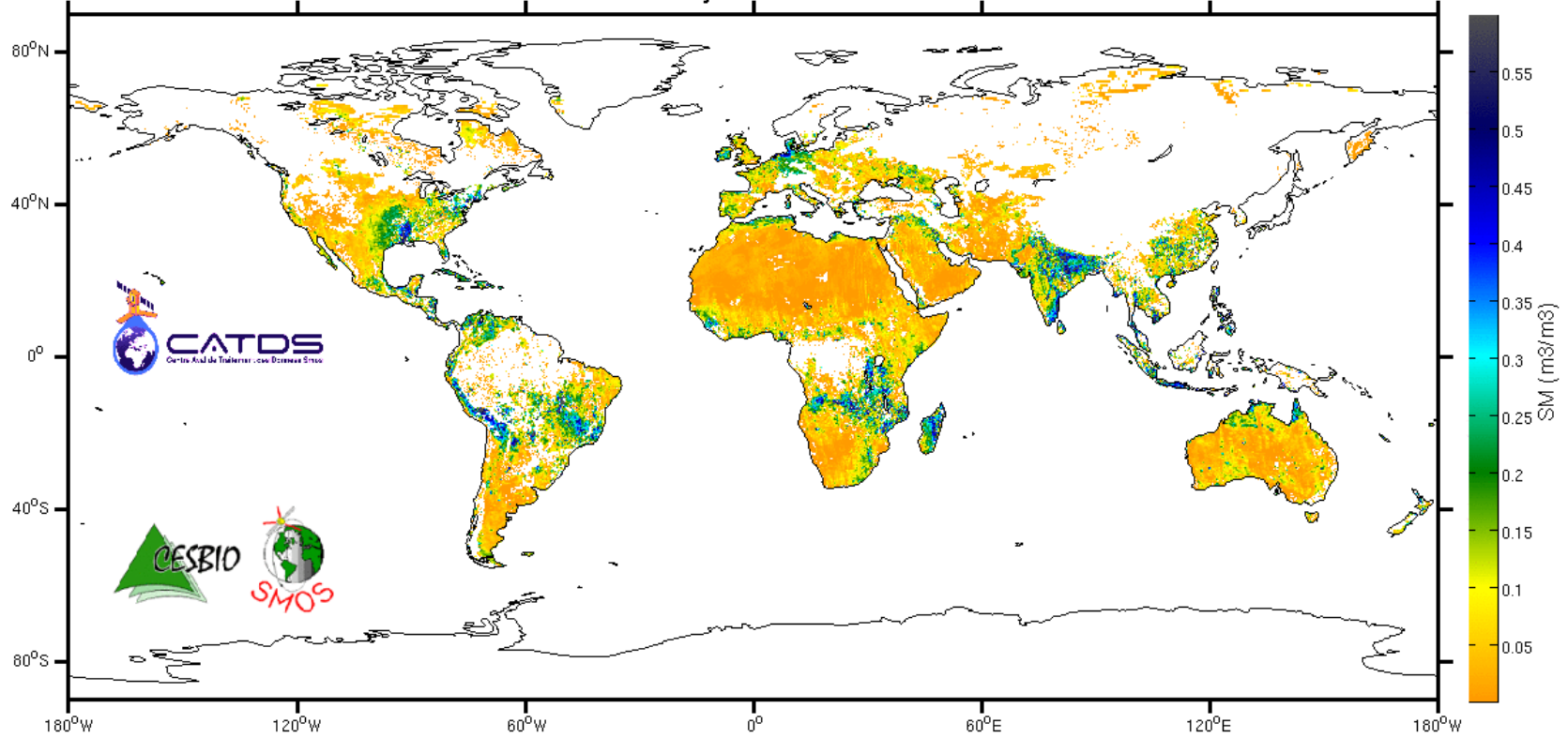




SMOS Land products developments and applications

Yann Kerr (CESBIO) and the ESL's
Steven Delwart, Matthias Drush and susanne
Mecklenburg

Soil Moisture Level 3 3-Day 20120101-20120103





Layout

- q SMOS status
- q Some results
- q Conclusion and way forward

SMOS

Principle and key points

- 2D **L band** Interferometric fully polarimetric radiometer
- Complete coverage of the globe in less than 3 days at both 6 am and 6 pm and multiangular acquisitions
- 43 km average (**real**) resolution
- Estimates of
 - Soil moisture, Vegetation opacity
 - Sea surface salinity
 - Wind speed (Hurricane)
 - Thin sea ice
 - drought, RZM
 -
- Launched november 2 2009





SMOS

Principle and key points

Soil Moisture
August 2011

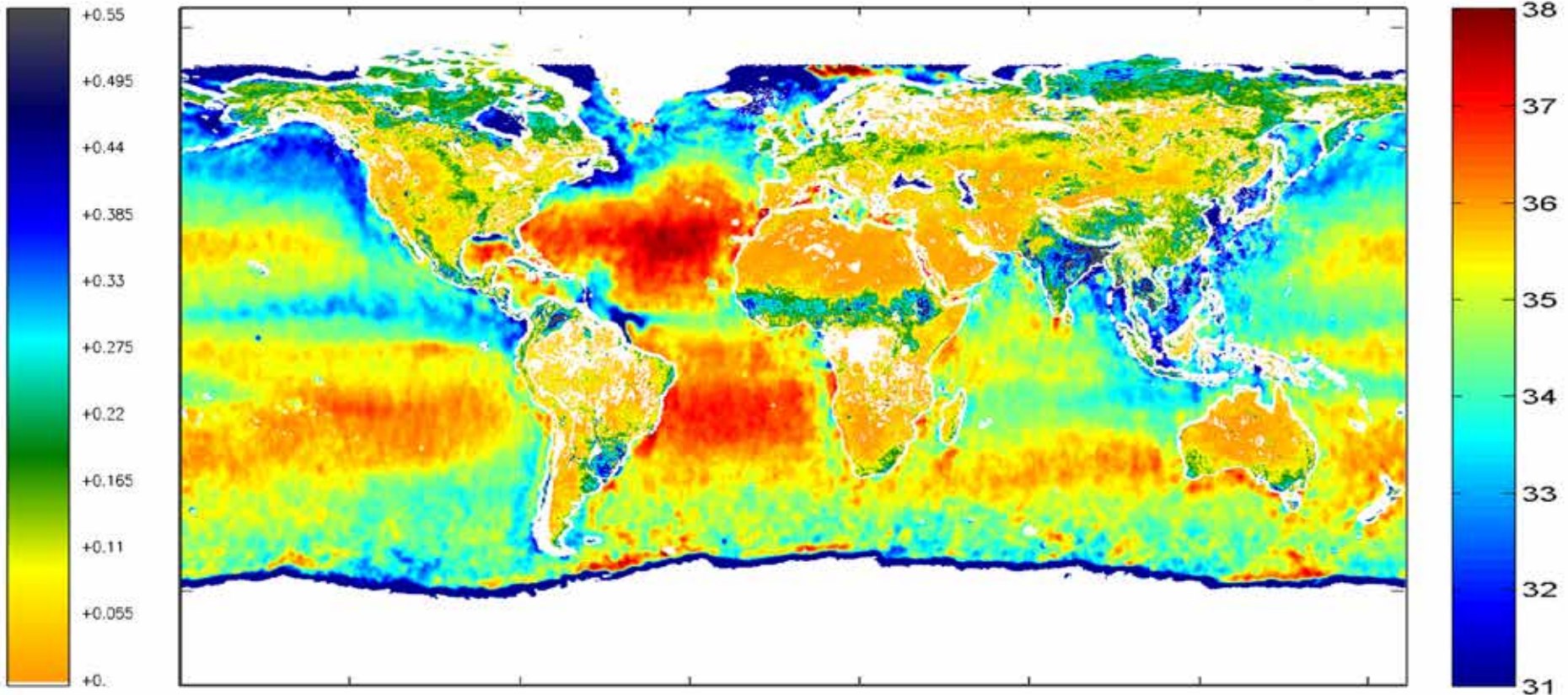
Ocean Salinity
August 2010



Ifremer



Morning orbits





A few facts

- q SMOS has now cumulated more than 4 years in operation (Jan 2010 to now)
- q SMOS Extension for two years granted
- q Operational Near real time Tb data disseminated
- q Operational SM to be implemented at ECMWF
- q Operational users of the data (latest is USDA)
- q Pre operational products under test

- q New products
 - √ Release of V620 this fall (ESA L1-L2) or now (CATDS L3)
 - o Improved L1
 - o Improved L2 / L3
 - √ Level 4
 - o Drought index
 - o Root zone soil moisture
 - o Precip, floods etc...
- q SMAP to be launched this November

Reprocessing

SMOS actually planning 2nd full mission reprocessing on v620.

L1 available next fall, L2 next winter.

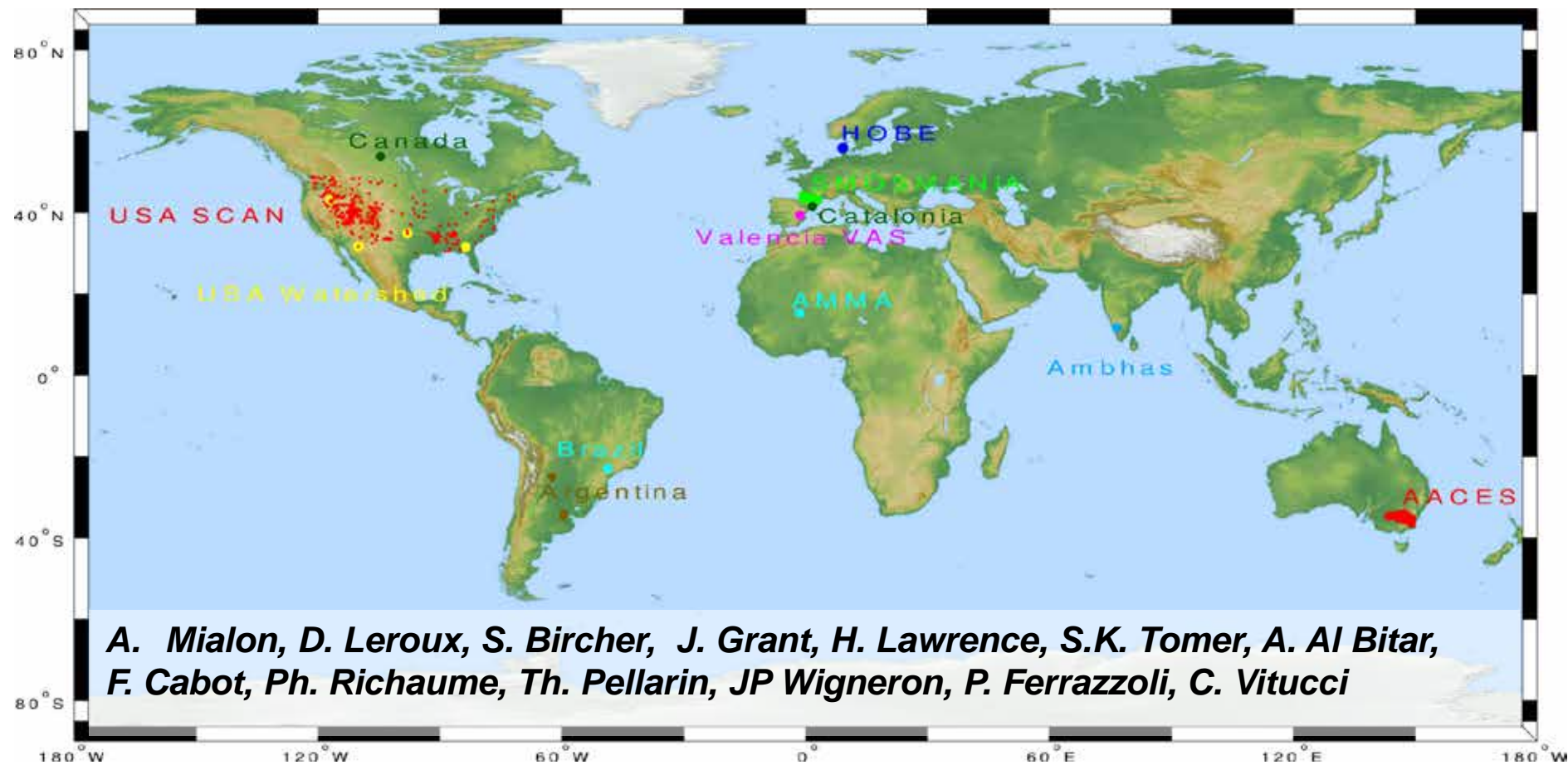
Based on preliminary results bias at Dome C down to below 1K

Aquarius soon distributing v3.0.

Cabot F.

	SMOS		Aquarius		Bias Aquarius-SMOS			Inter Version bias	
	V504	V611beta	V2.0	V2.7.1	V2.0/V504	V2.7.1/V611	V611	SMOS (v611-v504)	Aquarius (v271-v2)
H	202,16	205,40	207,87	206,16	5,71	0,76		3,24	-1,71
V	189,40	192,78	195,62	193,07	6,22	0,29		3,38	-2,55
H	208,70	210,57	212,43	210,65	3,73	0,08		1,87	-1,78
V	186,99	188,86	192,15	189,38	5,16	0,52		1,87	-2,77
H	211,22	213,88	215,63	213,68	4,41	-0,20		2,66	-1,95
V	182,04	184,91	187,47	185,14	5,43	0,23		2,87	-2,33

Cal val : Many in situ datasets



Collaborations : T. Jackson, R. Bindlish, E. Lopez, M. Sekhar, J. Walker, C. Rudiguer, E. Wood



