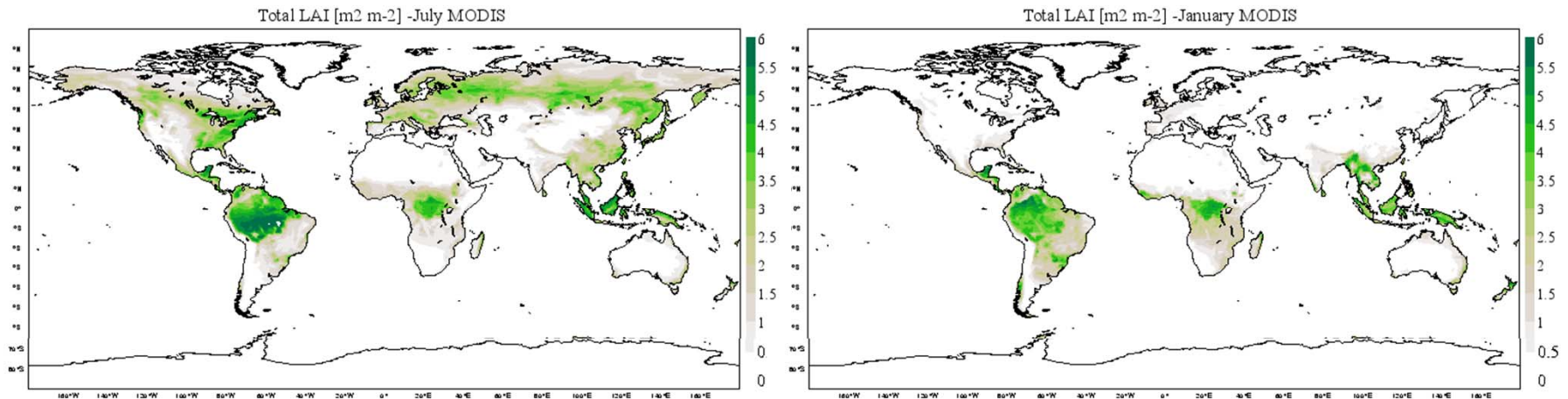
Anomaly correlation (1988-2014) measured with ESA-CCI soil moisture remote sensing (multi-sensor) product.

→ Provides a global validation of the usefulness of increase soil vertical resolution.

Seasonal Varying Leaf Area Index



Boussetta, Balsamo et al

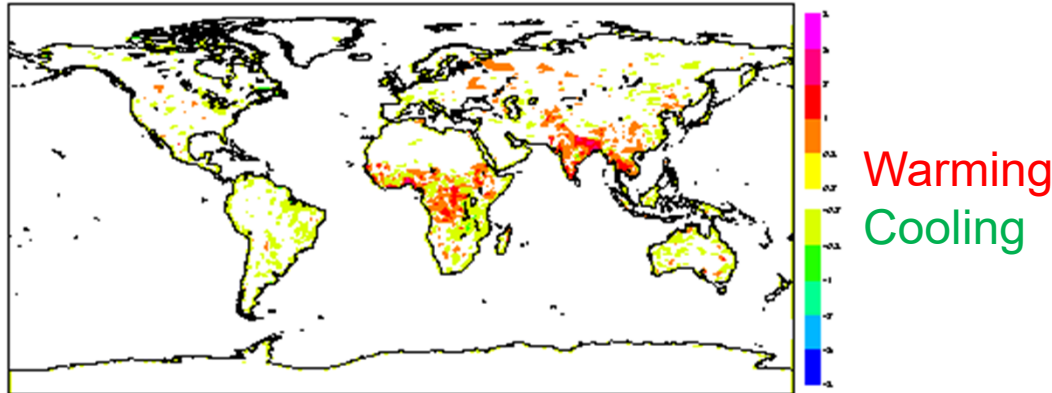


derived 8years (2000-2008) climatological time series from MODIS S5 products

Satellite-based LAI climatology introduce a more realistic seasonal variability of the vegetation state compared to the constant LAI map which used to overestimate LAI especially in winter and during the transition periods of spring and autumn

