

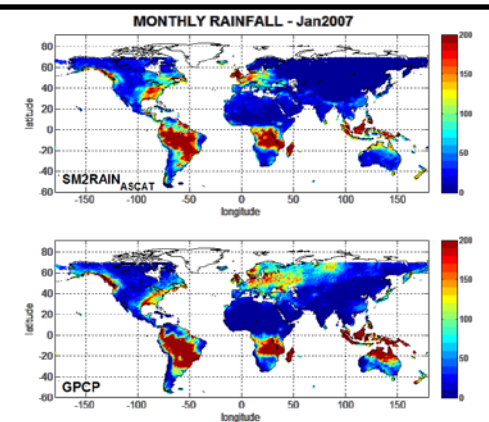
NEW INFORMATION CONTENT FROM SOIL MOISTURE DATA: RAINFALL ESTIMATION AND FLOOD MODELLING



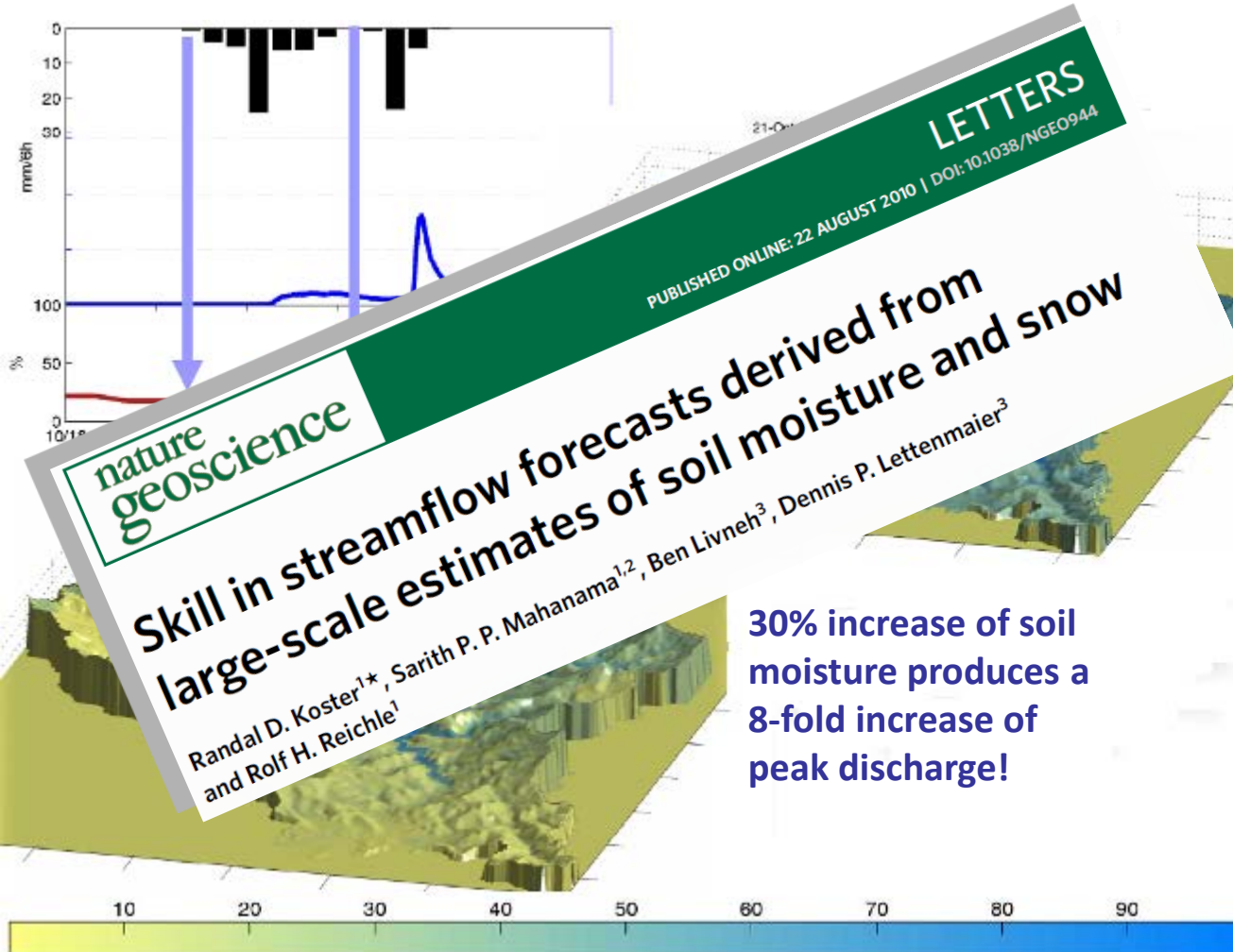
Luca Brocca, Christian Massari, Luca Ciabatta, Tommaso Moramarco

Research Institute for Geo-Hydrological Protection (IRPI-CNR), Perugia, Italy

...and many others: W. Wagner, W. Dorigo, S. Hahn, S. Hasenauer, C. Albergel, P. De Rosnay, S. Puca, S. Gabellani, R. Parinussa, R. De Jeu, P. Matgen, D. Penna, J. Martinez-Fernandez, V. Levizzani, ...



INTRODUCTION



Runoff generation
Brocca et al. (2010, HESS+WRR, 2012 TGRS, ...)

Drought monitoring
Rahmani et al. (2015, JAG, in press)

Erosion modelling
Todisco et al. (2015, HESSD)

Landslide prediction
Brocca et al. (2012, RS)

Numerical Weather Prediction
Capecchi & Brocca (2014, METZET)

Plant production
Bolten et al. (2010, JSTARS)

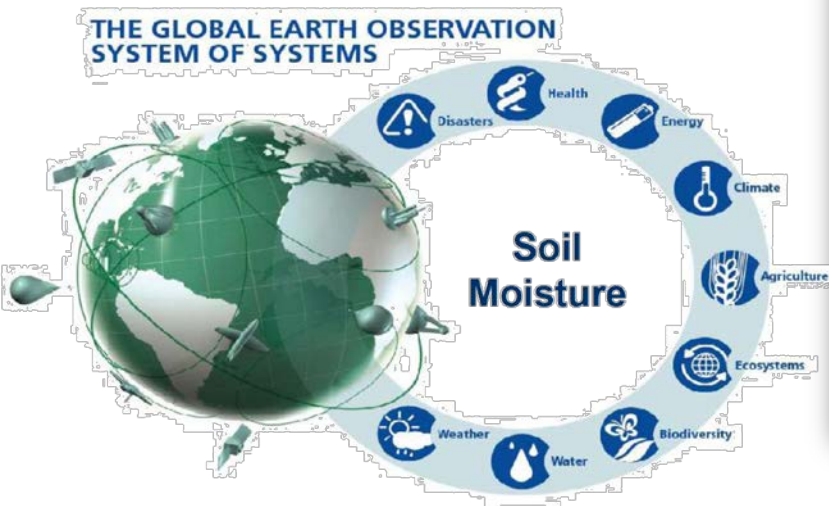
Water quality modelling
Han et al. (2012, HYP)

INTRODUCTION

Soil moisture is needed by all GEO Social Benefit Areas and was ranked the **second** top priority parameter (behind precipitation) in a year 2010 GEO report on "Critical Earth Observation Priorities".

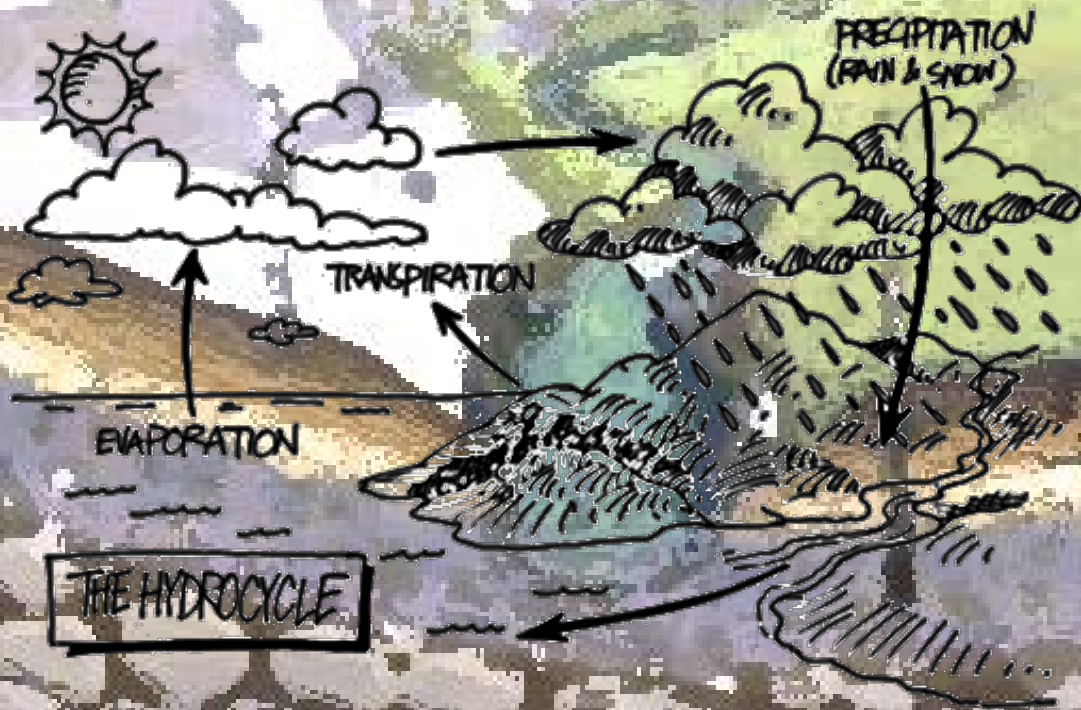
25 Highest-Ranked Earth Observations and Associated SBAs