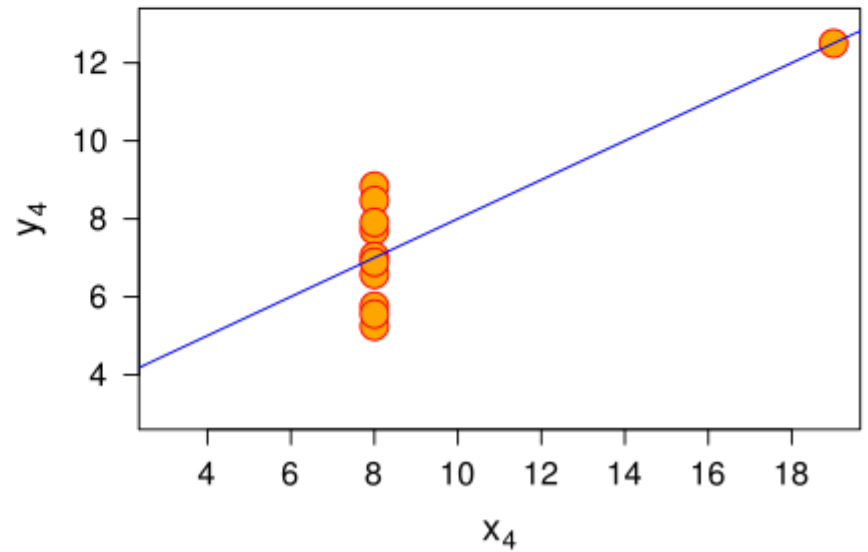
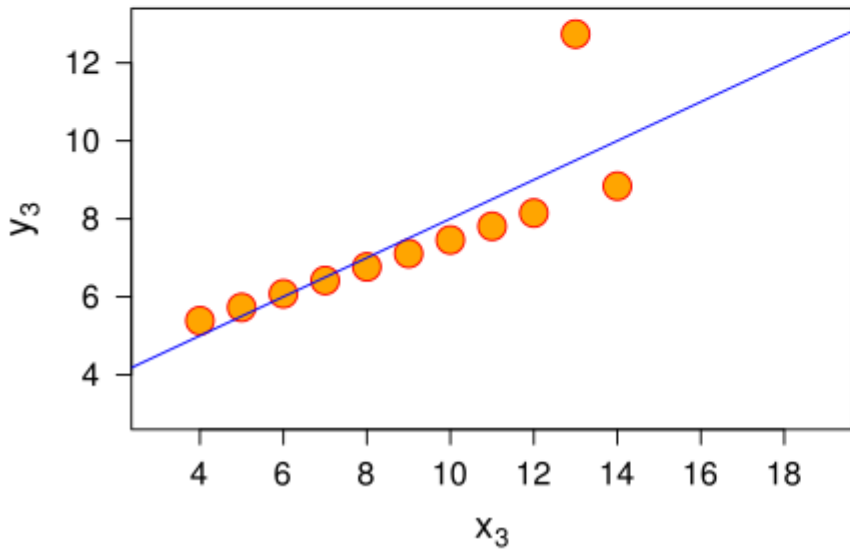
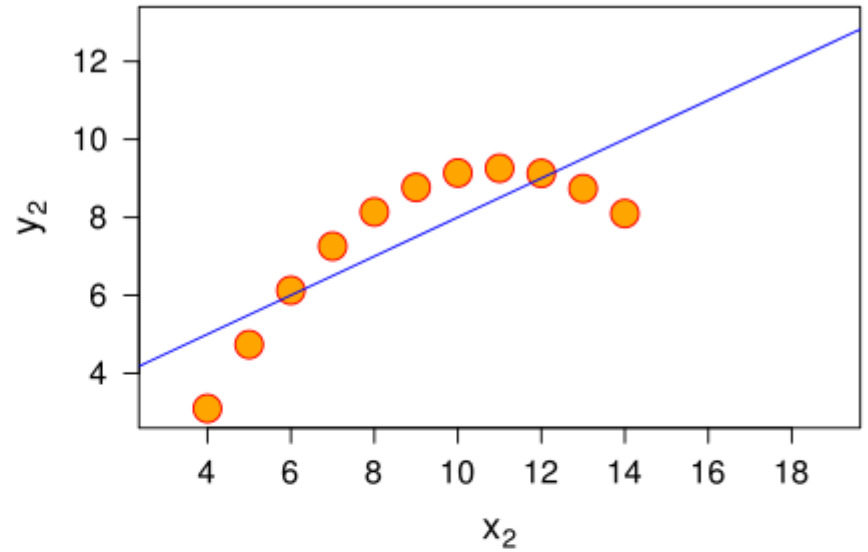
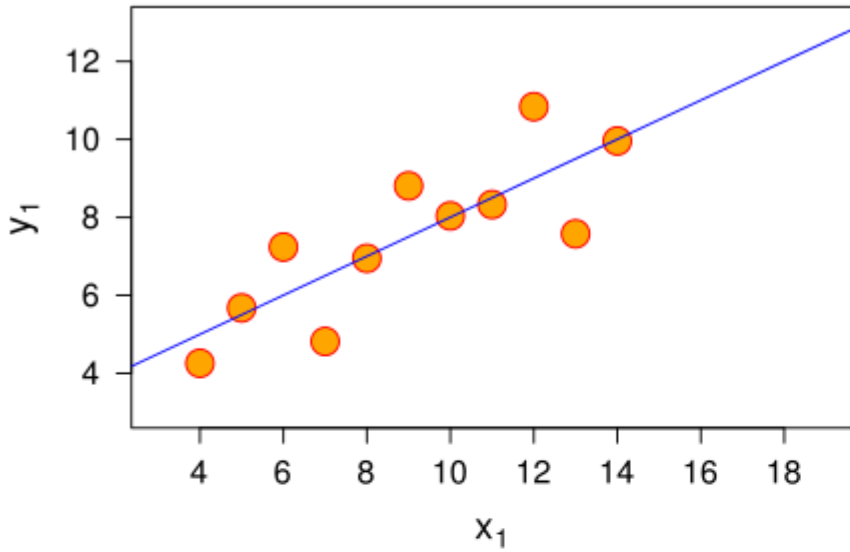
Cal Val Metrics?

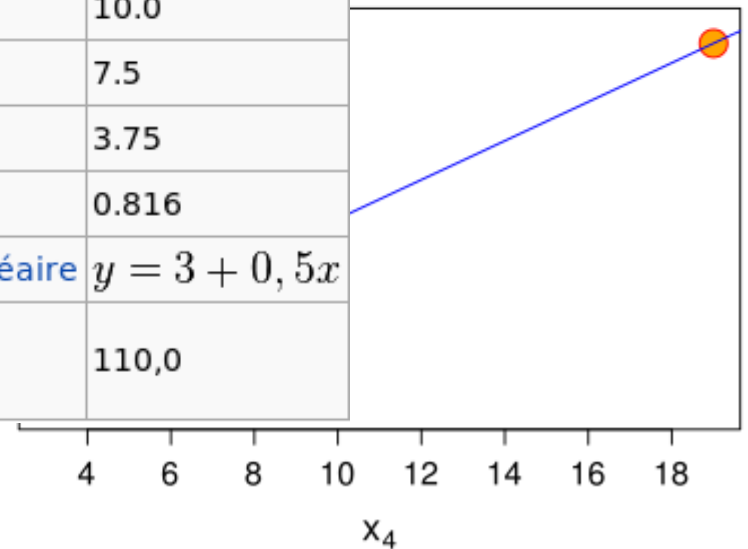
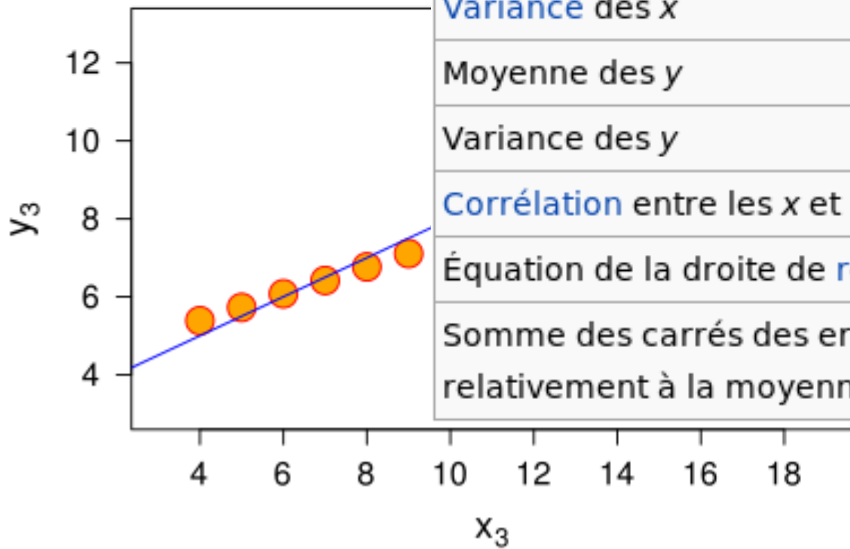
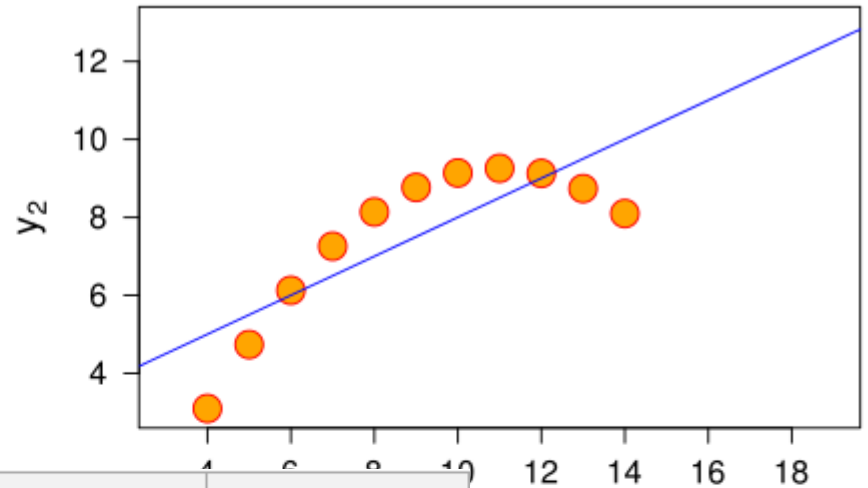
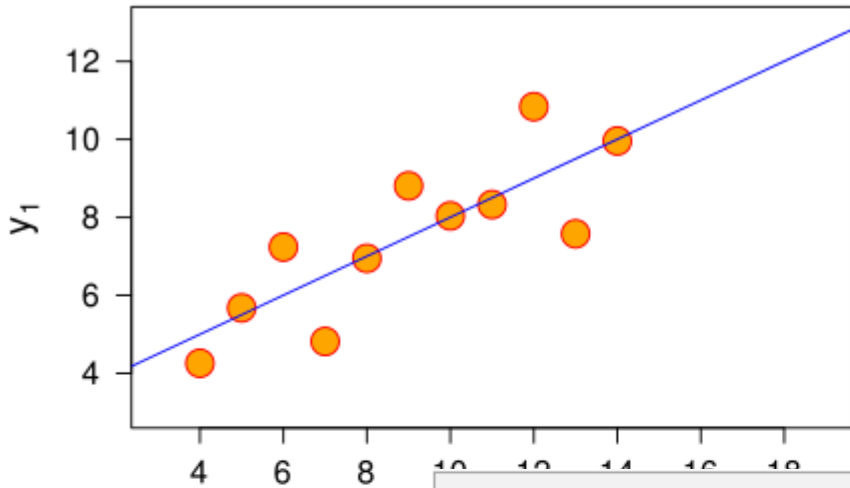
- q Correlation coefficient or RMSE?
- q Temporal or spatial?
- q A couple of provocative examples



Statistics – the solution?

- q Temporal evolution or statistics
 - ✓ 11 measurements ... which is the best (Ascombe quartet)

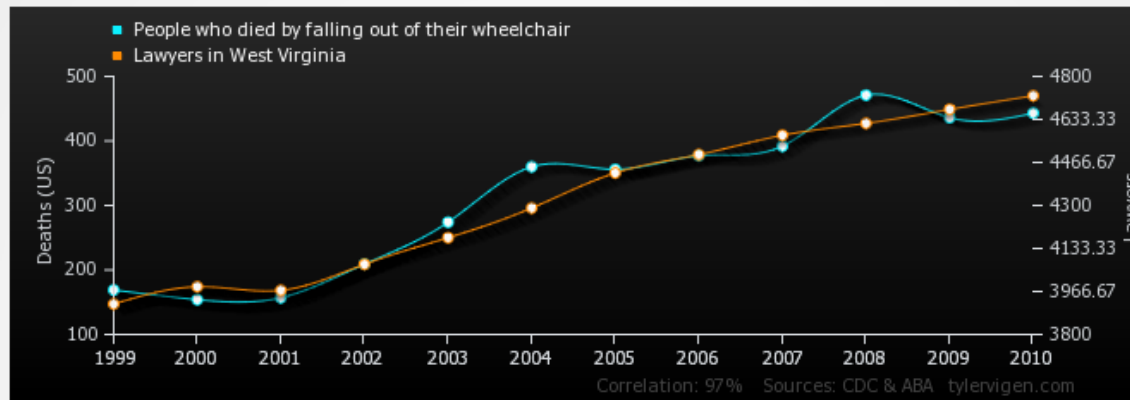




Propriété	Valeur
Moyenne des x	9.0
Variance des x	10.0
Moyenne des y	7.5
Variance des y	3.75
Corrélation entre les x et les y	0.816
Équation de la droite de régression linéaire	$y = 3 + 0,5x$
Somme des carrés des erreurs relativement à la moyenne	110,0

<http://www.tylervigen.com>

People who died by falling out of their wheelchair correlates with Lawyers in West Virginia



[Upload this image to imgur](#)

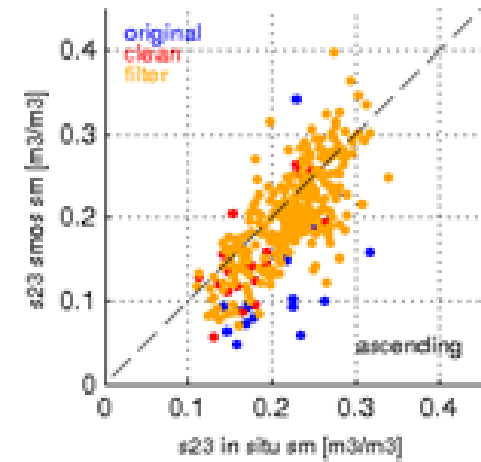
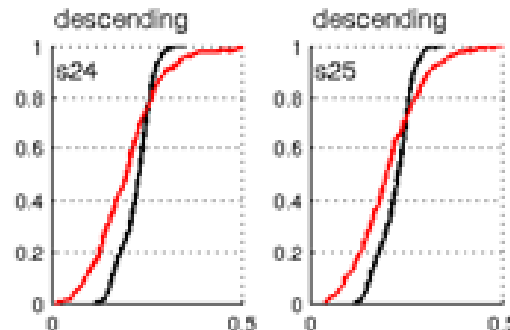
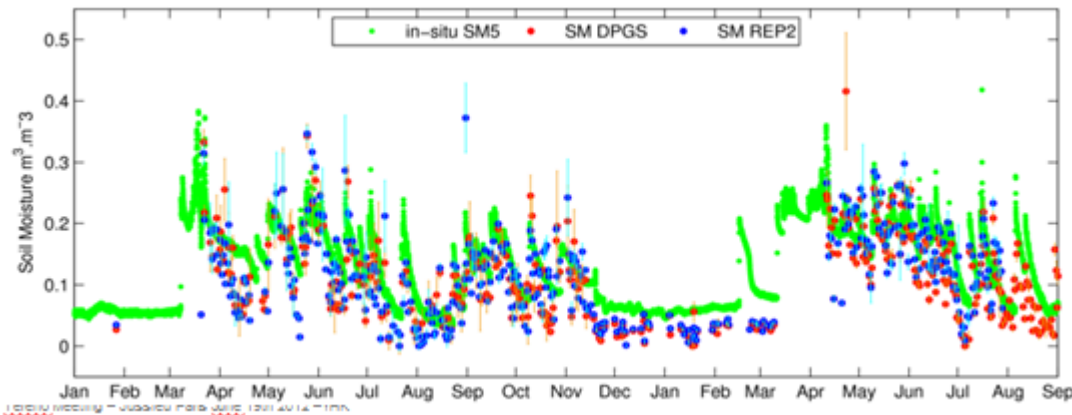
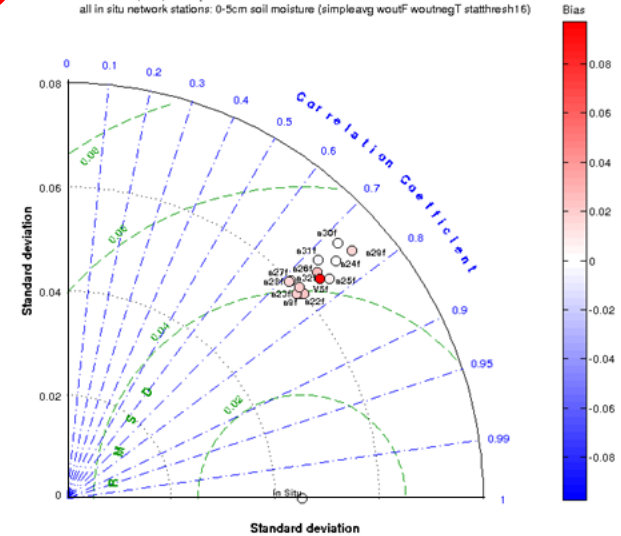
	1999	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010
<i>People who died by falling out of their wheelchair Deaths (US) (CDC)</i>	169	154	157	209	274	360	356	377	392	471	436	443
<i>Lawyers in West Virginia Lawyers (ABA)</i>	3,918	3,985	3,971	4,072	4,175	4,290	4,426	4,497	4,572	4,618	4,672	4,725