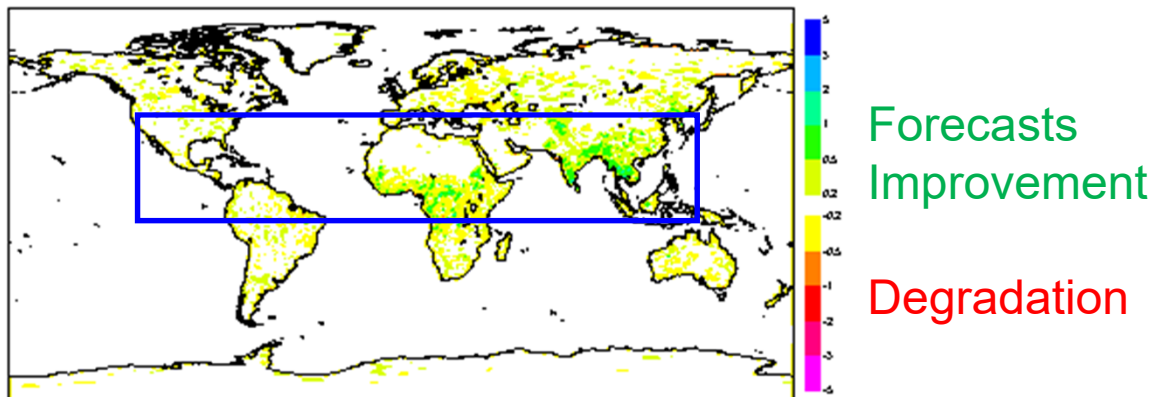
Seasonal Varying Leaf Area Index: Impact on T2m forecasts

Sensitivity T2m (New-Old)



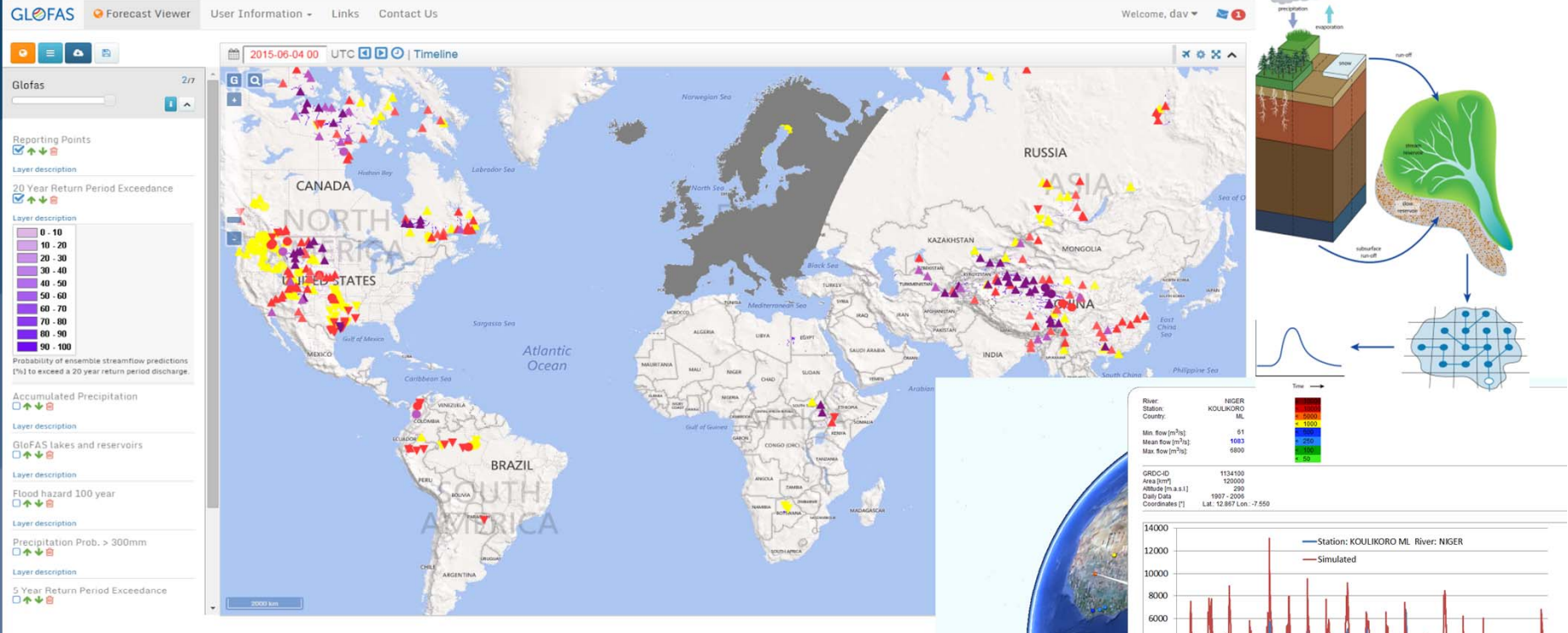
Experiments Feb-Sept 2008
- Old (fixed LAI)
- New (LAI seasonal)