Earth Observation Parameter	GEO Societal Benefits Areas*							
	Agriculture	Climate	Disasters	Ecosystems	Energy	Health	Water	Weather
Precipitation								
Soil Moisture								
Surface Air Temperature								
Surface Wind Speed								
Land Cover								
Surface Humidity								
Vegetation Cover								
Surface Wind Direction								
Normalized Difference Vegetation Index								
Sea Surface Temperature								

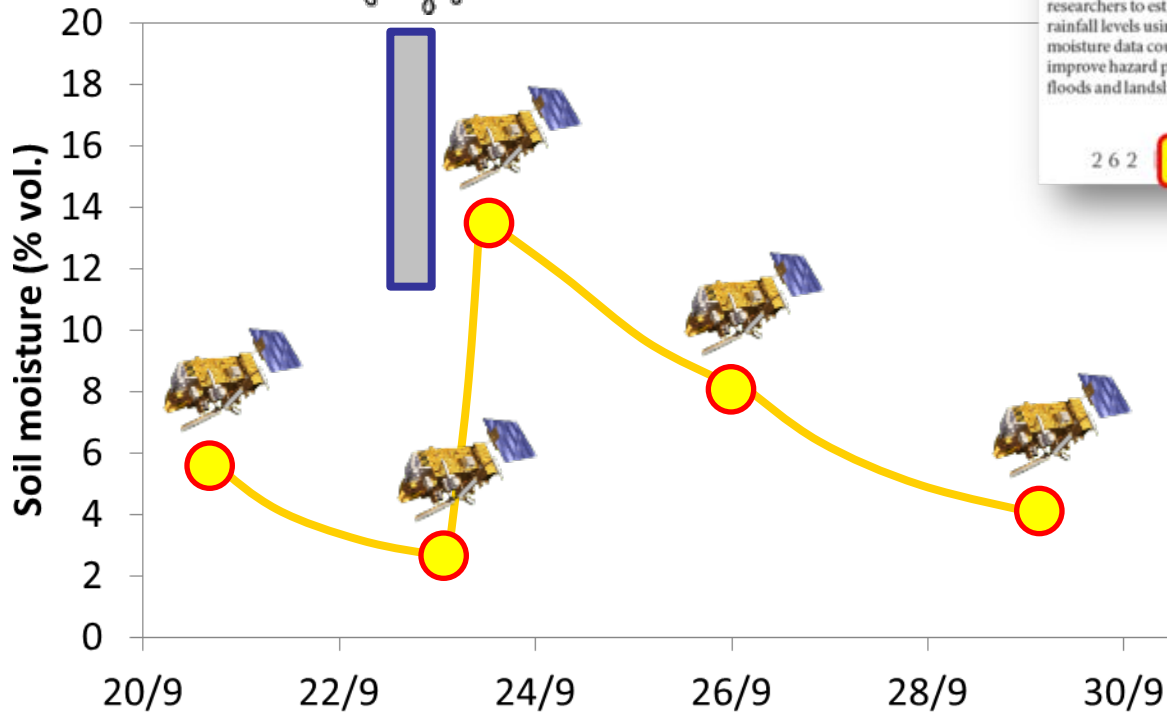
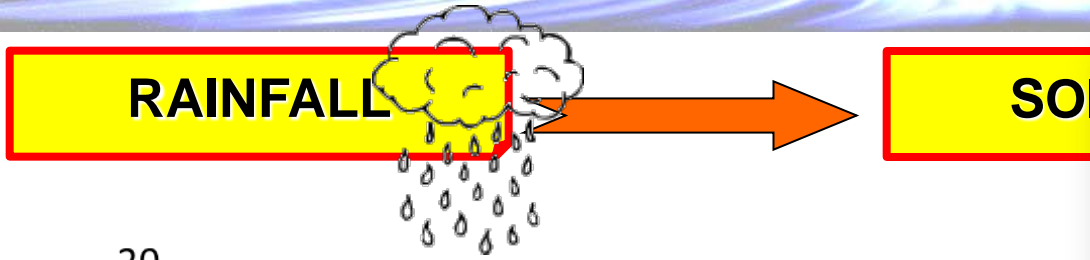


http://sbageotask.larc.nasa.gov/US-09-01a_SummaryBrochure.pdf

RAINFALL ESTIMATION FROM SOIL MOISTURE OBSERVATIONS: **SM2RAIN**



What is SM2RAIN?



RESEARCH HIGHLIGHTS

Selections from the scientific literature

ATMOSPHERIC SCIENCE

Detecting rainfall from the bottom up

A method that allows researchers to estimate global rainfall levels using soil-moisture data could help to improve hazard planning for floods and landslides.

To estimate rainfall in places that lack ground-based rain gauges, researchers rely on satellite data of atmospheric moisture, but this is notoriously inaccurate. Luca Brocca at the National Research Council in Perugia, Italy, and his colleagues developed an algorithm that calculates rainfall amounts on the basis of satellite data on soil moisture. They compared their estimates with rain-gauge data and found that their method accurately estimates rainfall in several regions around the world.

Moreover, their algorithm is better than a state-of-the-art method at detecting light rainfall events and precipitation at high latitudes.

J. Geophys. Res. Atmos.
<http://doi.org/sp7> (2014)

262 **NATURE** VOL 509 | 15 MAY 2014



The soil moisture variations are strongly related to the amount of rainfall falling into the soil. Therefore, we can use soil moisture observations for estimating rainfall by considering the “soil as a natural rain gauge”.

SM2RAIN

SM2RAIN algorithm

relative saturation precipitation surface runoff drainage

$$Z ds(t)/dt = p(t) - r(t) - e(t) - g(t)$$

soil water capacity = soil depth X porosity evapotranspiration

Soil water
balance
equation

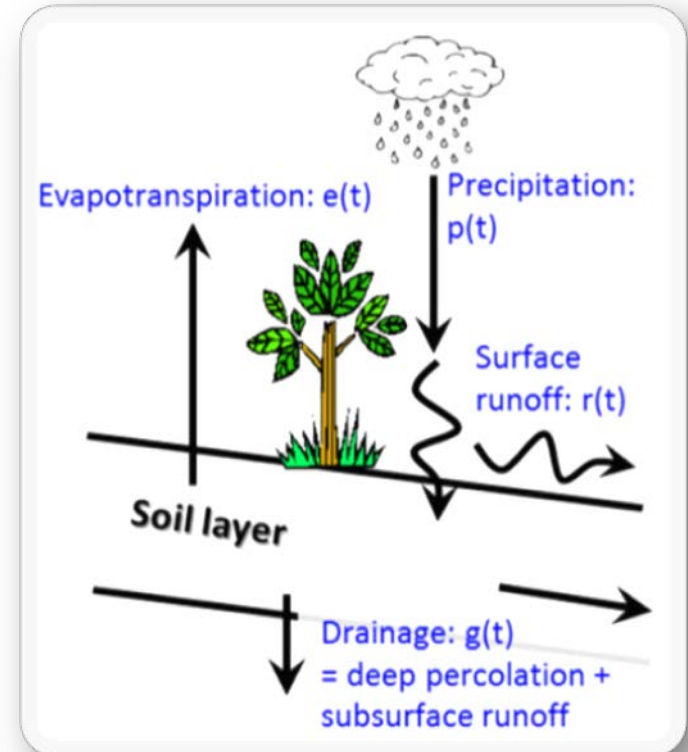
Inverting for $p(t)$:

$$p(t) = Z ds(t)/dt + r(t) + e(t) + g(t)$$

Assuming: $g(t) = a s(t)^b + e(t) = 0 + r(t) = 0$
during rainfall



$$p(t) \cong Z ds(t)/dt + a s(t)^b$$



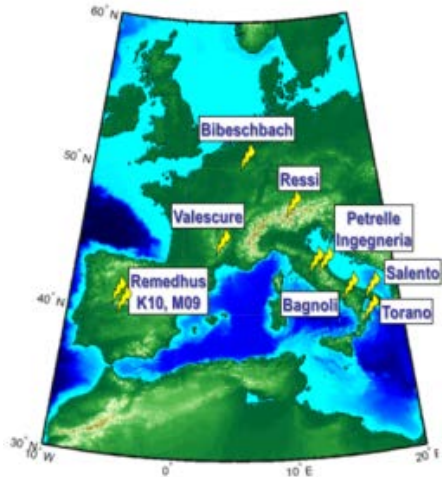
SM2RAIN: in situ observations

MOISST 2015
Stillwater
2nd June 2015
Brocca Luca

Hydrol. Hydromech., 63, 2015, X, XXX-XXX
DOI: 10.1515/johh-2015-0016

Rainfall estimation from in situ soil moisture observations at several sites in Europe: an evaluation of the SM2RAIN algorithm