Correlation: 0.972699

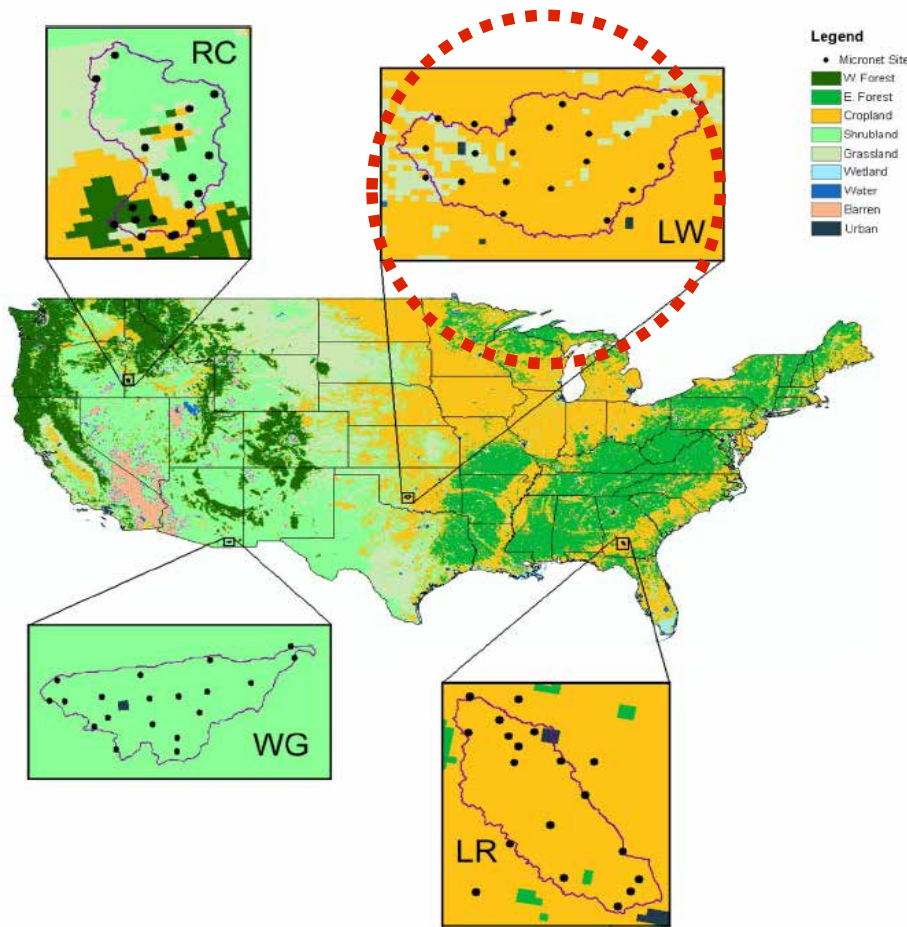
What do we need?

- q Some common sense first
- q Some reliable and characterised in situ data
- q A panoply of tools
- q Some mastering of statistics
- q And a close look

SML2PP series s9s22s23s24s25s26s27s28s29s30s31s32V5 ascending
 HOBE dgg2002029
 01-Jun-2010-19-Mar-2012
 1 (filter: sm dqx<0.070>chl2<2fl fraction<0.04
 all in situ network stations: 0-5cm soil moisture (simpleavg woutF woutnegT stathresh16)



Little Washita

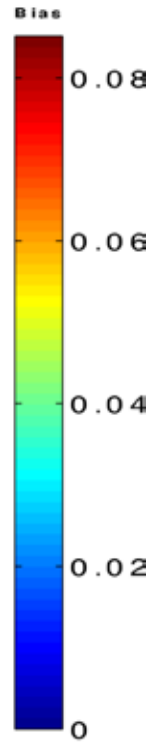
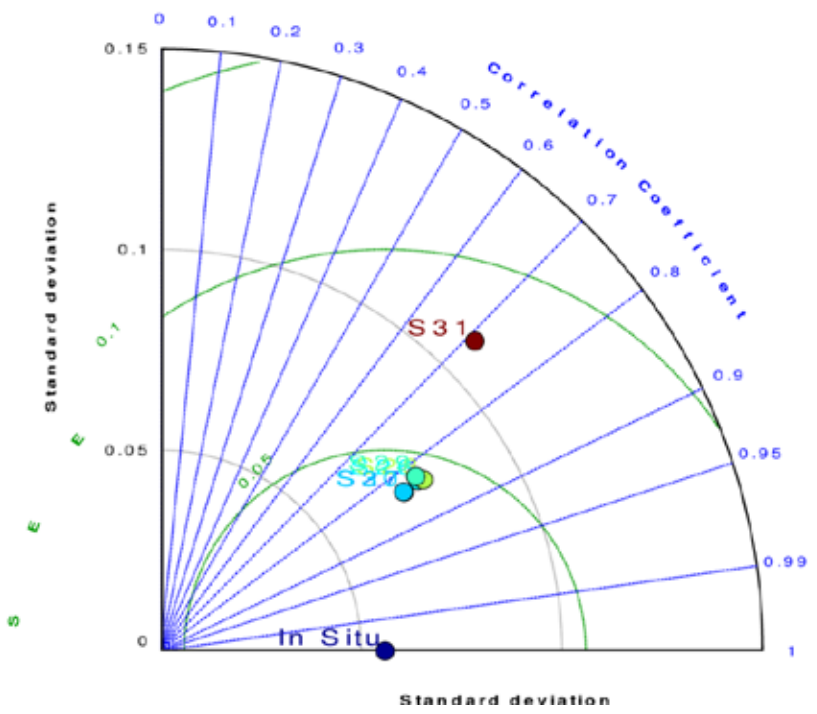


Climate : sub humid
Topography : rolling
Land use : range, wheat

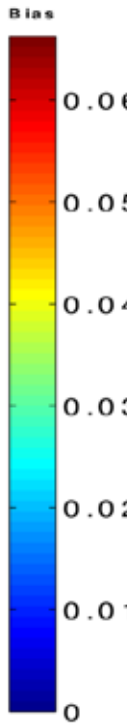
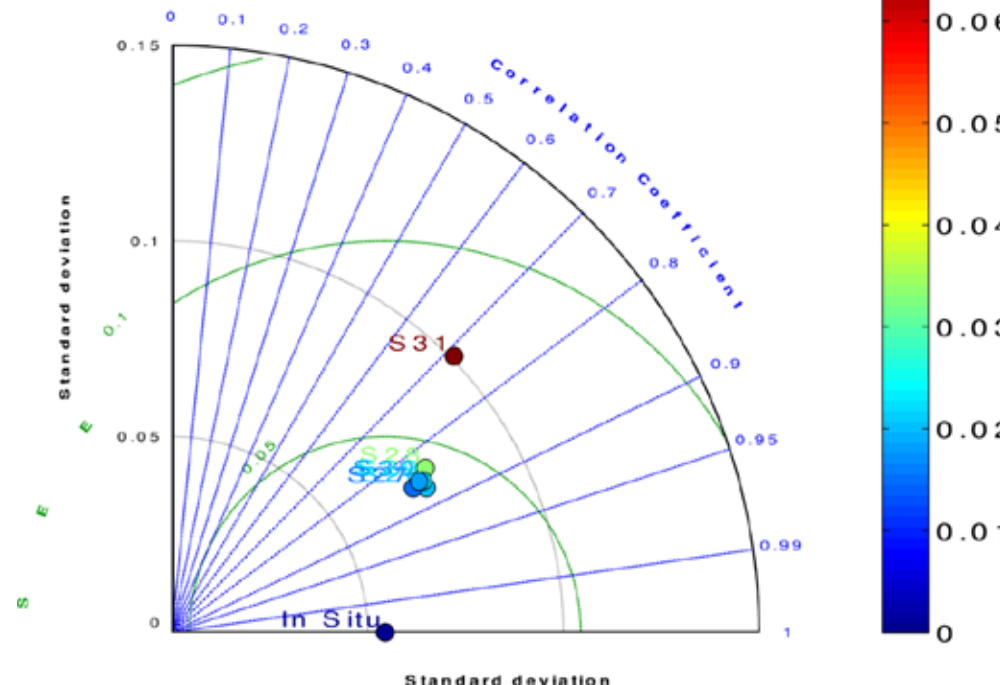


Jackson et al., Validation of AMSR soil moisture products, IEEE Transactions on Geoscience and Remote Sensing, vol. 48, 2010.

Little Washita

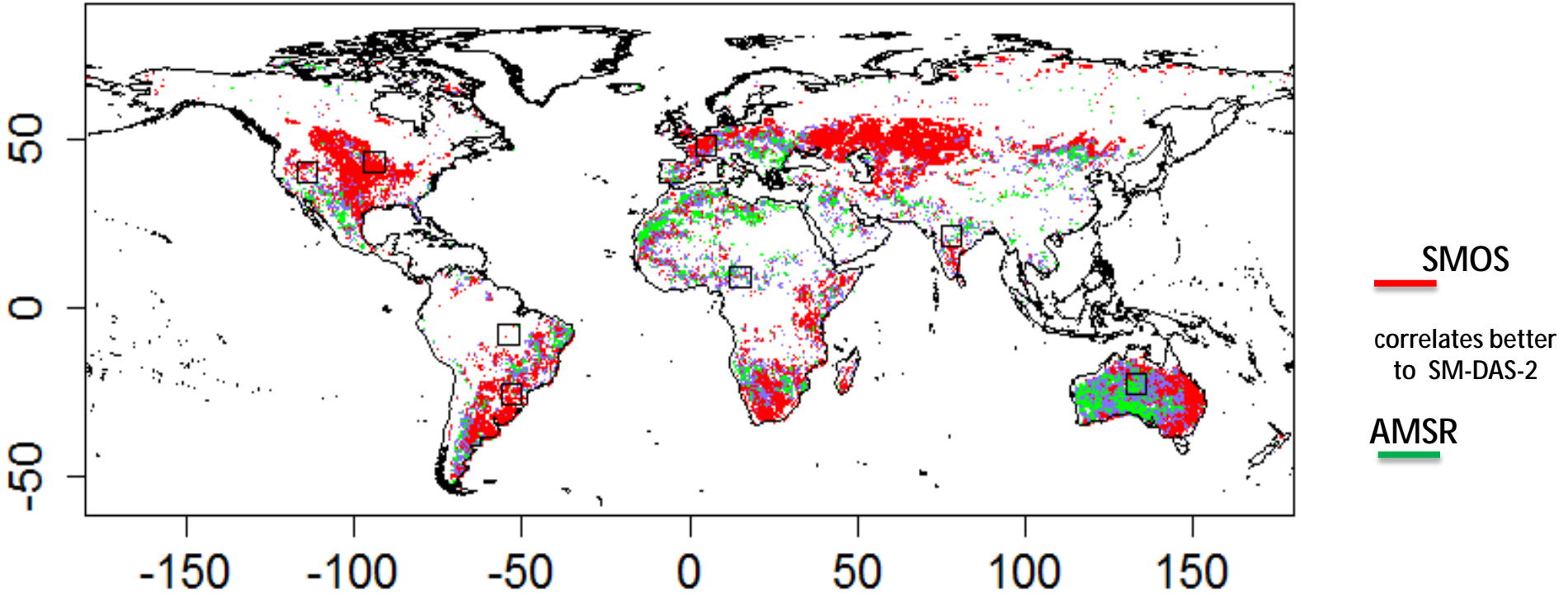


Little Washita -- DES



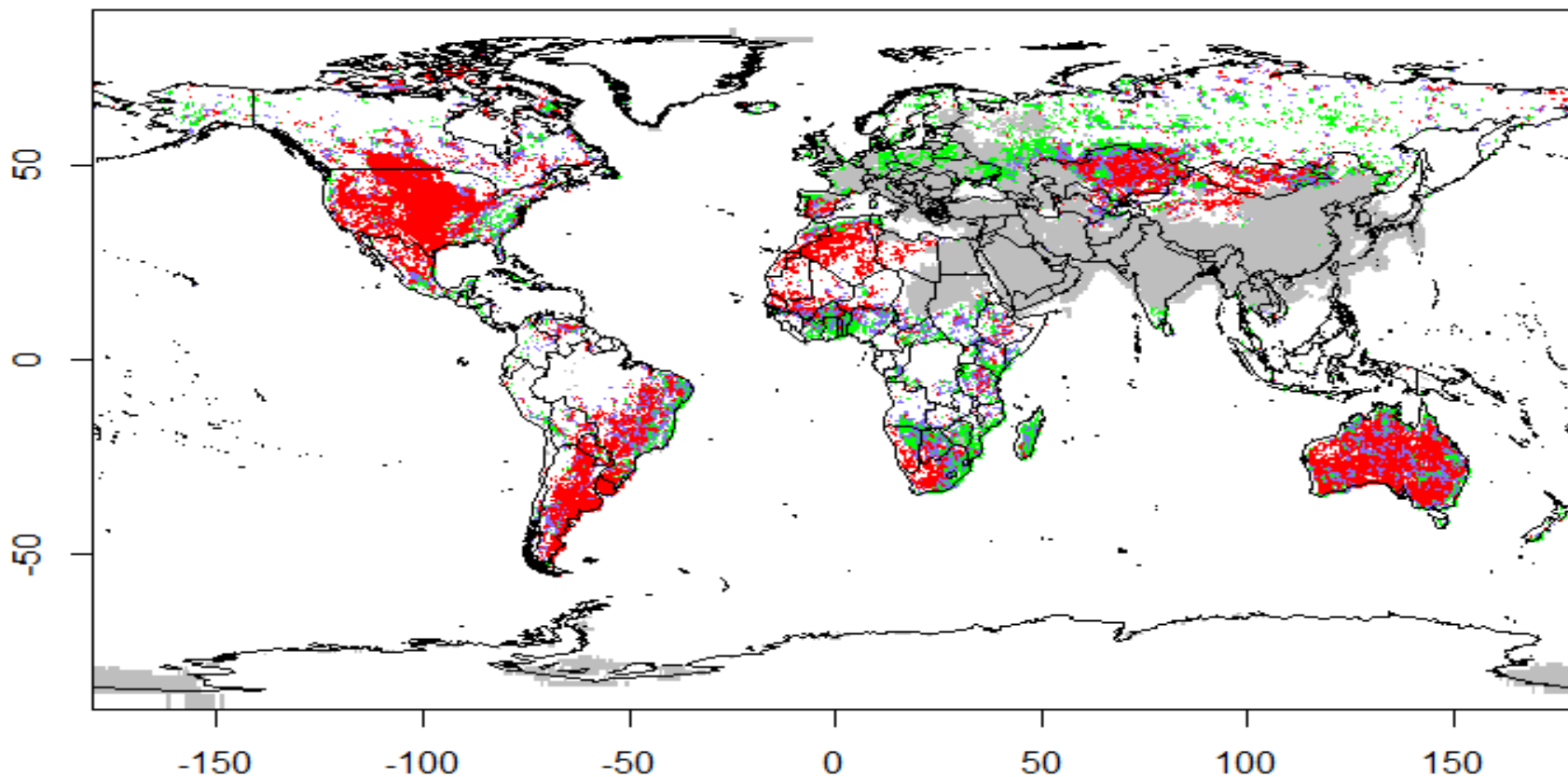
**Blind test V620
Leroux Bindlish**

Global Comparison between SMOS-L3 and AMSR surface soil moisture with SSM calculated by SM-DAS-2 .