Impact T2m MEA Old – New

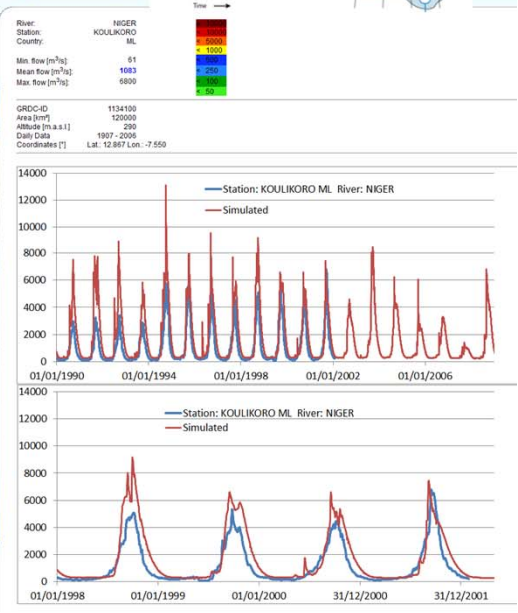


- Satellite LAI → a consistent warming seen in FC36h (12UTC) due to reduction of LAI in spring (reduced ET).
- Beneficial impact on near surface temperature forecast by reducing t2m bias by ~0.5degree

The Global Flood Awareness System



Output from global ECMWF NWP land-surface forecast is fed into a routing model (Simplified LISFLOOD (JRC)) to produce flood forecasts – benefiting from all the improvements in the ECMWF Integrated Forecasting System (model and assimilation)!



Summary (1/2)

Satellite data used for snow and soil moisture in the ECMWF IFS

- Snow: NOAA NESDIS/IMS 4km snow cover data (multi-sensor product).
No use of Snow Water Equivalent products used for NWP
- Soil moisture: ASCAT-A/B IFS DA operational
- L-band TB: SMOS IFS Monitoring operational, SMAP Early Adopter
- SMOS SM: NRT (NN) processor implementation, offline NN SM DA tests
- Reanalyses: ERA5 use of Scatterometer series ERS/SCAT and Metop ASCAT
- Root zone retrieval from ASCAT (H-SAF): H14 (NRT) and H27 Climate data record

Summary (2/2)

- Flood forecasts: benefits from overall improvements in the ECMWF IFS, including soil and snow data assimilation.
- Current developments:
 - Hybrid EDA-SEKF analysis for stronger land-atmosphere assimilation coupling (Quasi SCDA)
- Future
 - SEKF in OOPS
 - Use of MW data to analyse snow depth
 - LST DA in the SEKF
 - Future WCOM mission relevant for both SWE and SM
 - Integrated hydrological variables such as river discharges
- Observation latency : crucial for NWP applications (<3h)
- In situ data: essential for DA (snow, T2m, etc) and evaluation (SM)

ECMWF/ESA Workshop on Using Low Frequency Passive Microwave Measurements in Research and Operational Applications

ECMWF | Reading | 4-6 December 2017

[Learning homepage](#)

[Training](#)

[Workshops](#)

[Poster guidelines](#)