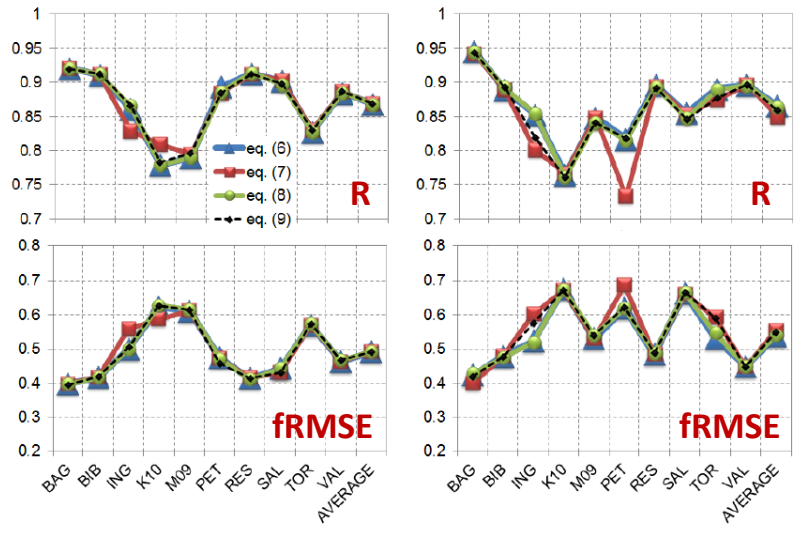
Luca Brocca^{1*}, Christian Massari¹, Luca Ciabatta¹, Tommaso Moramarco¹, Daniele Penna²,
Giulia Zuecco³, Luisa Pianezzola³, Marco Borga³, Patrick Matgen⁴, José Martínez-Fernández⁵