R coeff.
SM Anomalies (34 day window)

Correlation of SMOSL3 and ASCAT SM vs MERRA/land SM product (P value < 0.05) 2010–2012 period (Anomalies)



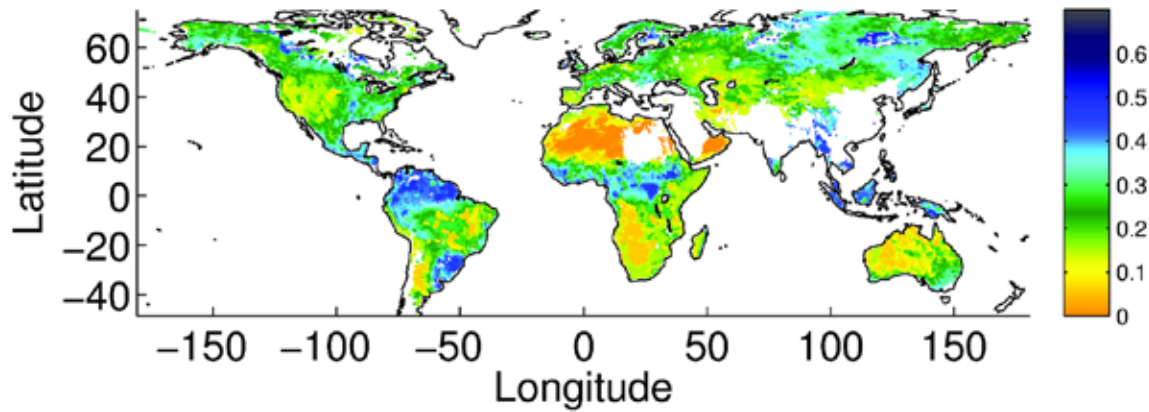
ASCAT: Global R=0.22
SMOS: Global R=0.29

A. Alyaari

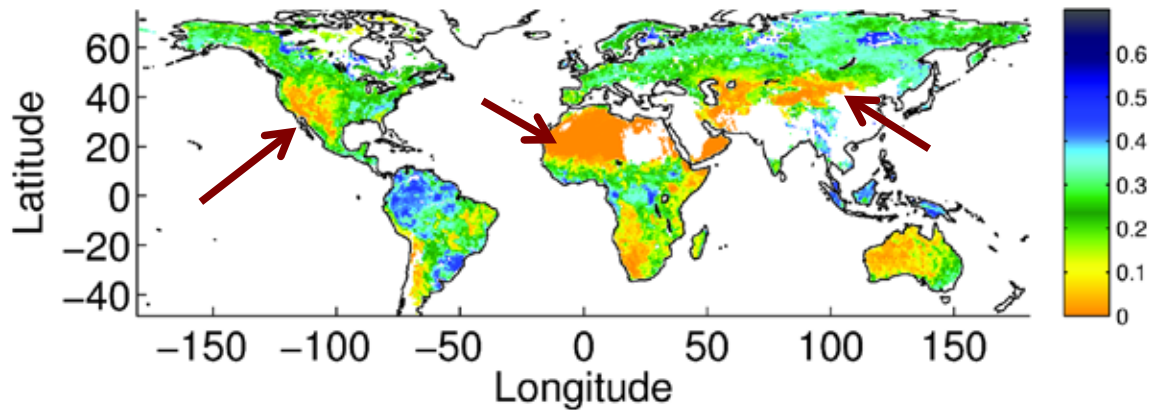
NN retrievals

Juillet 2010

SM ECMWF



SM NN



- Somewhat drier than ECMWF

Evidence of positive bias of ECMWF

- Muñoz-Sabater et al. (in prep)

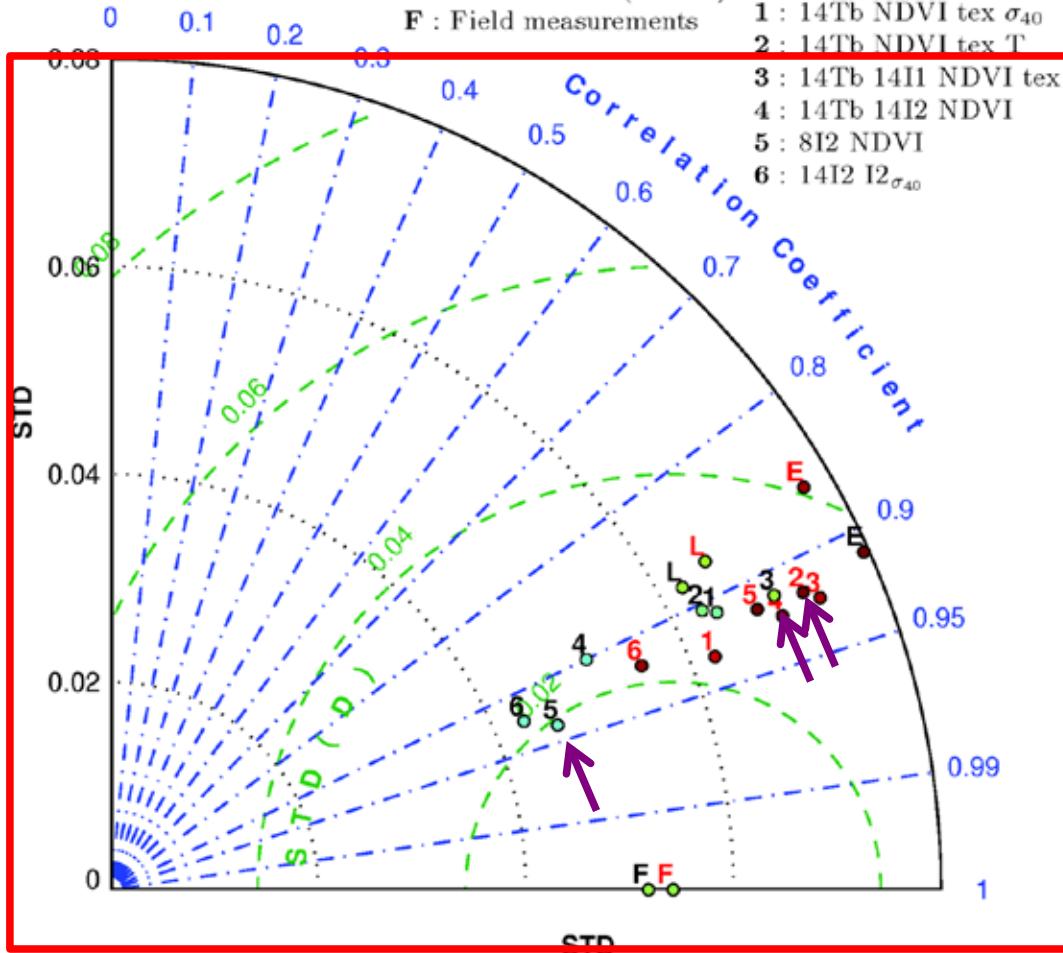
- Albergel et al. 2012

Little Washita

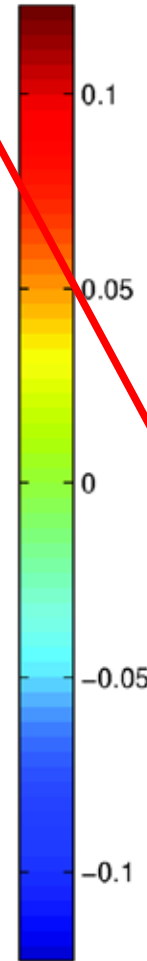
LW-2: Training with ECMWF (red) and L3SM (black)

L : SMOS L3SM
 E : ECMWF SM (0-7 cm)
 F : Field measurements

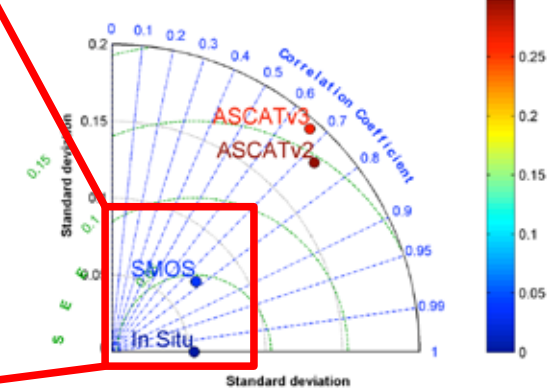
NN model
 1 : 14Tb NDVI tex σ_{40}
 2 : 14Tb NDVI tex T
 3 : 14Tb 14I1 NDVI tex
 4 : 14Tb 14I2 NDVI
 5 : 8I2 NDVI
 6 : 14I2 I2 σ_{40}



Bias



Little Washita - morning



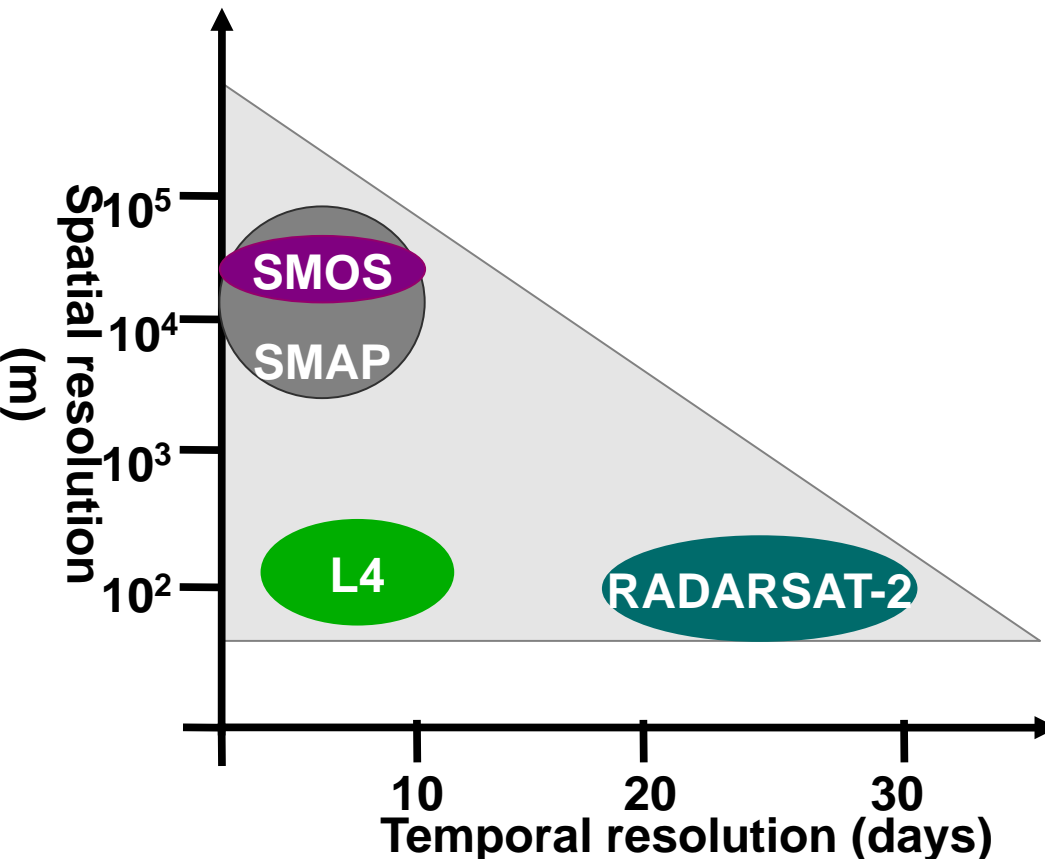
N Rodriguez

Different disaggregation schemes

- q SMAP
 - ✓ à N Das et al approach
 - ✓ yesterday's presentation by Dara
 - ✓ Yesterday's presentation by Jeff
- q Using Thermal and Infra red
 - ✓ Yesterday's presentation by Maria-Jo
- q With sparse active microwaves
 - ✓ Radar every so often
 - ✓ Sat Kumar et al approach

Active Passive disaggregation (S Tomer)

- ▼ L4: Combined high resolution active and passive Microwave soil moisture product