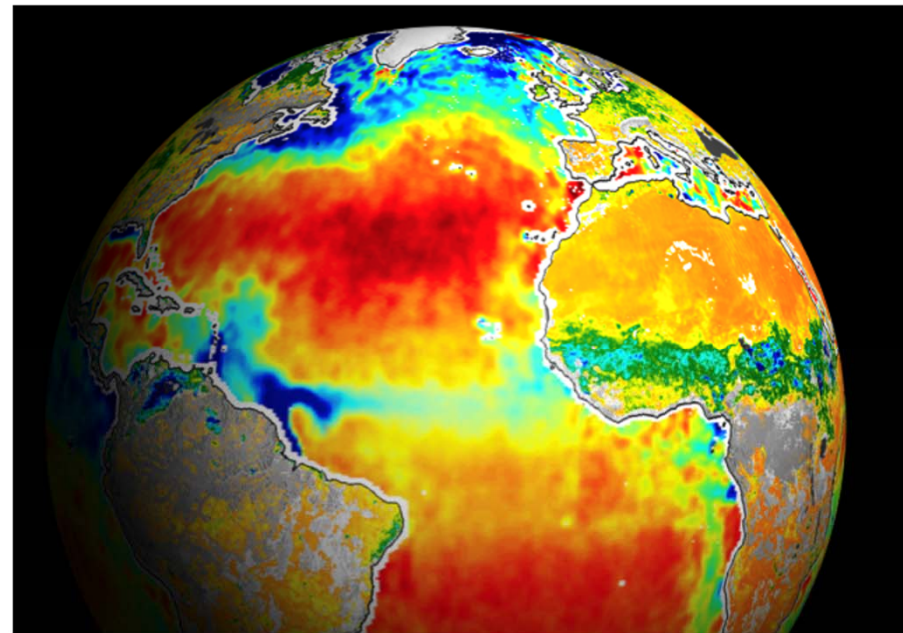
[Past workshops](#)

[Seminars](#)

[Education material](#)

[Programme \(to follow\)](#)

[Local information](#)

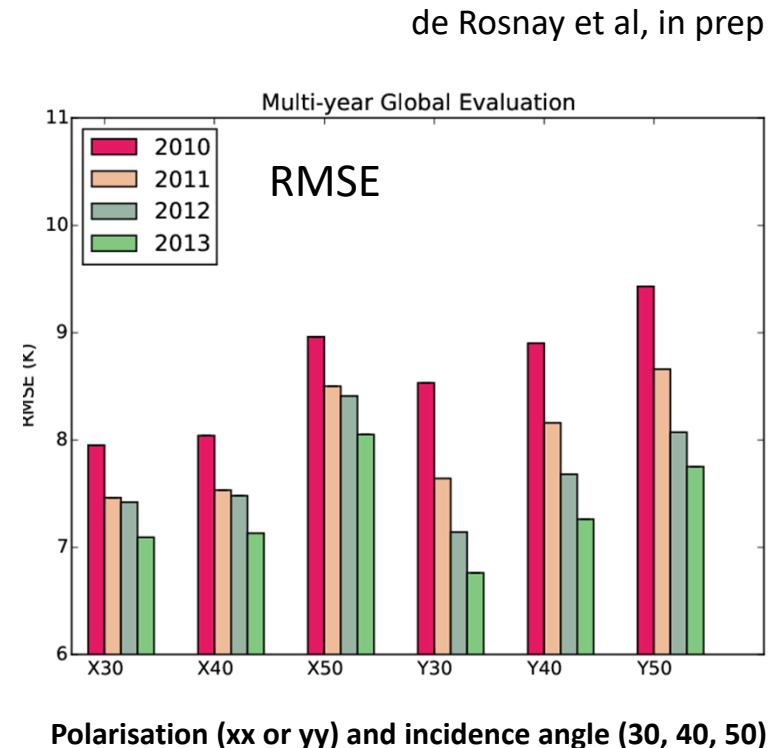
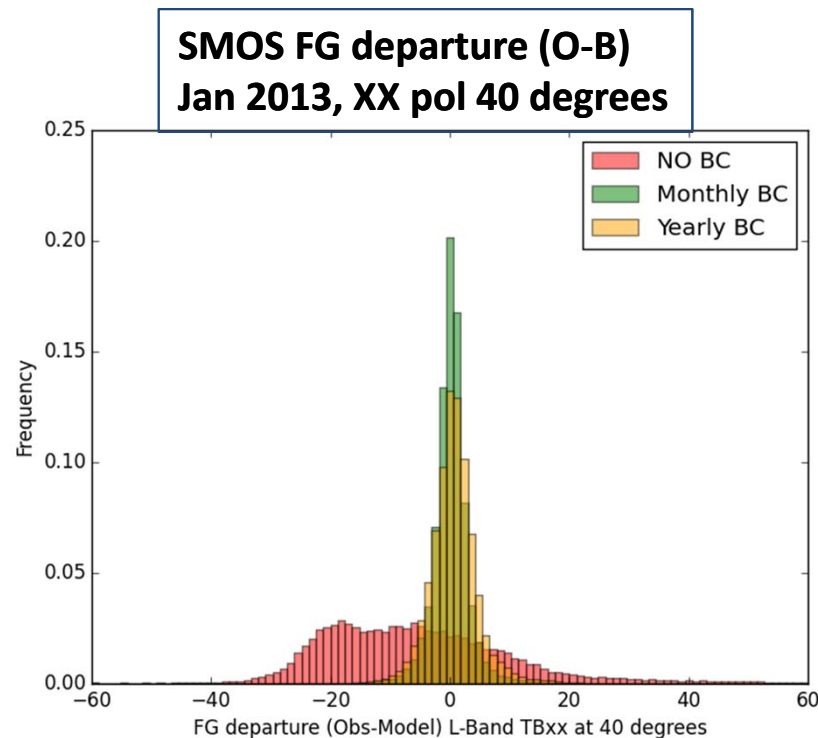