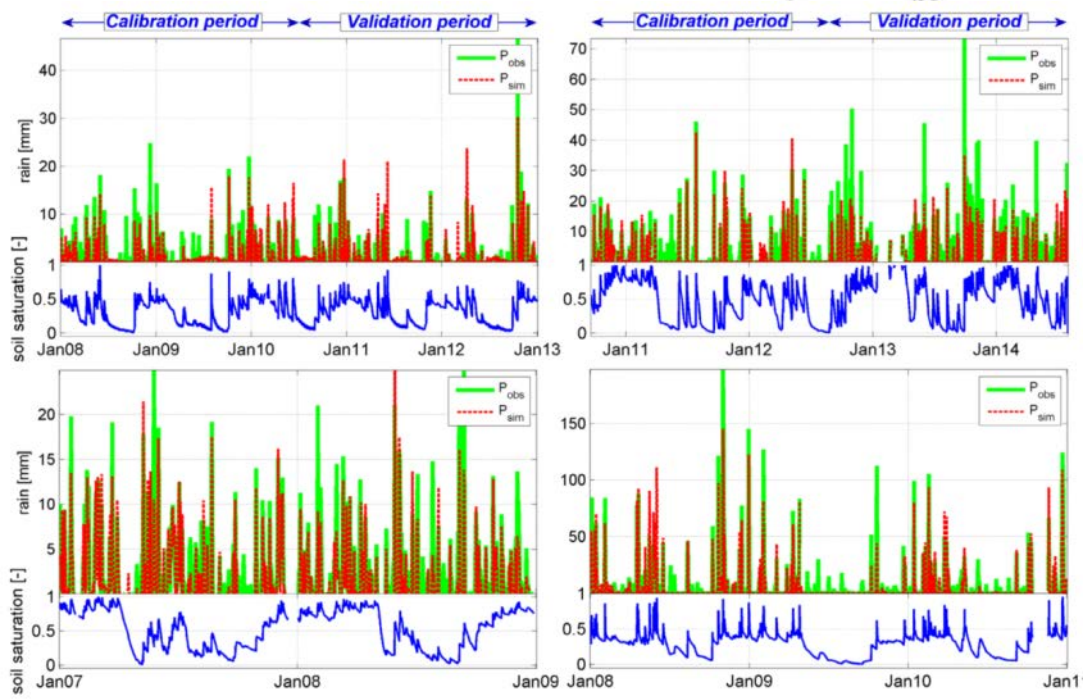
0.75 < R < 0.95

calibration

validation

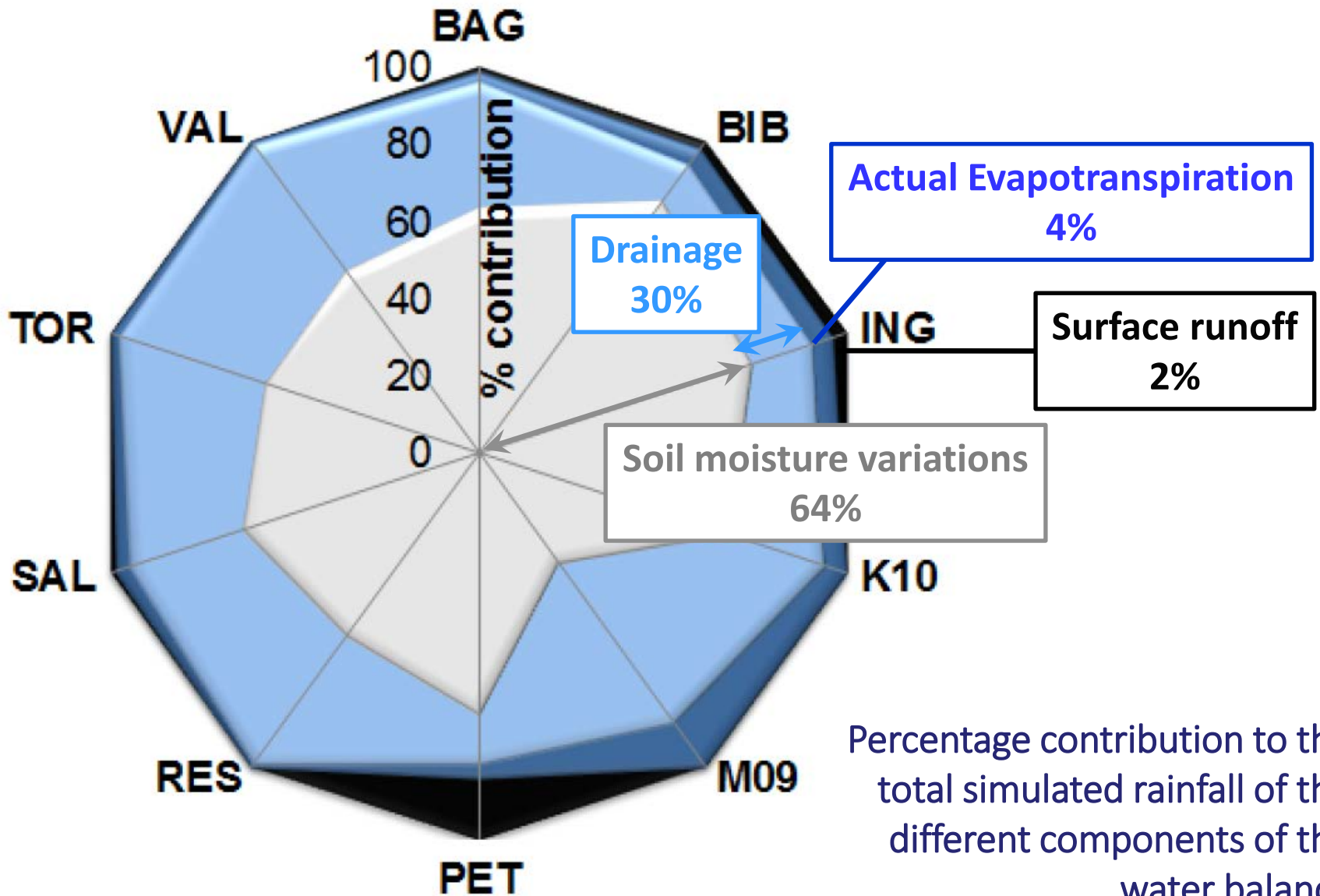


In situ soil moisture observations



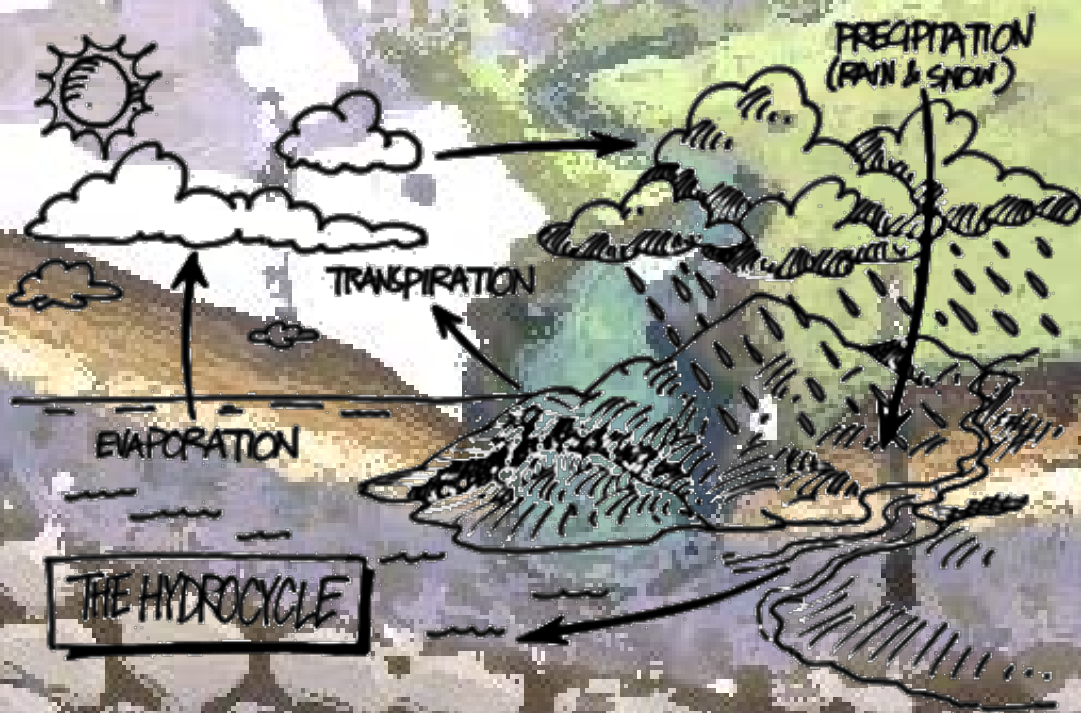
SM2RAIN: in situ observations

MOISST 2015
Stillwater
2nd June 2015
Brocca Luca



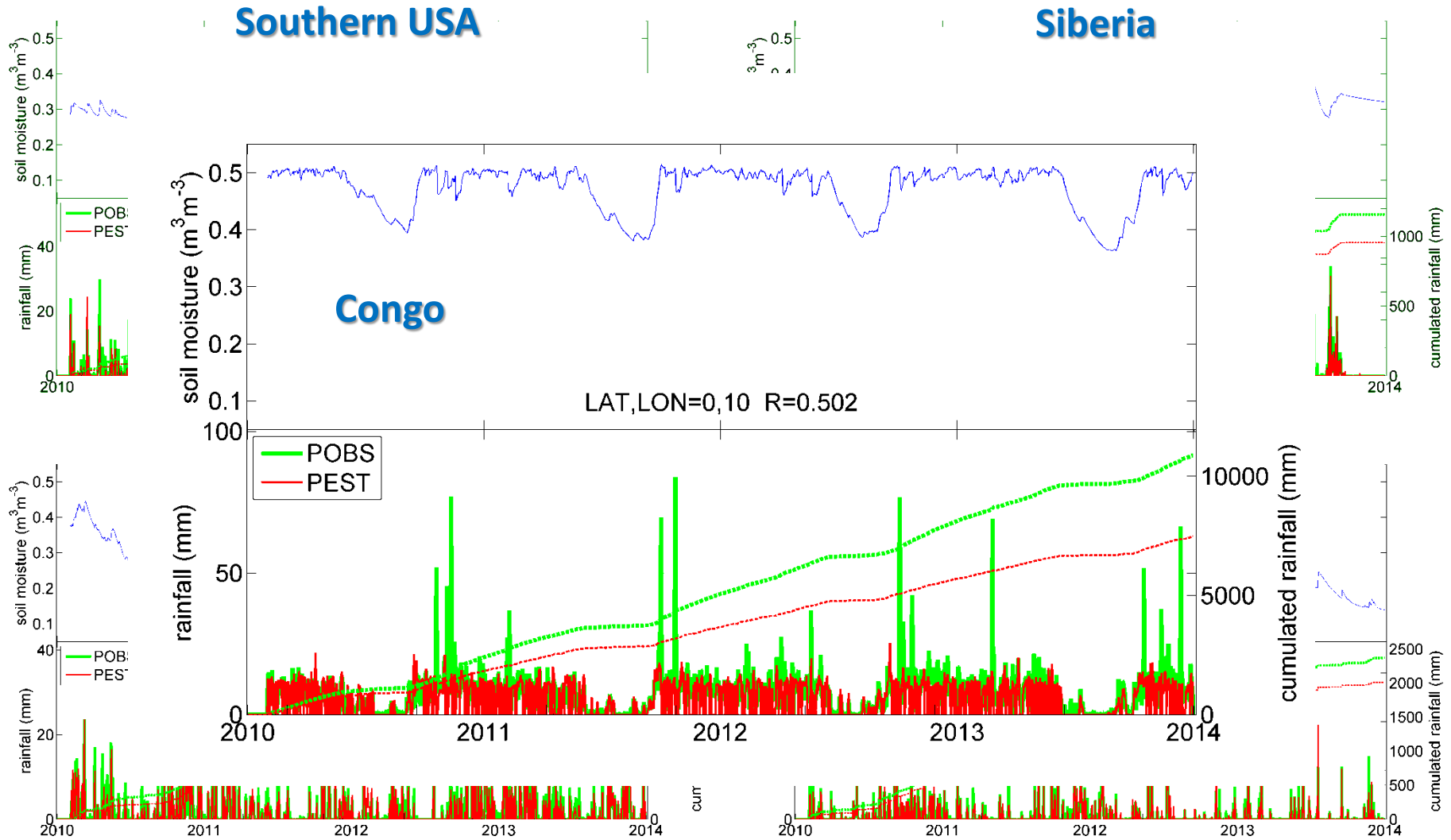
Percentage contribution to the total simulated rainfall of the different components of the water balance

GLOBAL SCALE APPLICATIONS WITH SATELLITE SOIL MOISTURE PRODUCTS



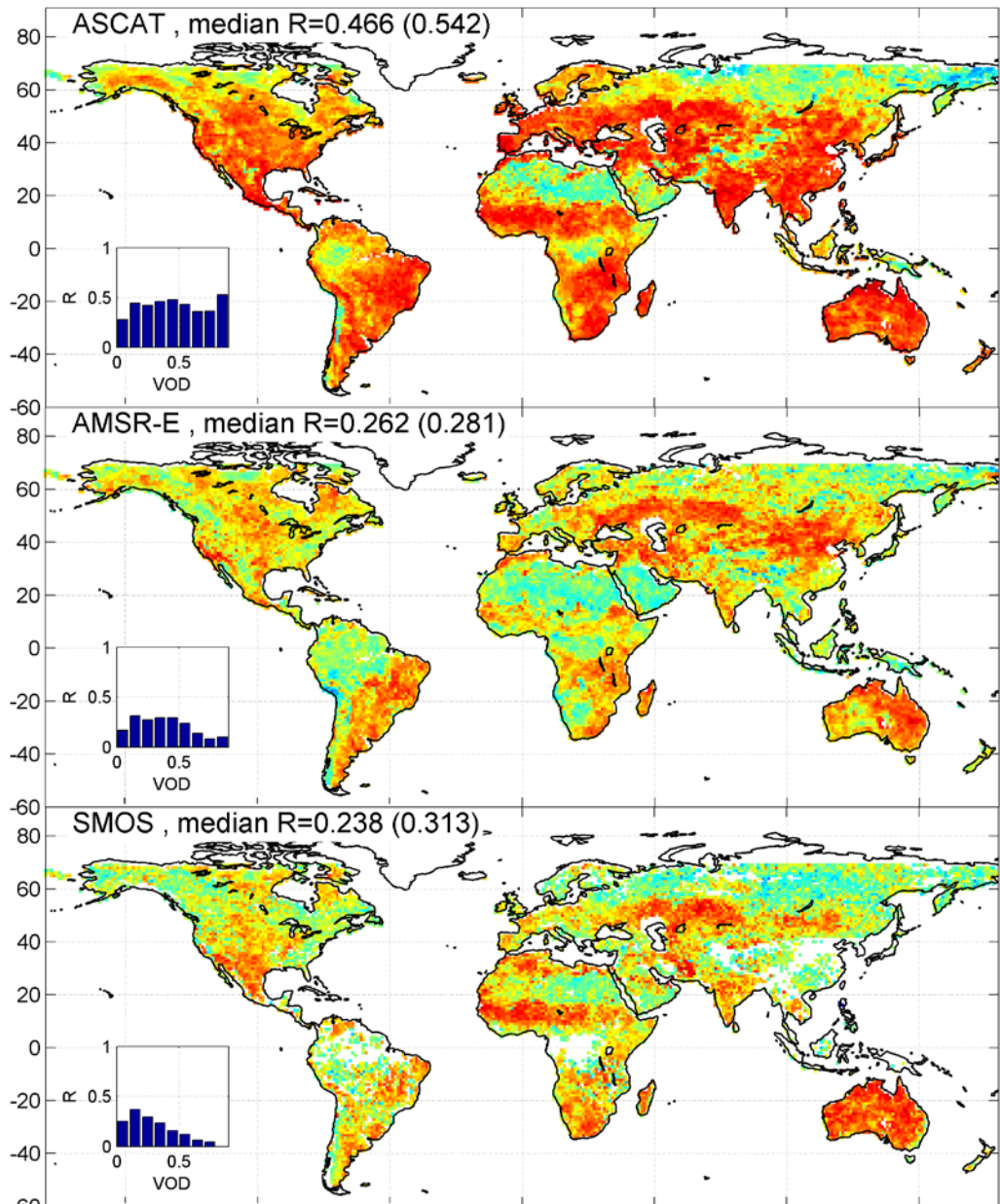
CRASH TEST: timeseries

MOISST 2015
Stillwater
2nd June 2015
Brocca Luca

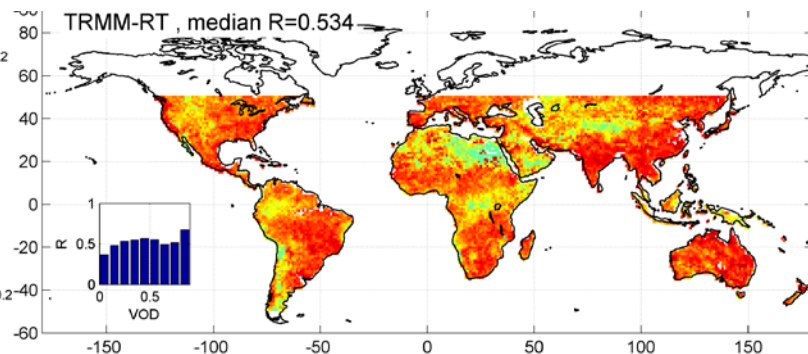


Global scale: real satellite observations

MOISST 2015
Stillwater
2nd June 2015
Brocca Luca



Correlation map between 5-day rainfall from GPCP and the rainfall product obtained from the application of SM2RAIN algorithm to ASCAT, AMSR-E and SMOS data plus TMPA 3B42RT (VALIDATION period 2010-2011)