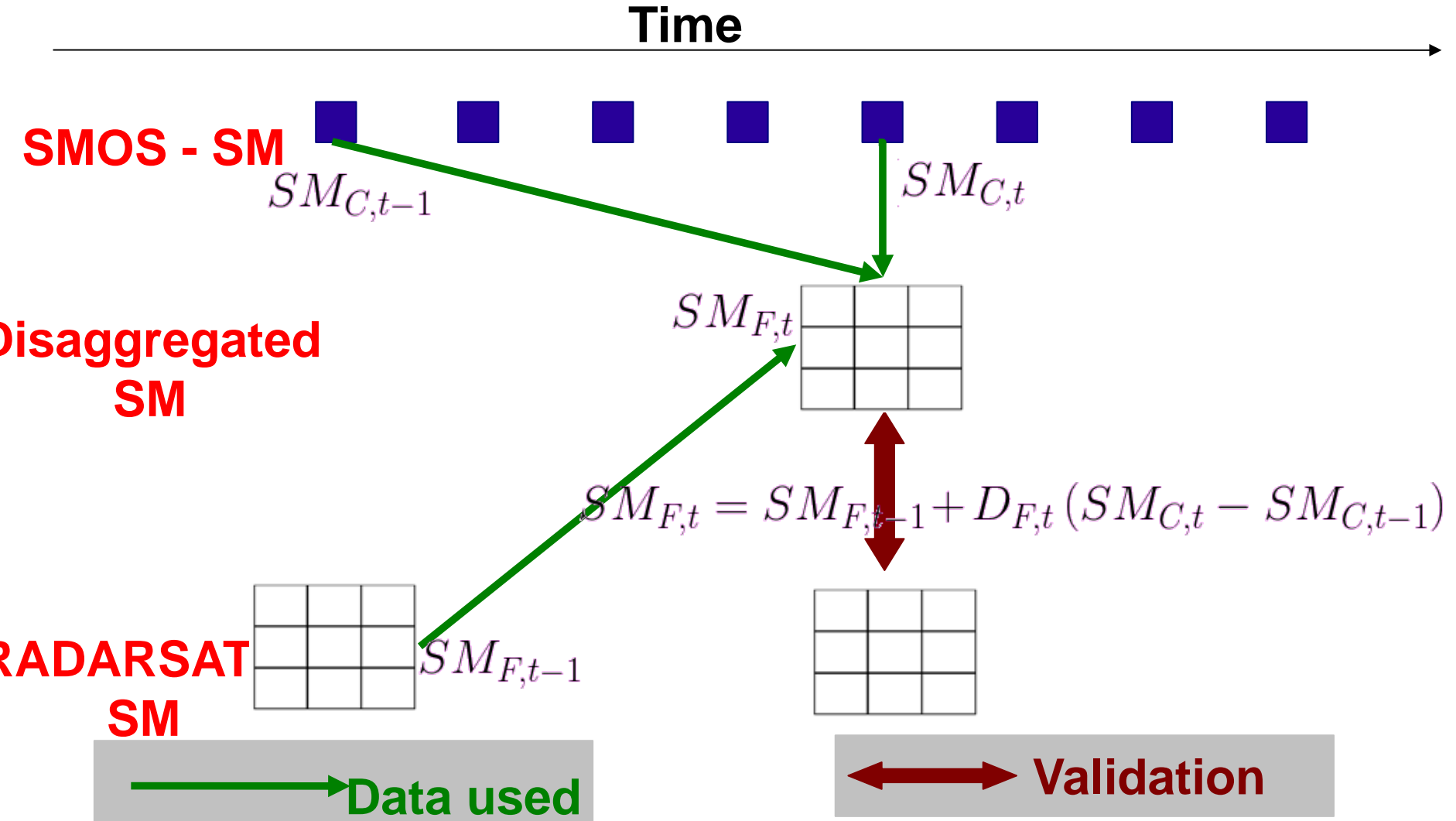
Passive (SMOS)

Spatial res. --> ~25 km
 Temporal res. --> ~3 days

Active (RADARSAT-2)

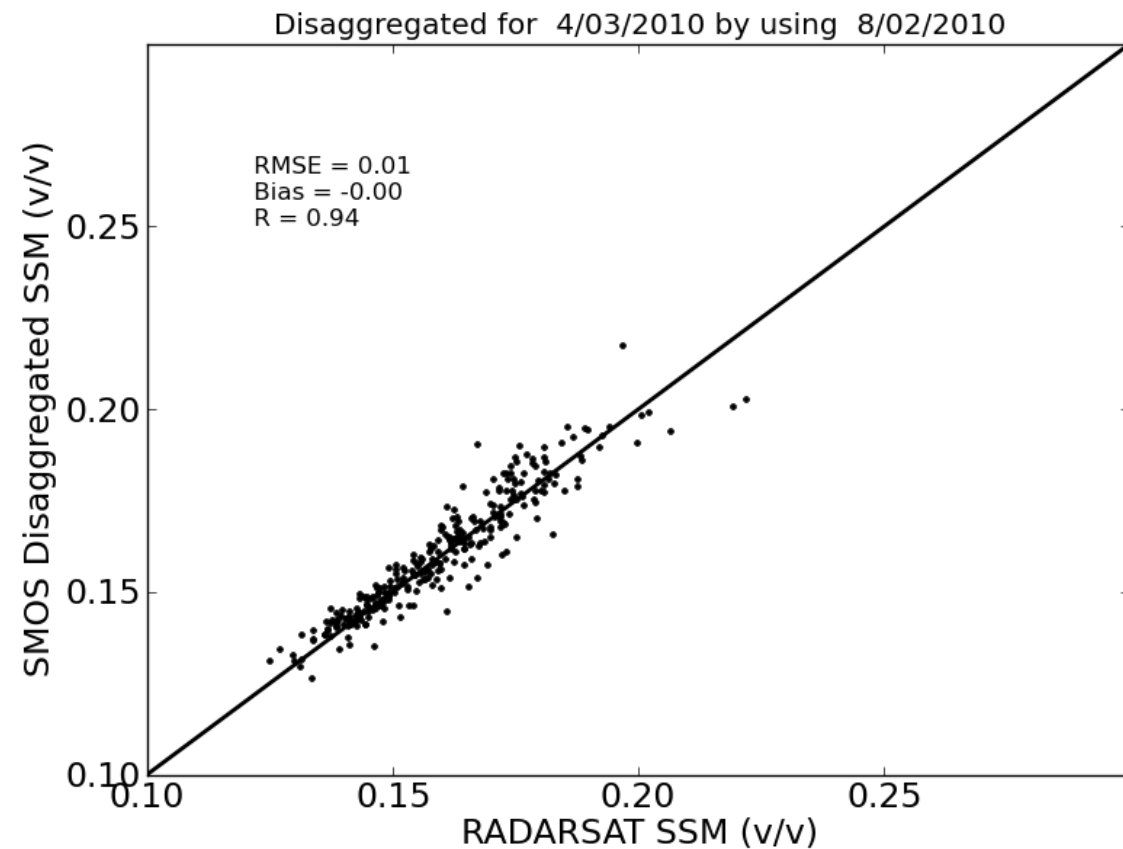
Spatial res. --> ~100m
 Temporal res. --> ~24 days

Spatio-temporal disaggregation

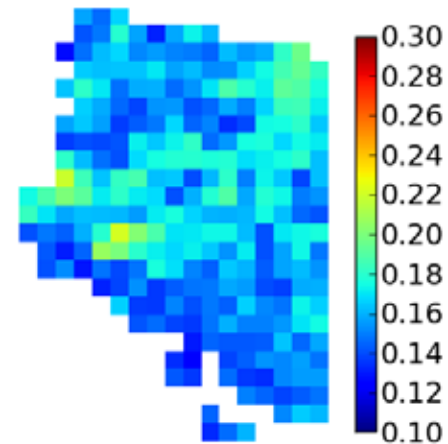




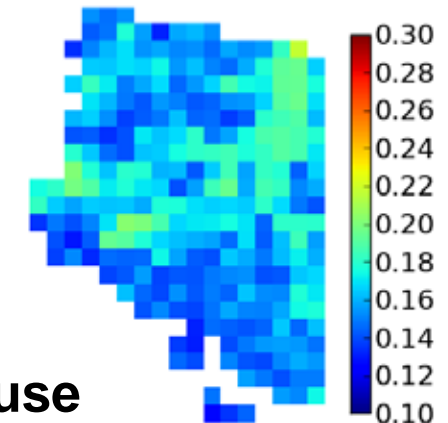
Validation of downscaled SMOS soil moisture with respect to RADARSAT-2 soil moisture



RADARSAT-2 retrieved



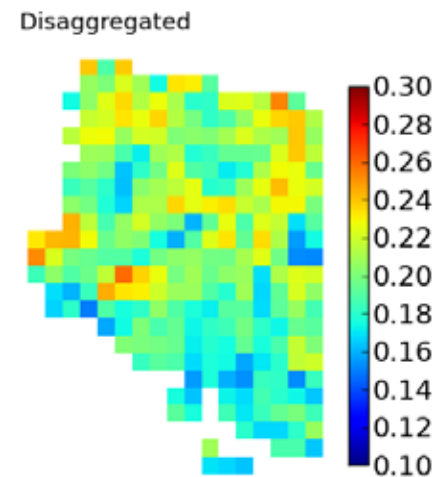
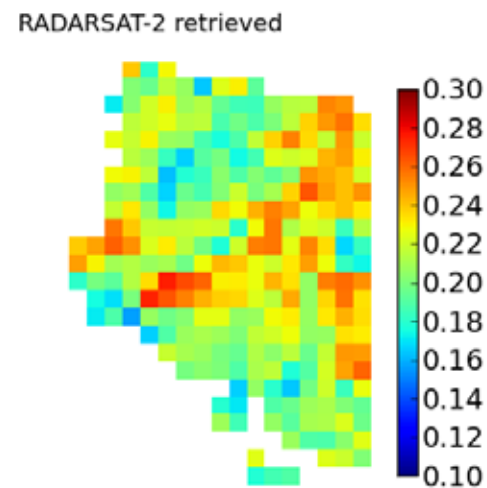
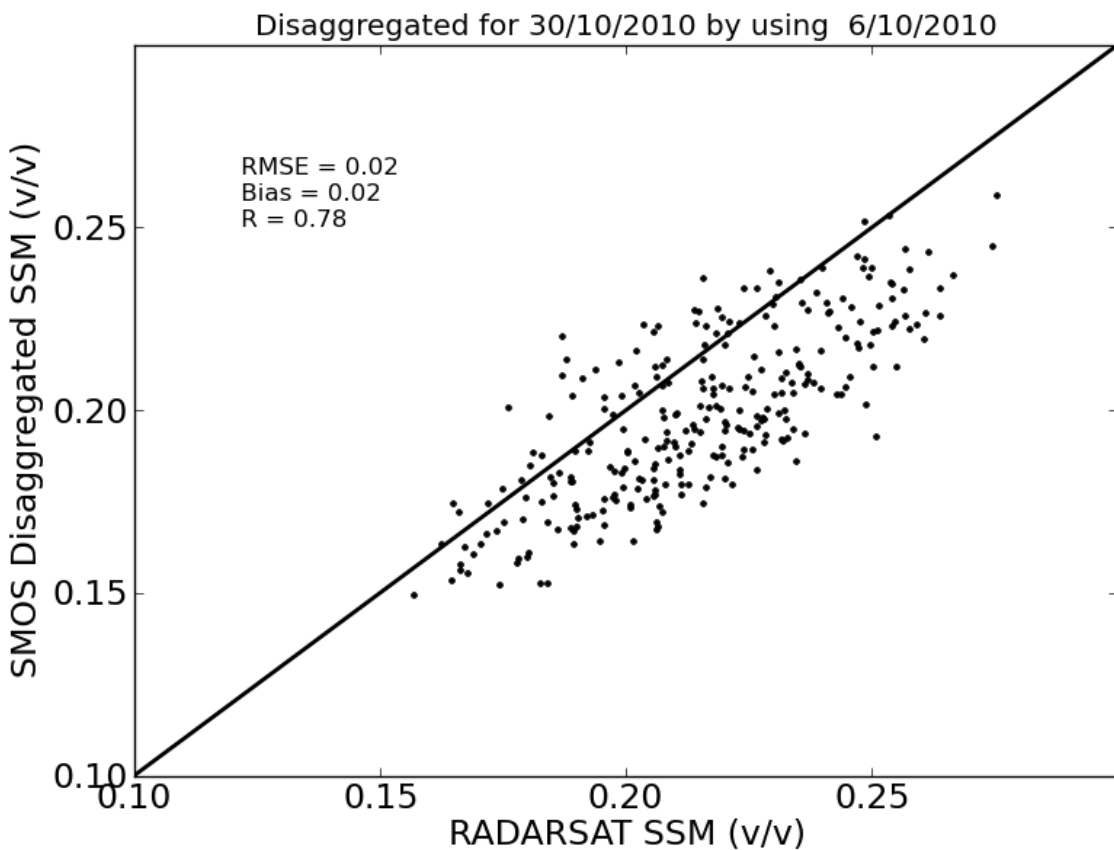
Disaggregated



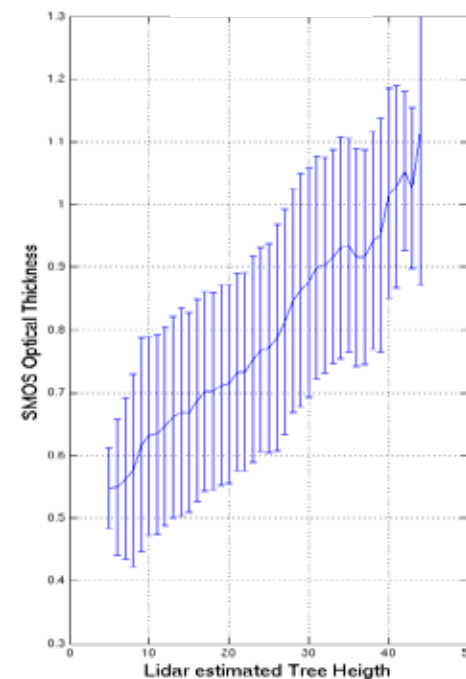
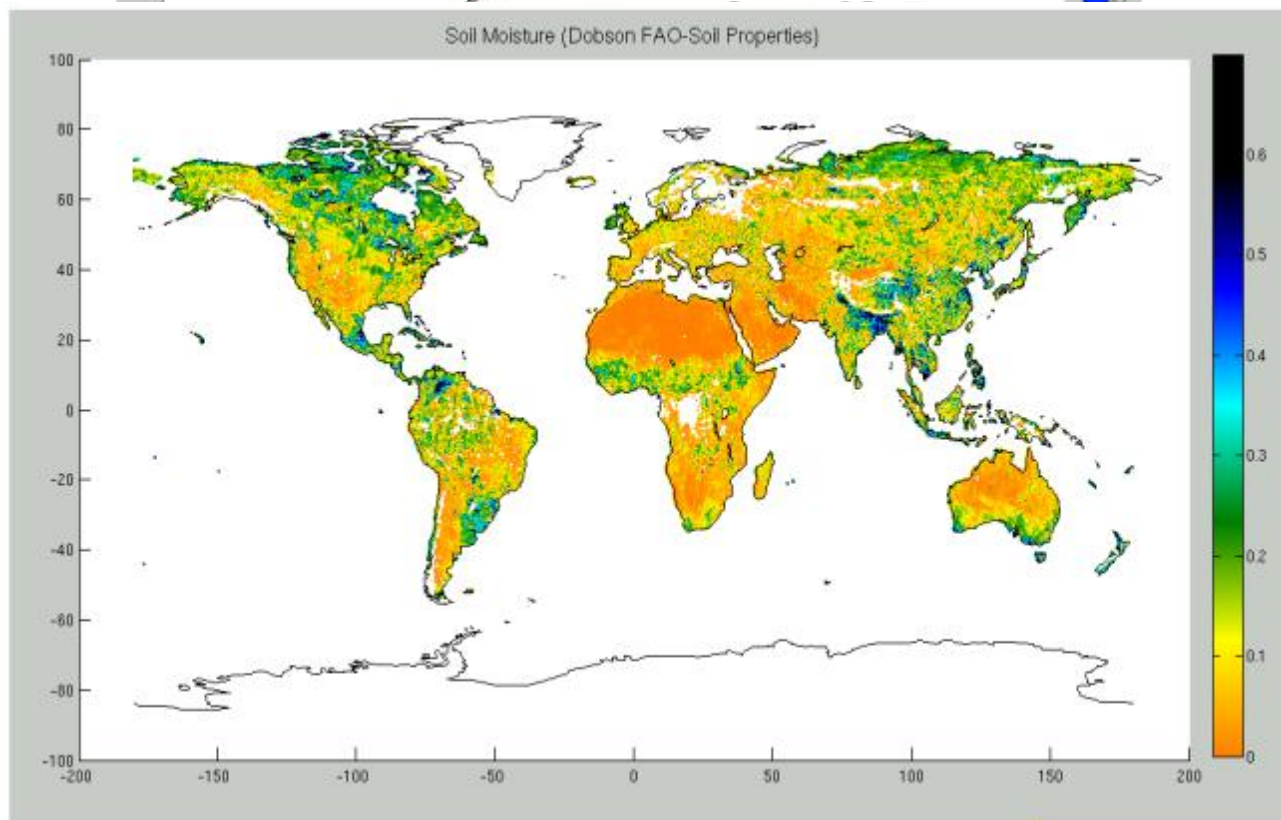
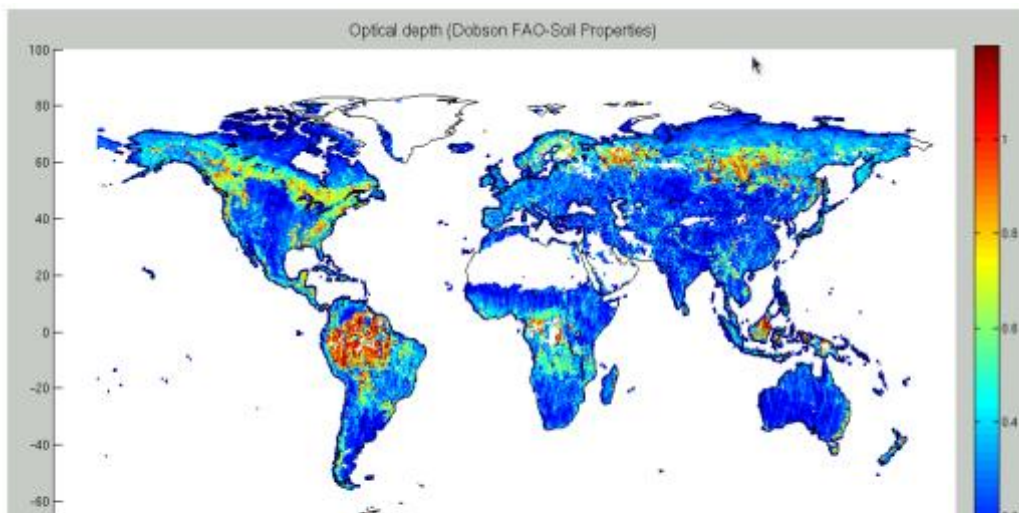
⊗ Disaggregation is not performed for the forest land use