© ESA/CATDS

<https://www.ecmwf.int/en/learning/workshops/workshop-using-low-frequency-passive-microwave-measurements-research-and-operational-applications>

SMOS Forward modelling and Bias correction

- CMEM: ECMWF Community Microwave Emission Modelling Platform
→ produce reprocessed ECMWF SMOS TB for 2010-2013
- Comparison between ECMWF TB and SMOS NRT TB (both reprocessed)
- **Consistent improvement of SMOS data at Pol xx and yy, for incidence angles 30, 40, 50 degrees**



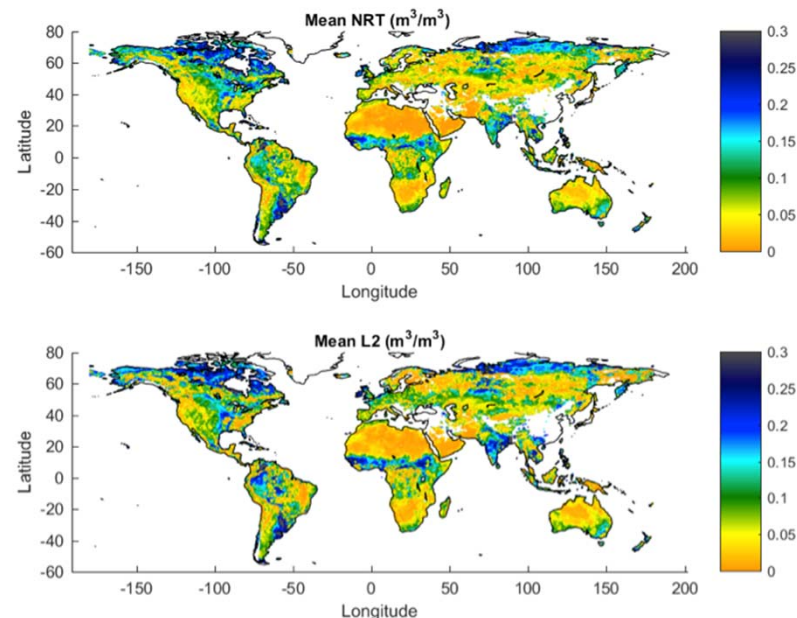
New level 2 SMOS NRT Soil Moisture product based on Neural Networks

Designed by CESBIO/Estellus, Implemented by ECMWF (Rodriguez et al, HESS 2017)

- Neural Network used to retrieve SMOS L2 SM from NRT brightness temperature
- Trained on SMOS L2 Soil moisture

→ NRT (4h latency) SMOS L2 SM

- Available in NetCDF, since March 2016 on ESA SMOS Online Dissemination service <https://smos-ds-02.eo.esa.int/oads/access> also on EUMETCAST and GTS



Comparison between L2 NRT and L2 v6.20 soil moisture

Evaluation against in situ stations (USCRN and SCAN)

Input	STD	R	Bias
NN	0.049	0.55	-0.024
SMOS L3	0.064	0.50	-0.026