AGU PUBLICATIONS

JGR

RESEARCH ARTICLE

Soil as a natural rain gauge: Estimating global rainfall from satellite soil moisture data

Luca Brocca¹, Luca Ciabatta¹, Christian Massari¹, Tommaso Moramarco¹, Sebastian Hahn², Stefan Hasenauer², Richard Kidd², Wouter Dorigo², Wolfgang Wagner², and Vincenzo Levizzani¹

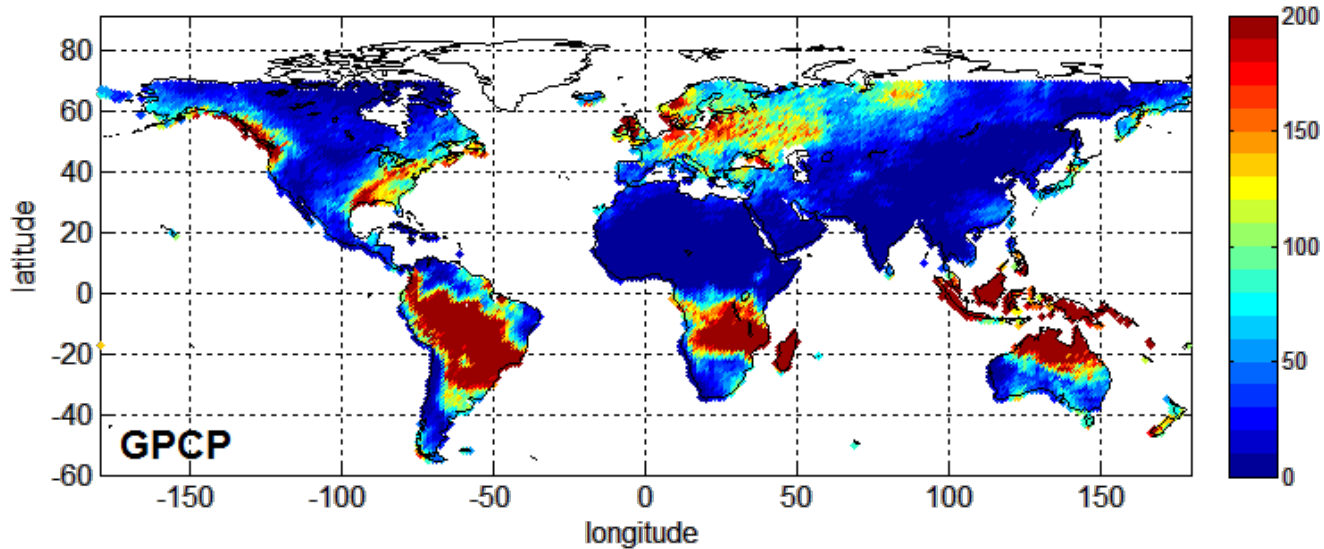
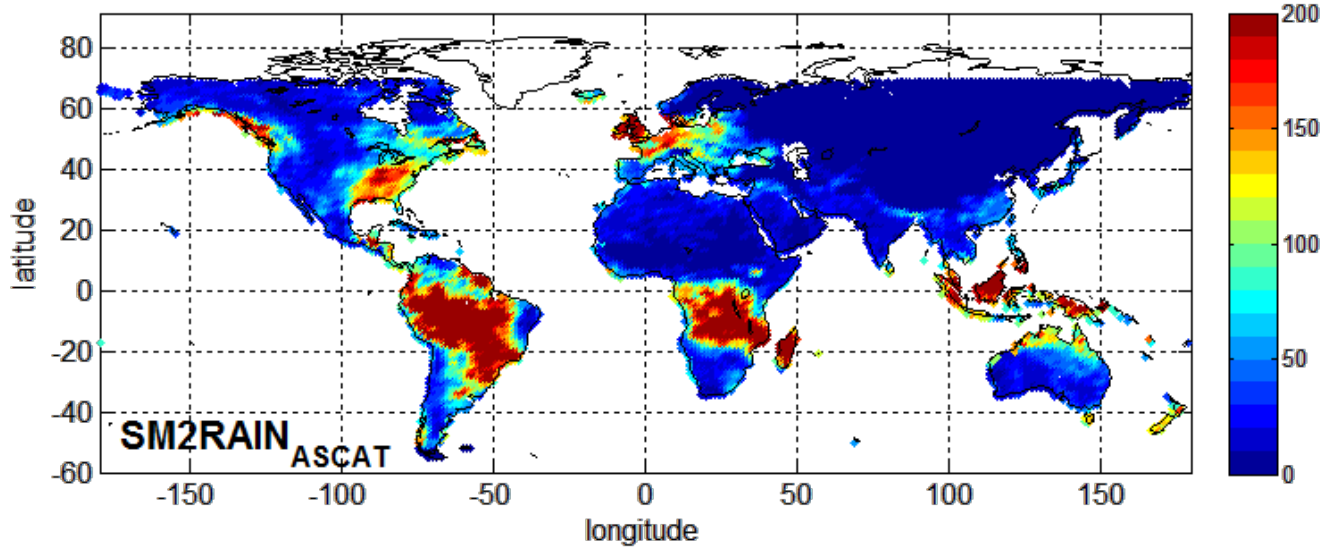
¹Research Institute for Geo-Hydrological Protection, National Research Council, Perugia, Italy; ²Department of Geodesy and Geoinformation, Vienna University of Technology, Vienna, Austria; ³Institute of Atmospheric Sciences and Climate, National Research Council, Bologna, Italy

Abstract Measuring precipitation intensity is not straightforward, and over many areas, ground observations are lacking and satellite observations are used to fill this gap. The most common way of retrieving rainfall is by hydrometereon. However, most applications are interested in how much water reaches the ground, a problem that is notoriously difficult to solve from a top-down perspective. In this study, a novel "bottom-up" approach is proposed that, by doing "hydrology backward", uses variations in soil moisture (SM) sensed by microwave

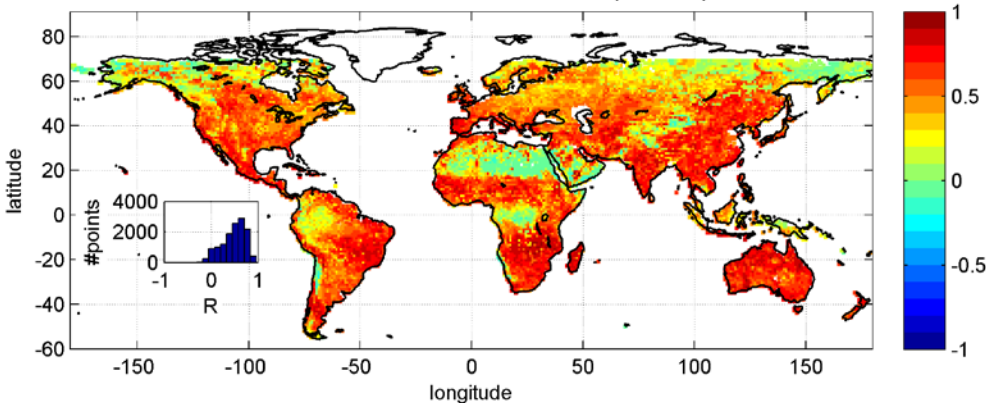
Global monthly rainfall from ASCAT

MOISST 2015
Stillwater
2nd June 2015
Brocca Luca

MONTHLY RAINFALL - Jan2007

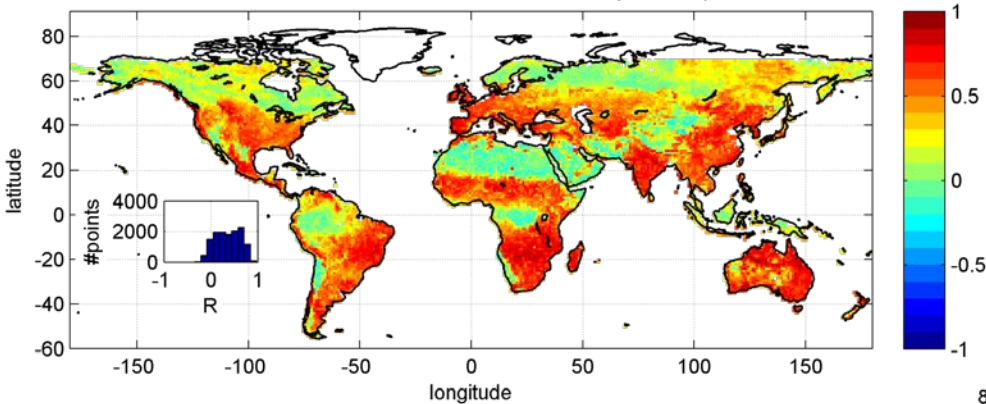


ASCAT, median R=0.526 (0.598)