Validation of downscaled SMOS soil moisture with respect to RADARSAT-2 soil moisture



Global Map of retrieved optical thickness



*For reference:
Forest height estimated by GLAS-ICESat Lidar (Simard et al., 2011)*

Ferrazzoli
Rahmoune



SMOS+Hydro Project

Sat Kumar Tomer, Ahmad Al Bitar, Olivier Merlin, Yann Kerr
Centre d'Etudes Spatiales de la Biosphere, 31401 Toulouse, France

Hans Lievens, Niko Verhoest
Laboratory of Hydrology and Water Management, Ghent University, 9000 Ghent, Belgium

Valentijn Pauwels, Jeffrey Walker, Gift Dumedah
Department of Civil Engineering, Monash University, 3800 Victoria, Australia

Eric Wood, Ming Pan, Alok Sahoo
Land Surface Hydrology Group, Princeton University, 08544 Princeton, USA

Gabrielle De Lannoy, Rolf Reichle
Global Modeling & Assimilation Office, NASA Goddard Space Flight Center, 20770 Greenbelt, USA

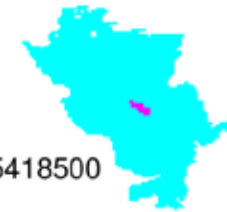
Matthias Drusch
European Space Agency, 2200 AG Noordwijk, The Netherlands



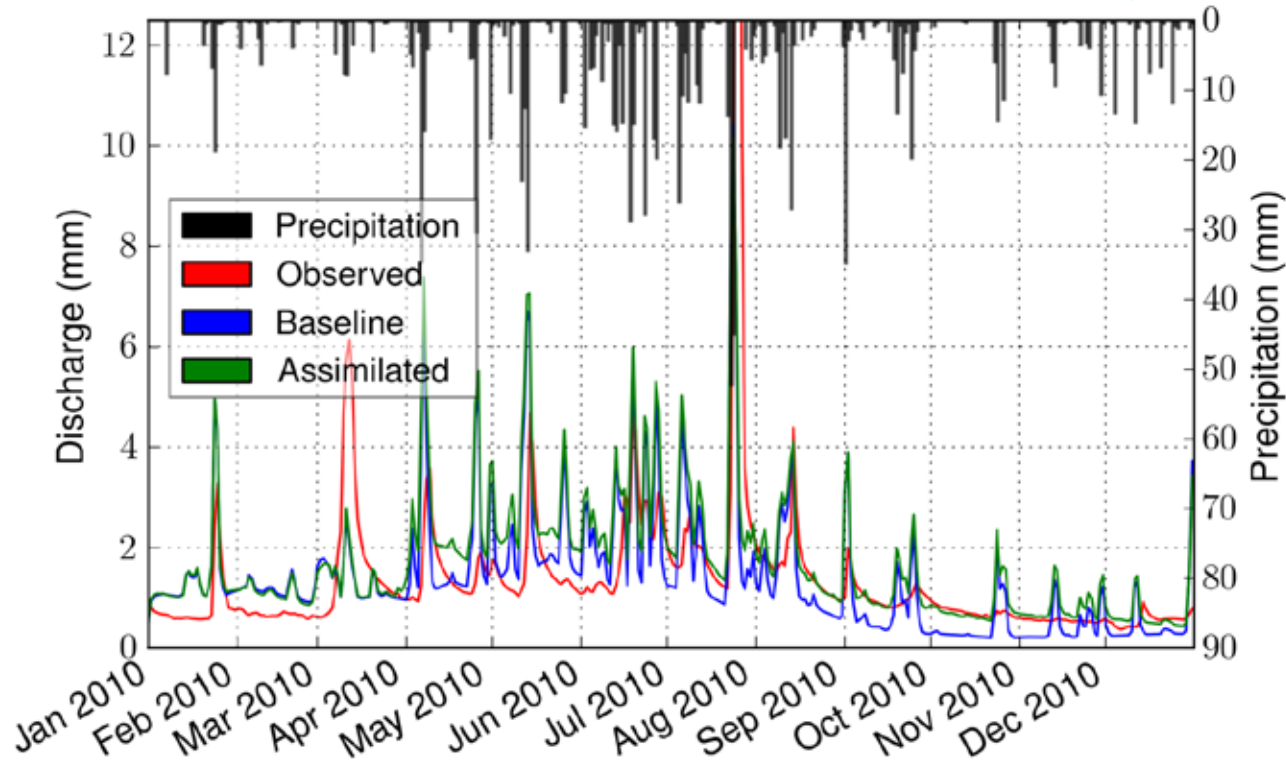
Objectives

- n **Assimilation of SMOS data into VIC model to improve the flood prediction**
 - **Soil moisture**
 - **Coarse scale**
 - **Fine scale**
 - **Temperature brightness**
- n **Streamflow routing**
- n **Test the model setup at two different basins UMB (humid) and MDB (arid)**

	Baseline	Assimilation
KGE	0.41	0.49
Ratio of std	1.45	1.37
Bias	-0.17	0.20
Correlation	0.66	0.73



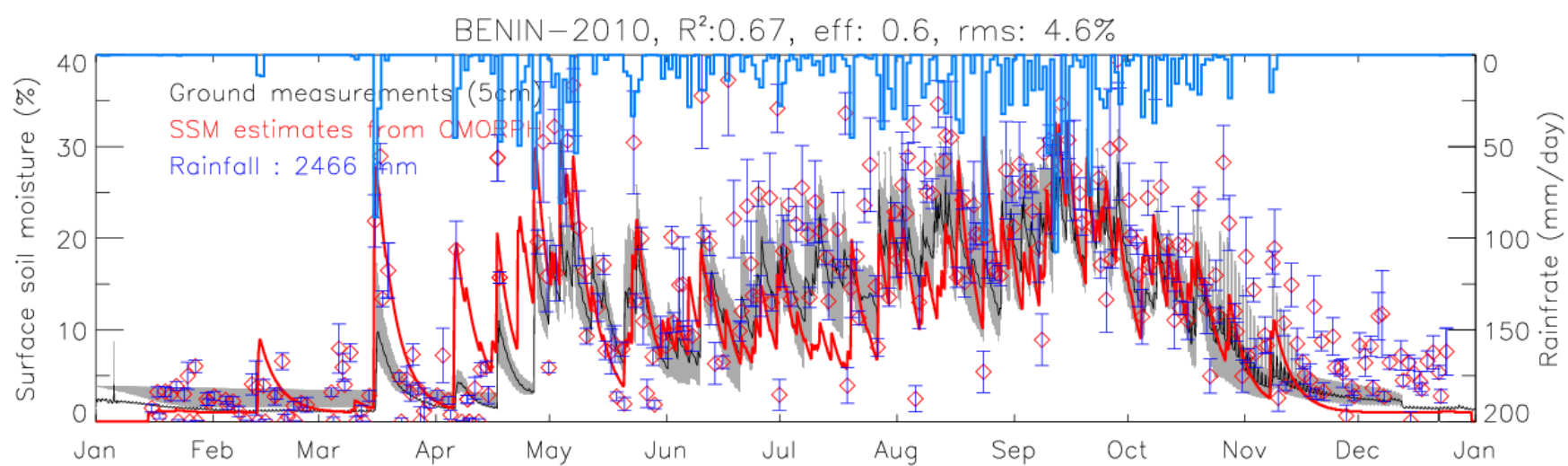
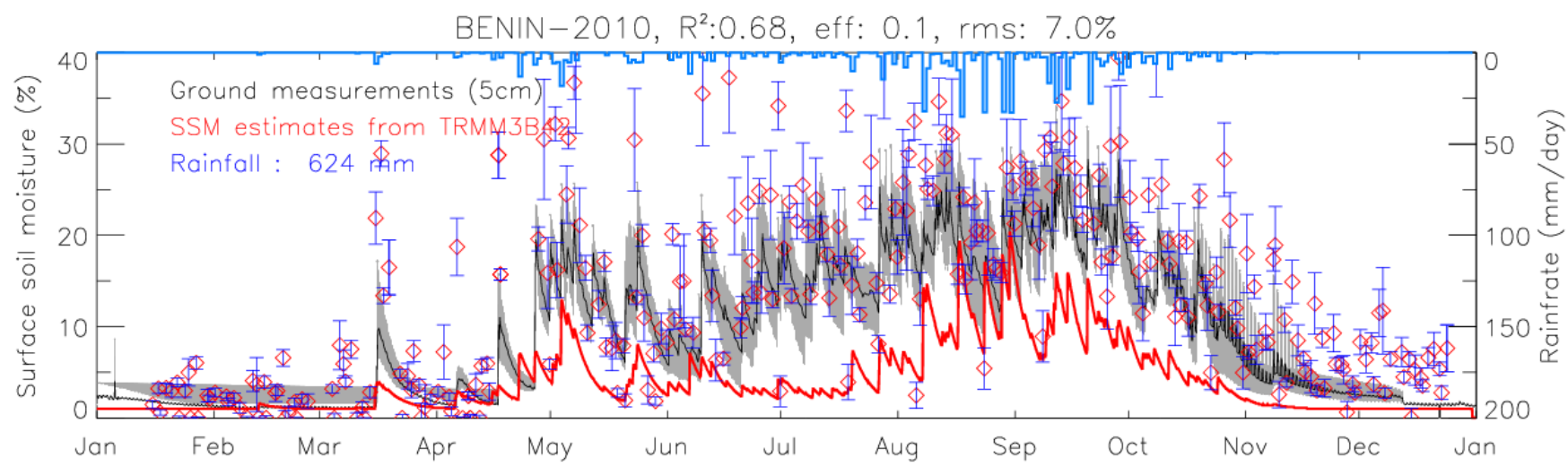
Gauge Site No. = 5418500



∅ Soil moisture assimilation is enhancing the streamflow except the peak flow

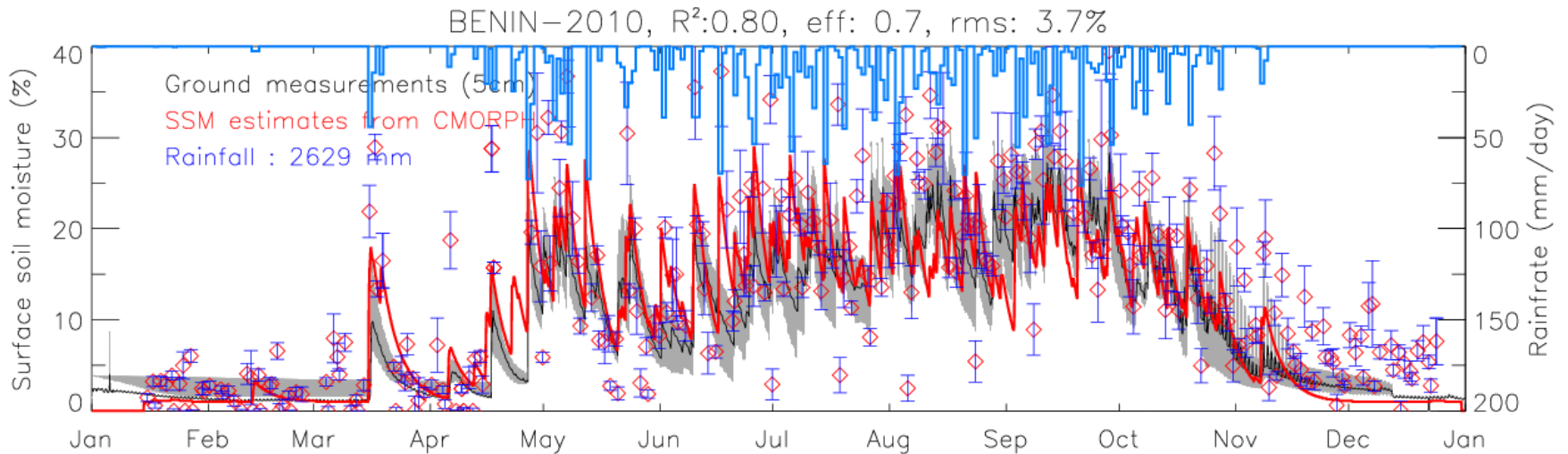
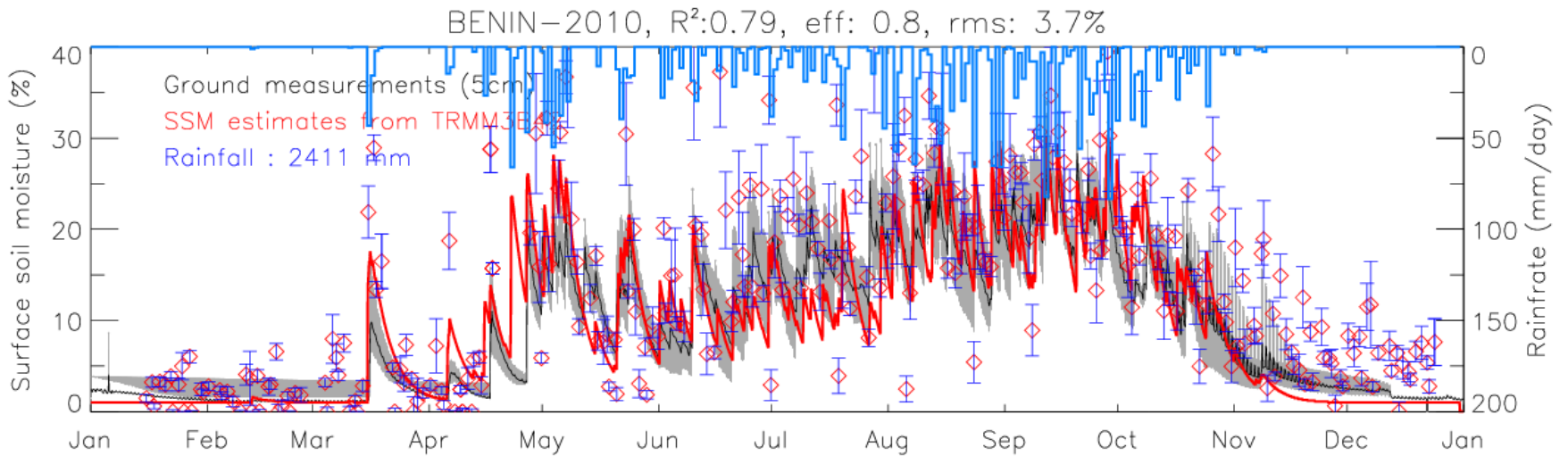


Estimated SSM without SMOS assimilation (Benin site) Using TRMM-3B42 (top) and CMORPH rainfall products (bottom)

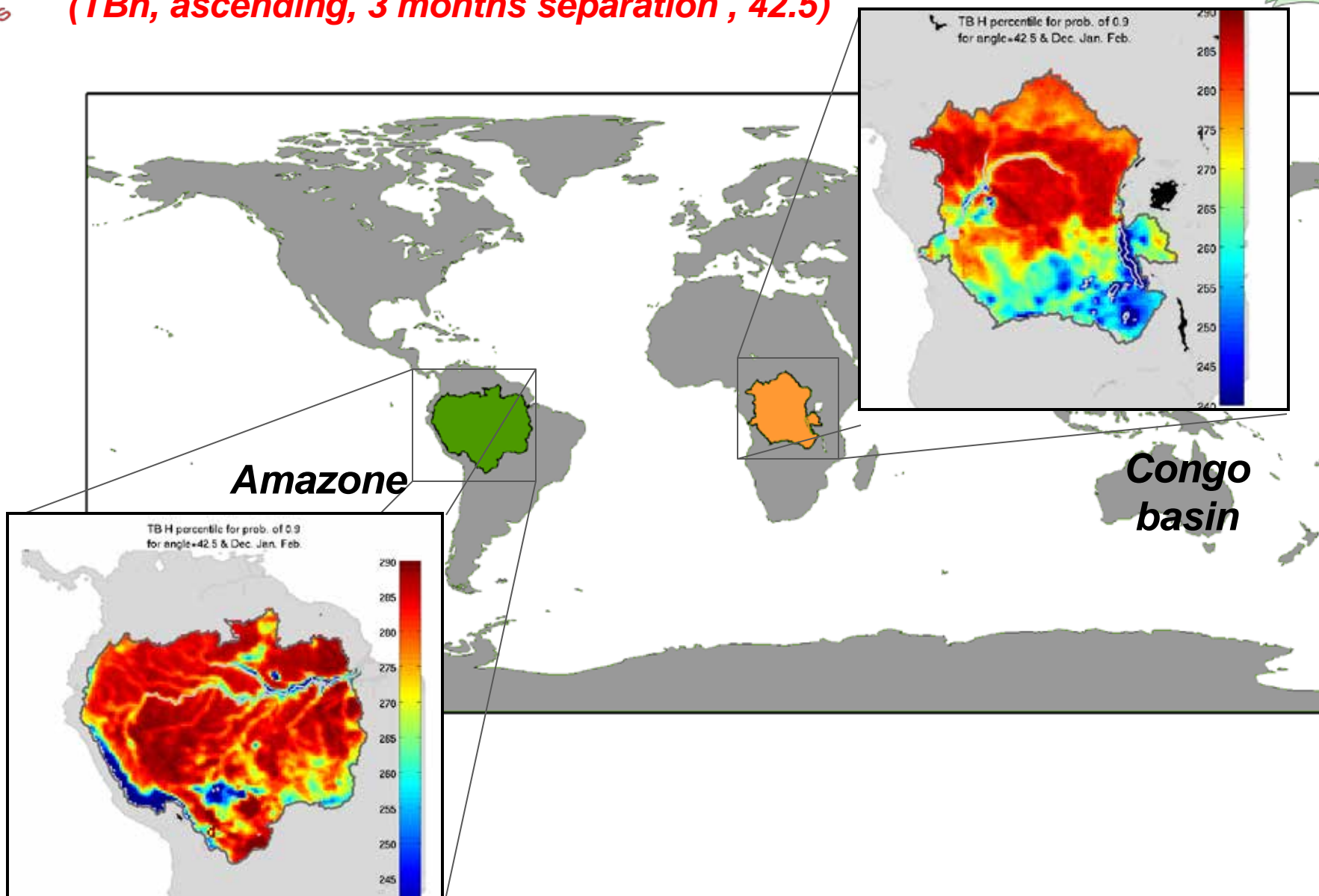




Estimated SSM with SMOS assimilation (Benin site) Using TRMM-3B42 (top) and CMORPH rainfall products (bottom)



Seasonal dynamics of brightness temperatures (TBh, ascending, 3 months separation, 42.5)



Amazon

Congo basin

Root zone soil moisture

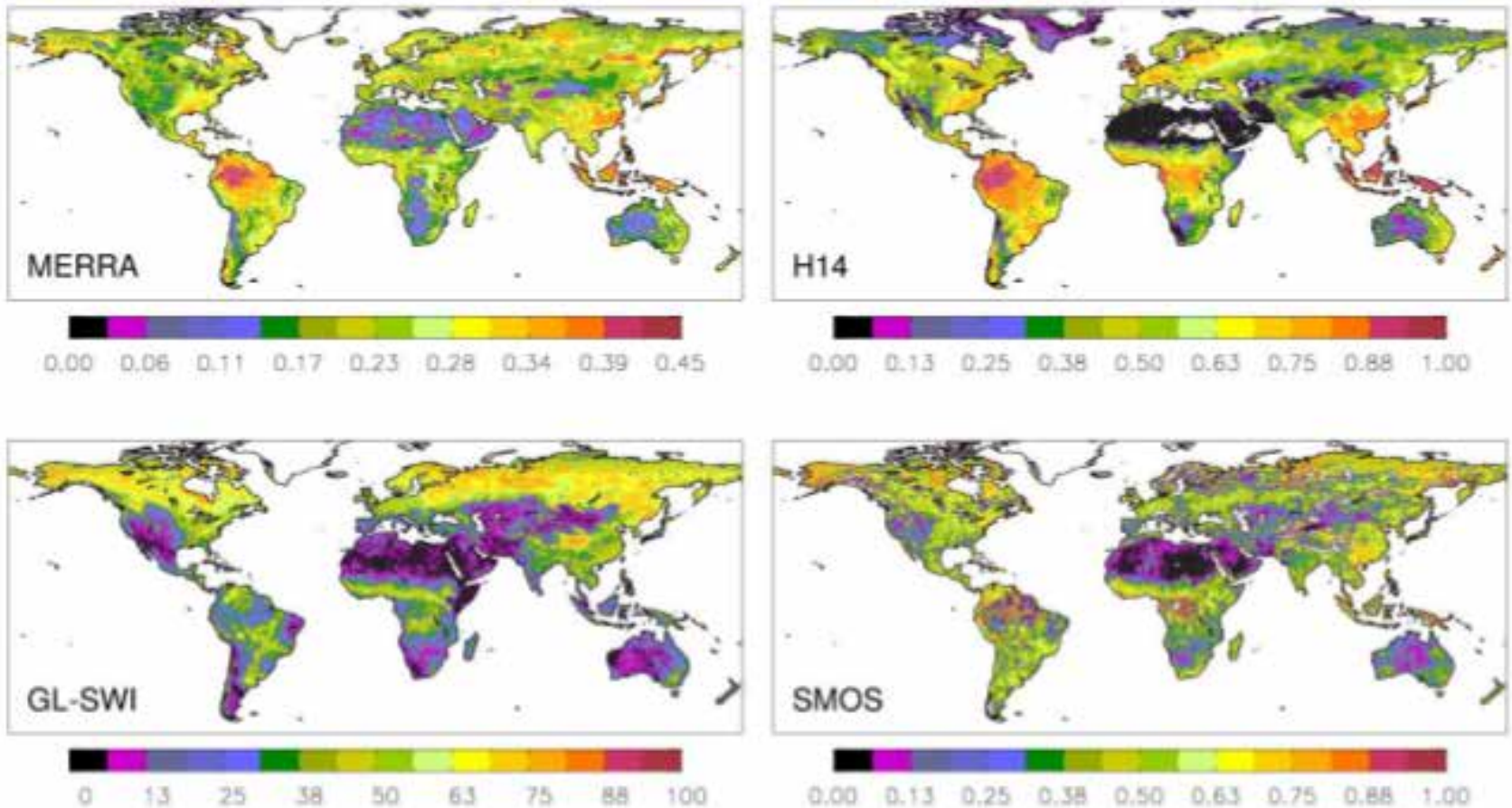
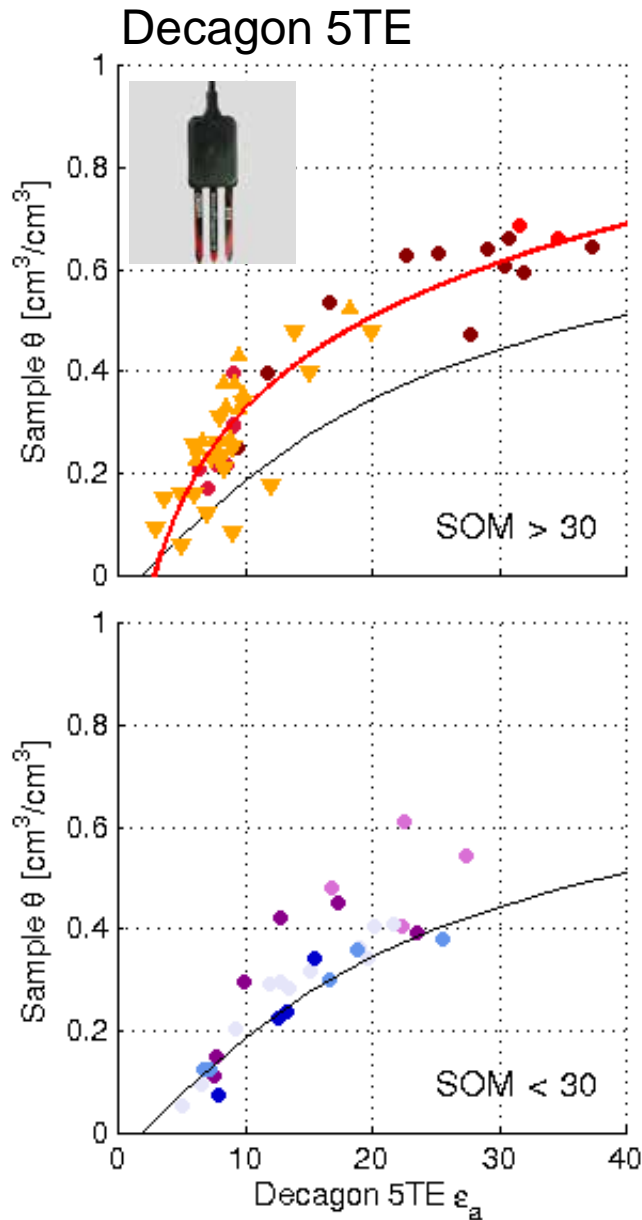


Figure 1: Annual mean root-zone soil moisture maps for MERRA, H14, GL-SWI and SMOS.

Soil moisture sensor calibration



Organic (SOM >30%)

- ▲ HOBE Forest 1 Lab
- ▼ HOBE Forest 1 Field
- HOBE Heath
- FMI Forest
- HOBE Forest 2

- Default mineral
- Fit organic

Mineral (SOM <30%)

- FMI Forest
- HOBE Heath
- HOBE Forest
- FMI Heath 1
- FMI Heath 2

↑ Increasing SOM

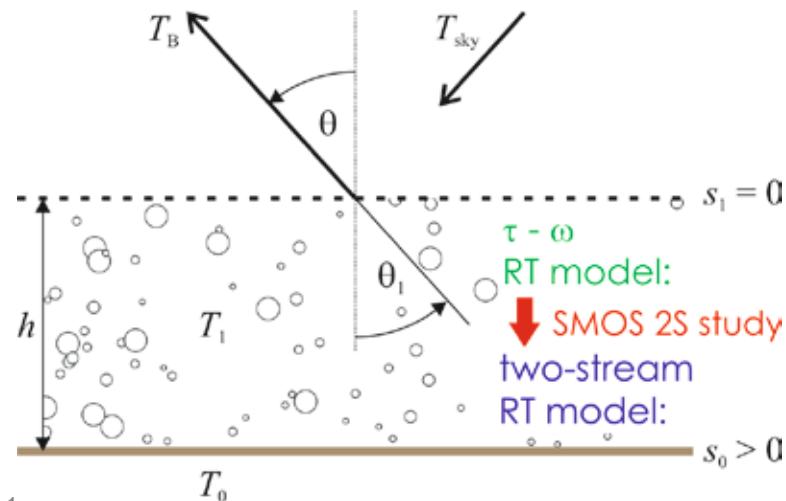
↑ Increasing SOM

→ Calibration curves fitted through organic data, to be used for re-calibration of in situ soil moisture of automatic network stations

→ Mineral calibration curve as $f(\text{SOM})$ possible if further data acquired...