- ✓ 2007-2009
- ✓ Improved algorithm
- ✓ Distributed calibration
- ✓ Application of filtering techniques (e.g. wavelet)

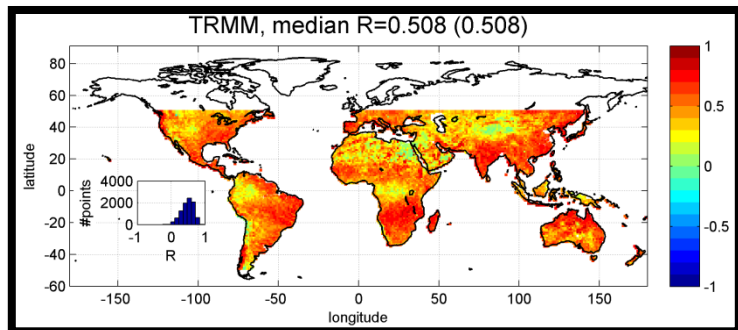
QSCAT, median R=0.369 (0.491)



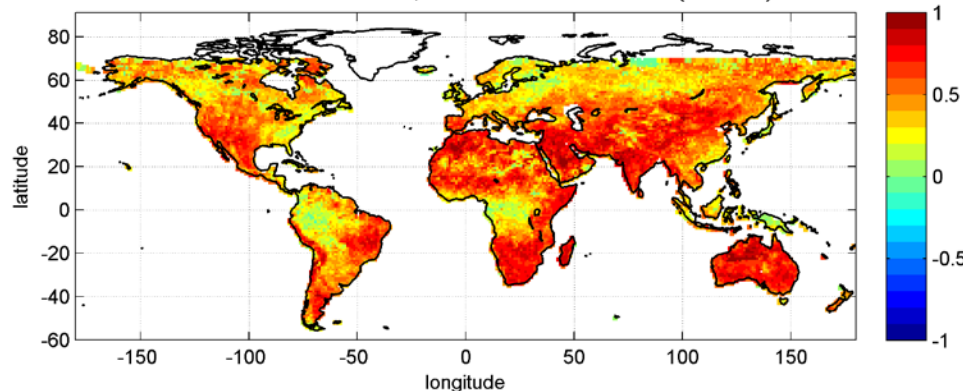
Application to **QUIKSCAT**
(Ku-band scatterometer),
2007-2009

Application to
AMSR2 (C+X-band radiometer)
2012-2014

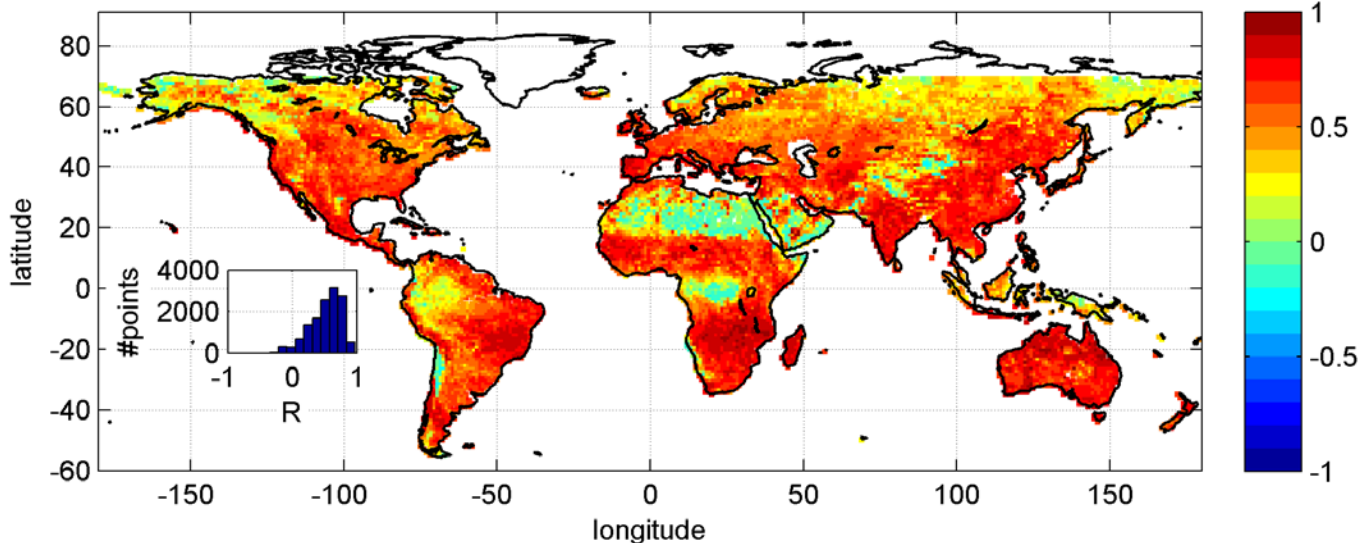
TRMM, median R=0.508 (0.508)



AMSR2-DES-best, median R=0.496 (0.572)

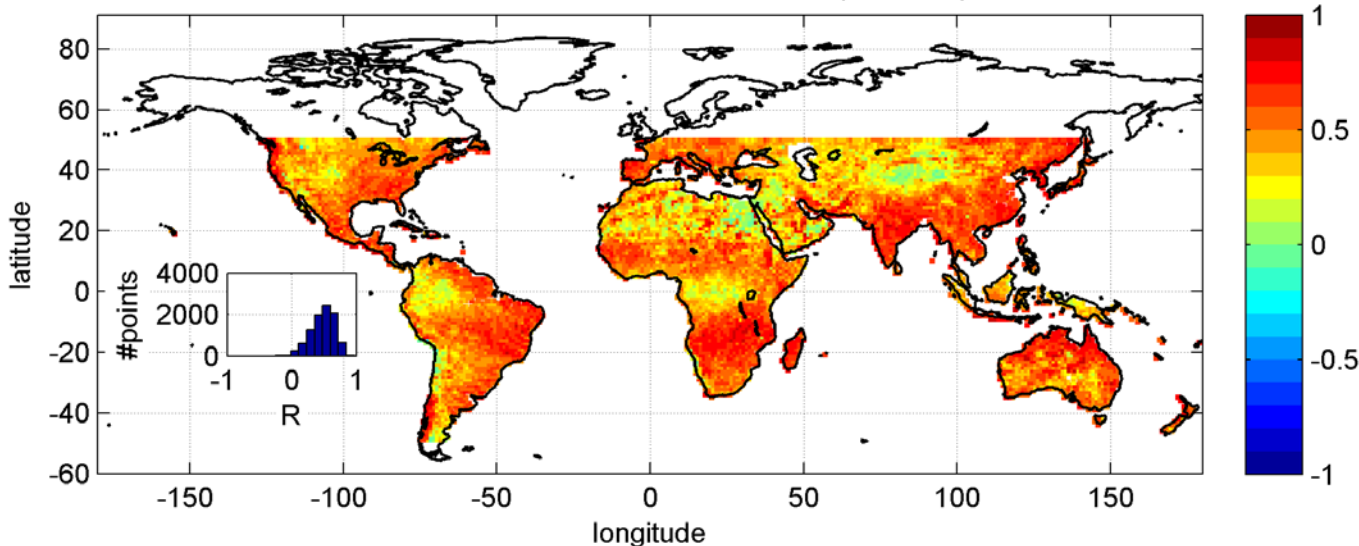


ASCAT-QSCAT, median R=0.567 (0.640)



- ✓ 2007-2009
- ✓ ERA-Interim as benchmark
- ✓ 5-day cumulated

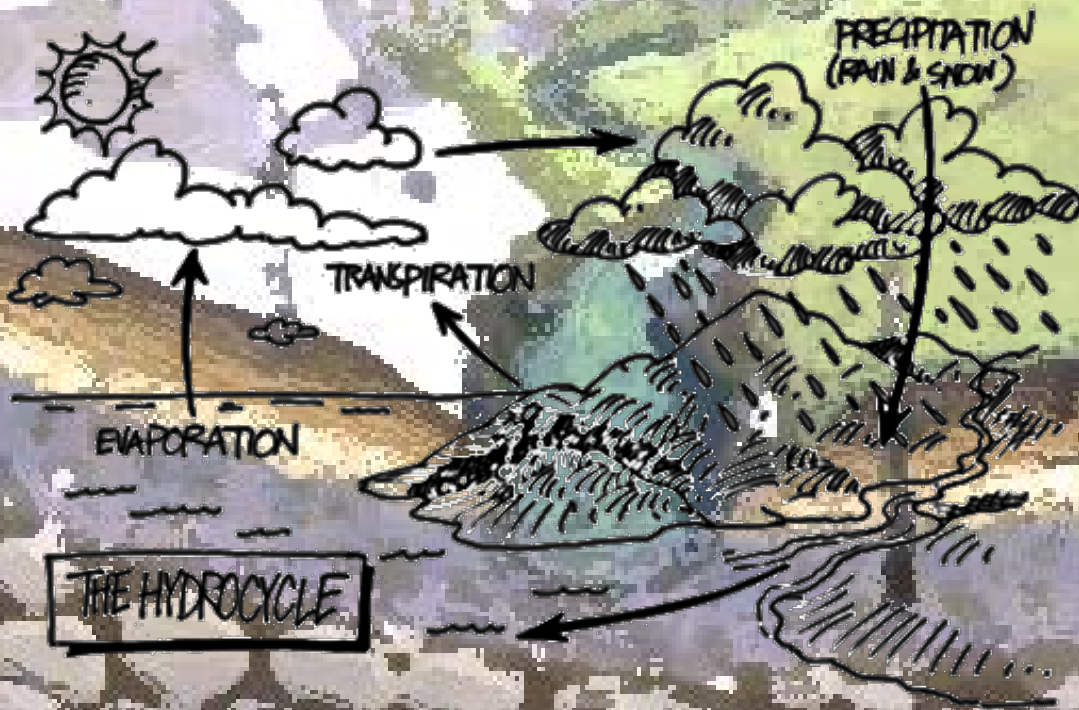
TRMM, median R=0.508 (0.508)