GAMMA REMOTE SENSING

Worbstr. 225, CH-3073 Gümligen, Switzerland
<http://www.gamma-rs.ch/>

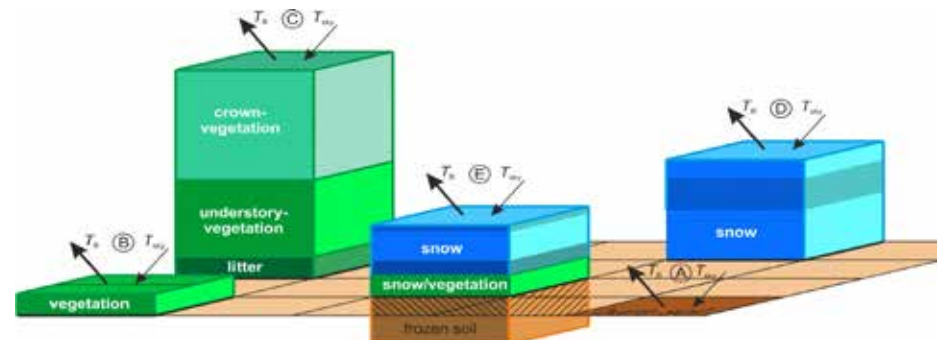
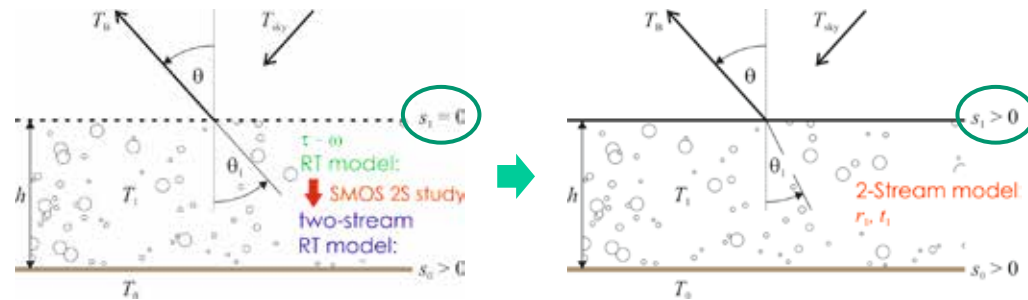


Major Project Goals:

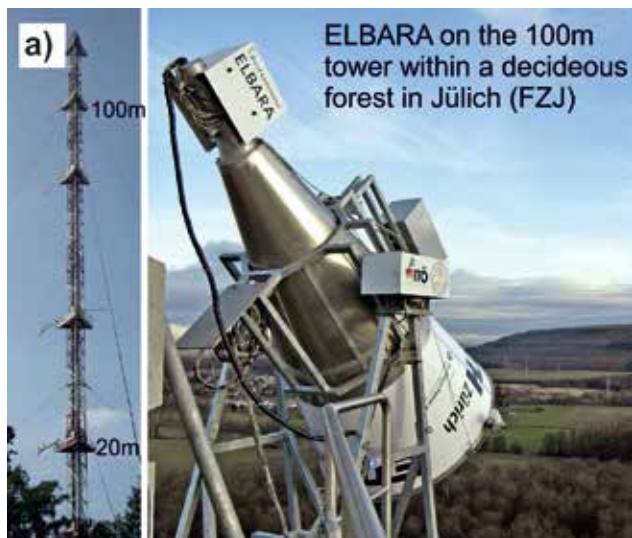
- Exploring the potential of replacing the $t - w$ (TO) RT model used in the SMOS L2 SM processor with the Two-Stream (2S) Radiative Transfer (RT) model.
- SMOS 2S study shall allow taking the decision on a possible implementation of the proposed retrieval update in the operational SMOS SM retrieval.

Pros of 2S RT over TO RT model:

- 1) 2S RT considers multiple reflections (higher order solution of RT equations).
- 2) "soft-layer" ($s_1=0, q=q_1$) assumption can be given up. "hard-layer" ($s_1>0, q>q_1$) necessary for e.g. soil beneath snow / litter.
- 3) 2S RT allows to represent multiple layered systems.
- 4) All this is relevant to advance full exploitation of SMOS TB (novel data products).

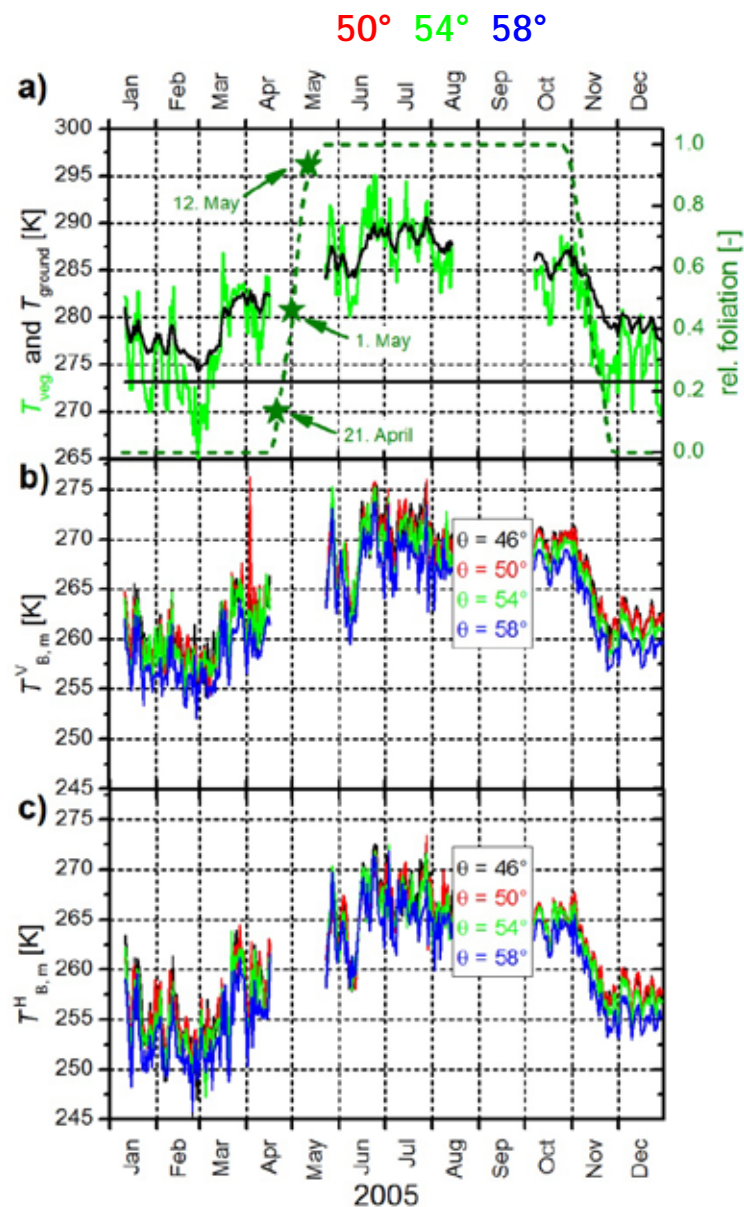


5) Comparative 3-P Retrieval Based on FOSMEX $T_B^P(\theta)$



forest development 2005

3-P retrievals:



Comparative 3-P Retrieval Based on FOSMEX $T_B^P(q)$

- Footprints observed are:
 - between 2000 m² and 6000 m² for the $q_k = 46^\circ, 50^\circ, 54^\circ, 58^\circ$ considered.
 - Deciduous forest comprising oak, birch, and beech:
 - tree age: 40 - 80 years
 - average crown height 24 m.
 - column density (dry) » 15 kg m⁻².
 - max. fresh leaf density » 1.14 kg m⁻²
- Configurations applied to the **original retrieval** and the **updated retrieval scheme**

retrieved parameters P_i :				constant parameters P_{const} :	
P_i	P_{i}^{guess}	P_i^S	$P_{constrain}$	P_{const}	value
WC	0.2	0.2	0.0 £ WC £ 1.0	$tt_H = r_{ttHV}$	1.0
t^*	0.8	0.5	0.0 £ t^* £ 2.5	DW^*	0.0
W^*	0.2	0.2	0.0 £ W^* £ 1.0	$T_{veg.}$ & T_{ground}	data shown
				$X_A = X_B$	0.3
				clay	0.160
				Q_H	0.0
				$N_{RV} = N_{RH}$	0.0
				H_R	1.2

$$CF = \frac{\overset{\text{measured}}{\uparrow} \overset{\text{simulated}}{\uparrow} \sum_{p=H,V;q_k} \frac{(T_{B,k,m}^P - T_{B,k,s}^P)^2}{(T_B^S)^2}}{\sum_{\#para.i} \frac{\sum_{p=H,V;q_k} (P_i^{guess} - P_i)^2}{(P_i^S)^2}}$$



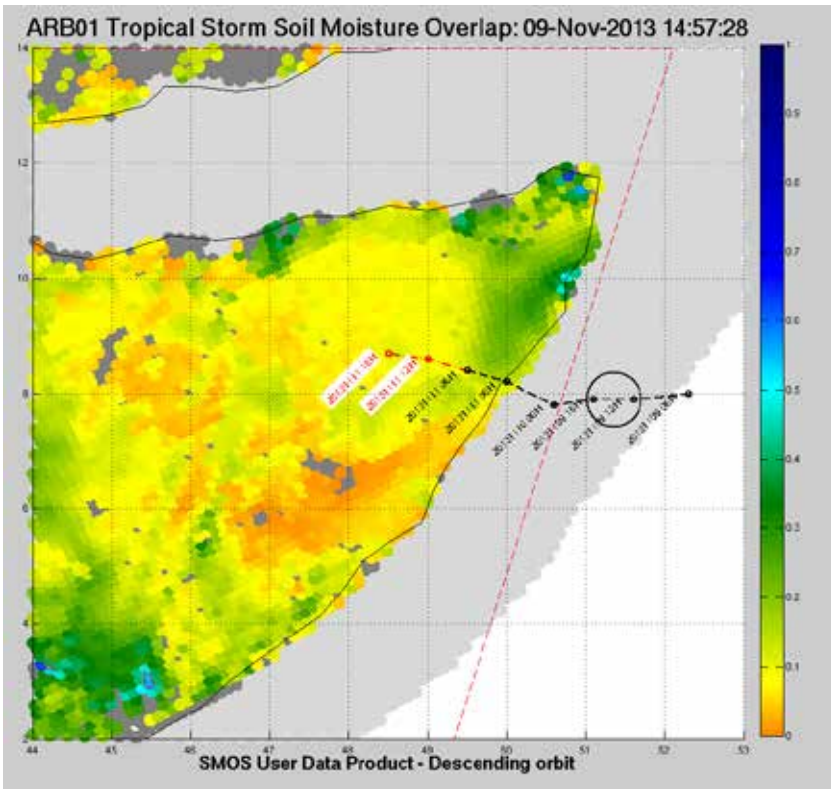
Summary



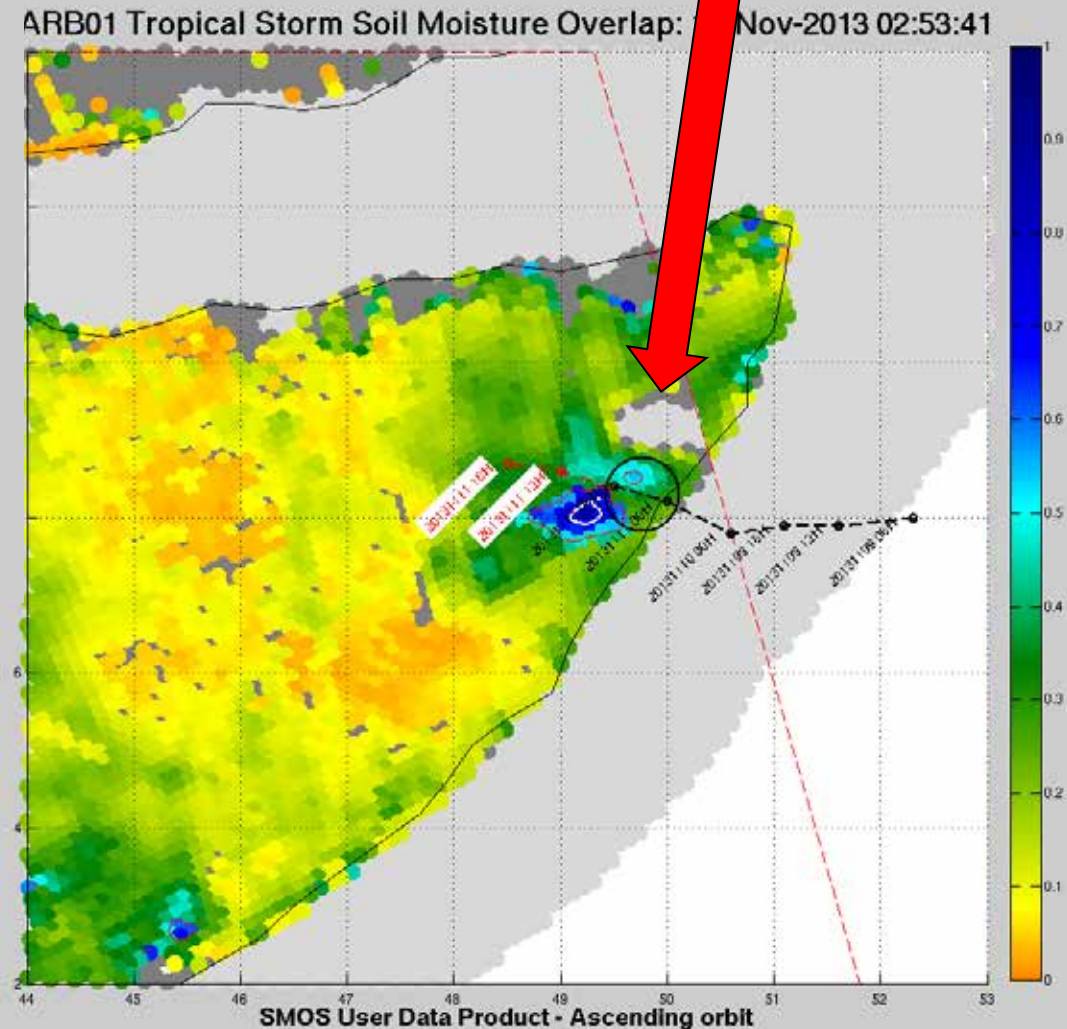
- q After several years in orbit **SMOS, is behaving very well**
- q SMOS as still **room for improvements** and is **evolving very positively**
 - ✓ Significant improvement in calibration and retrieval algorithms
 - ✓ Significant reprocessing efforts soon underway
 - ✓ Intercomparisons → Not a beauty contest but a base to learn and understand more
- q **First applications are very interesting**
 - ✓ Many of them
 - ✓ Identification of research areas
- q **Getting ready to bridging with SMAP** and establish long term data sets
- q Synergistic approaches studies (AMSR, ASCAT, GCOM-W, Aquarius...)
- q Next Step is the elaboration of an **Essential Climate Variable** by bridging several datasets

Example of Tropical storm Nov 2013

visit http://www.cesbio.ups-tlse.fr/SMOS_blog/



Rains > 20 mm/h for ECMWF



**Thank you for
your patience
Any questions?**