The correlation is 25% higher than TMPA real time rainfall product

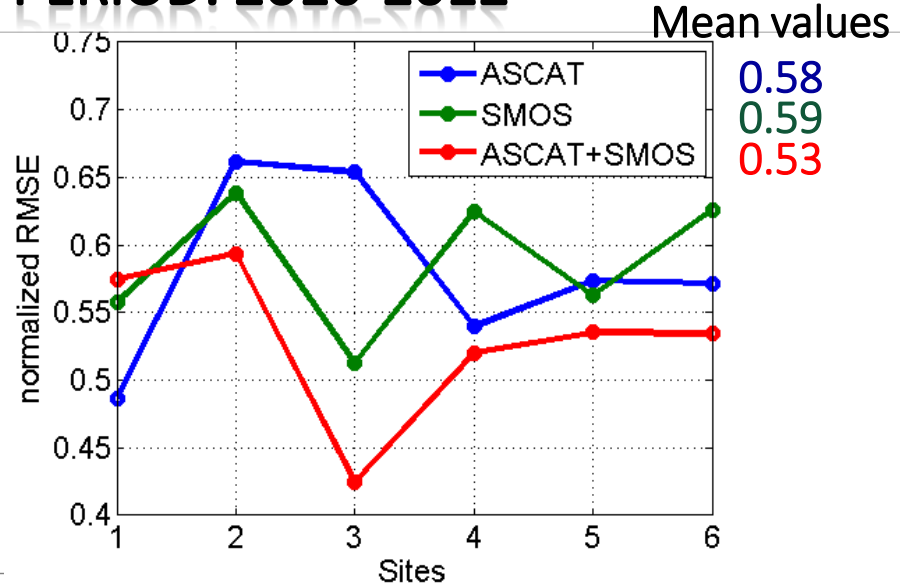
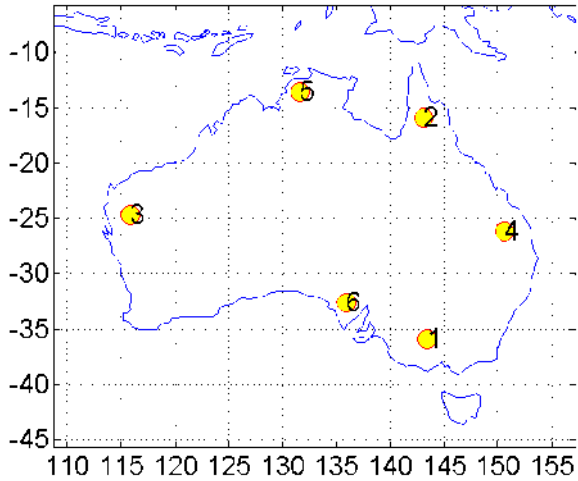
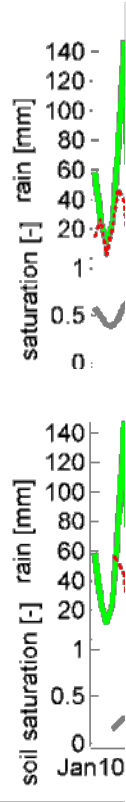
0.508 → 0.640

MERGING MULTIPLE SATELLITE SOIL MOISTURE PRODUCTS

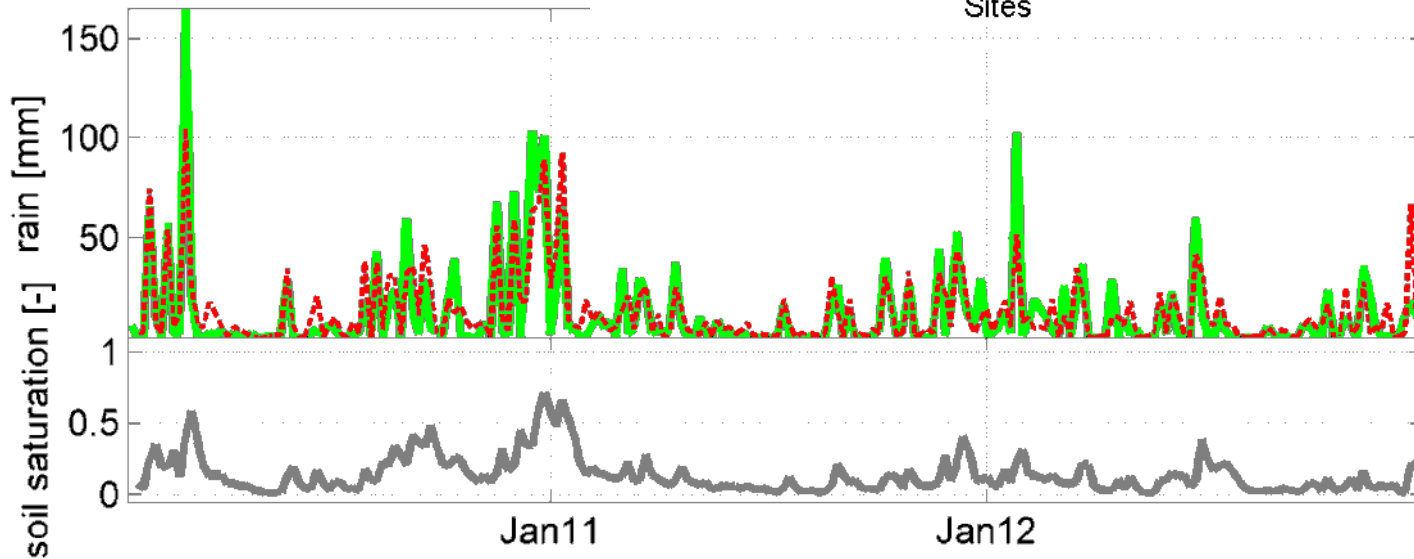


SMOS + ASCAT

EXTENDED PERIOD: 2010-2012



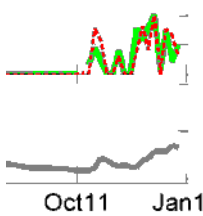
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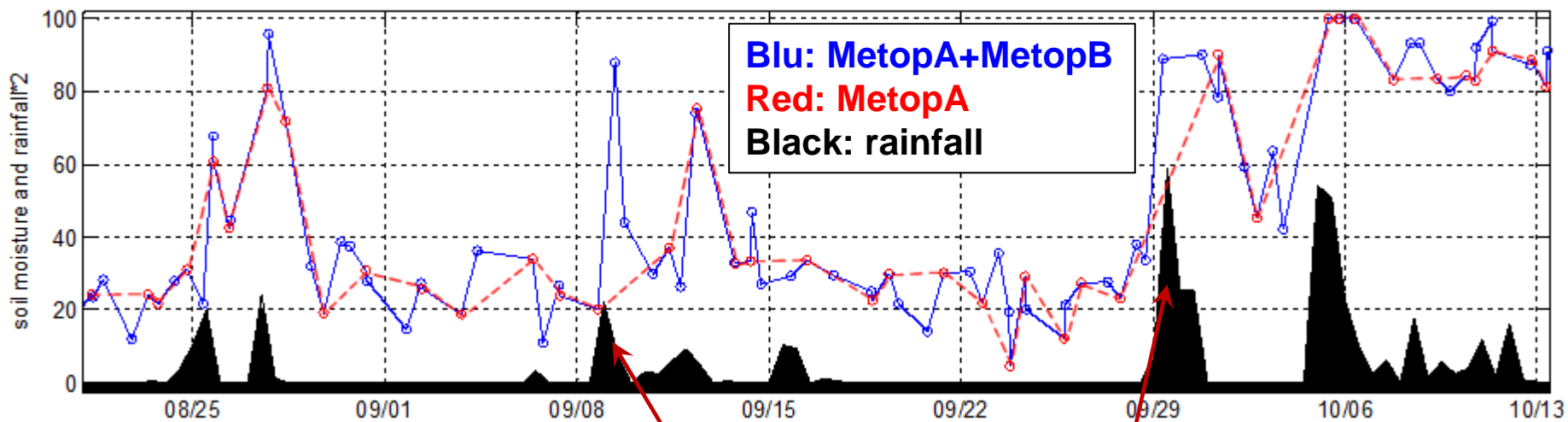
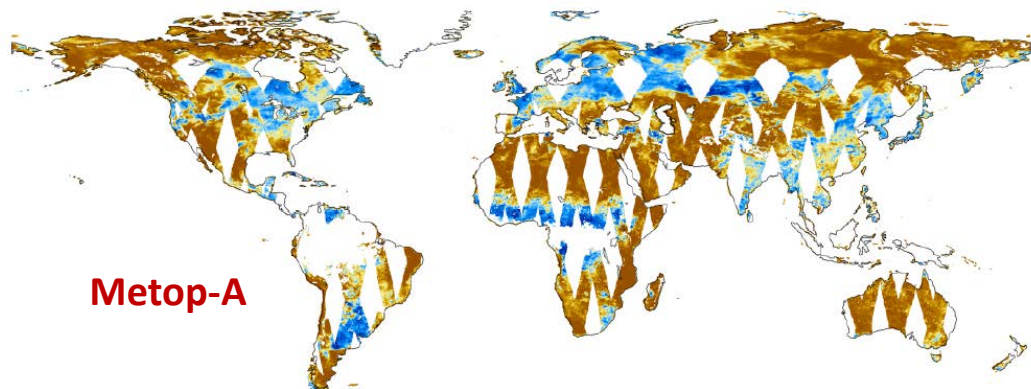


RMS

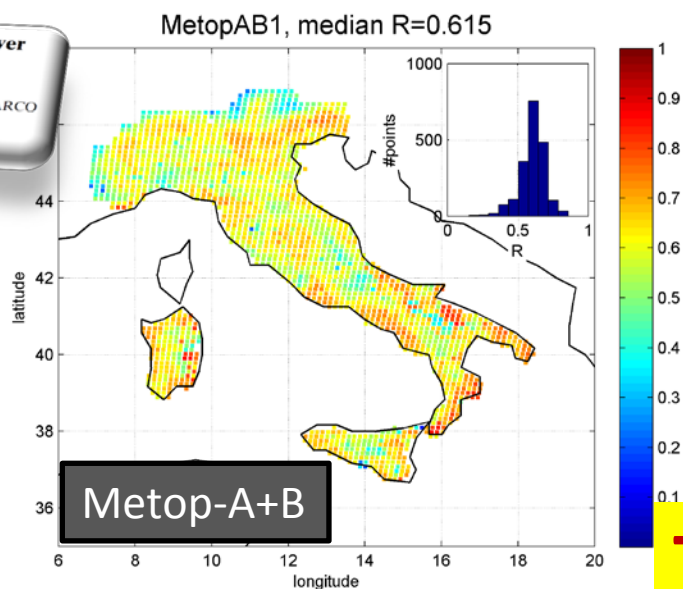
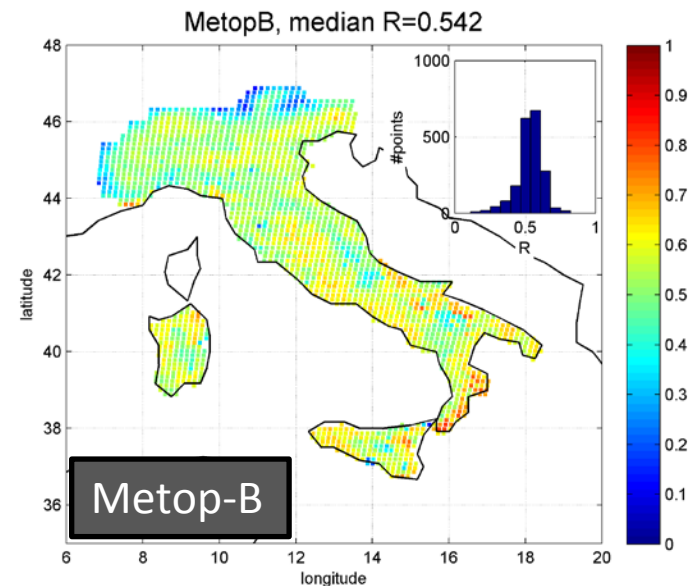
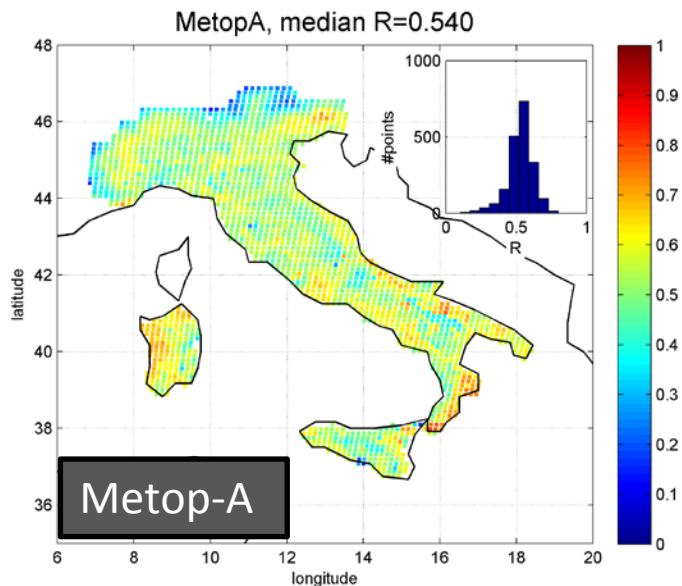
22.826 mm

ASCAT+
SMOS





Rainfall events detected only by considering both sensors Metop A and B



Metop A+B improves by
14% wrt single sensors

Metop A+B improves by
37% wrt Ciabatta et al.
JHM (calibration period)

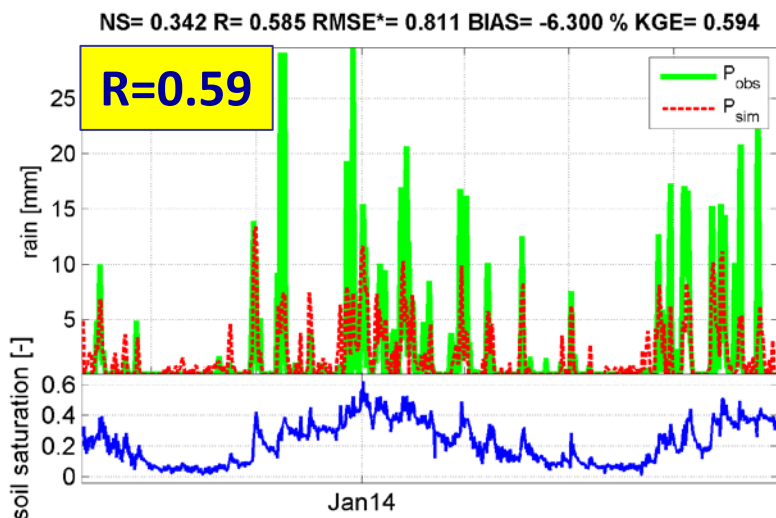
Time step: 1-day

Integration of Satellite Soil Moisture and Rainfall Observations over the Italian Territory
LUCA CIABATTA, LUCA BROCCA, CHRISTIAN MASSARI, AND TOMMASO MORAMARCO
Research Institute for Geo-Hydrological Protection, National Research Council, Perugia, Italy

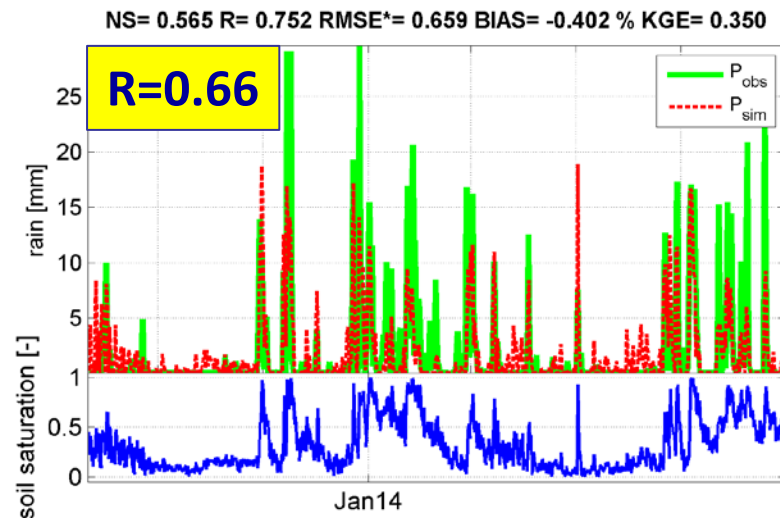
Ciabatta et al. (2015) – JHM
2010-2013
Median R= 0.44
T=0 (no SWI application)

Data Period:
2-may-2013 → 31-dec-2014

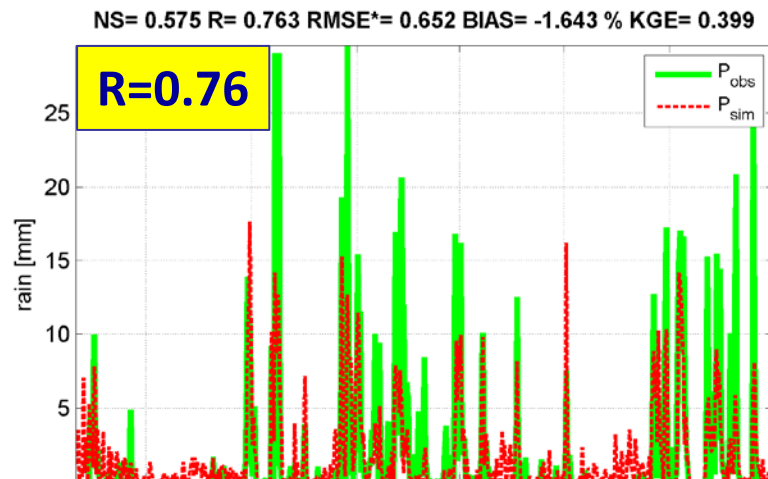
AMSR2



ASCAT



AMSR2+ASCAT



Lon,Lat: -5,40

Period: may 2013-dec 2014

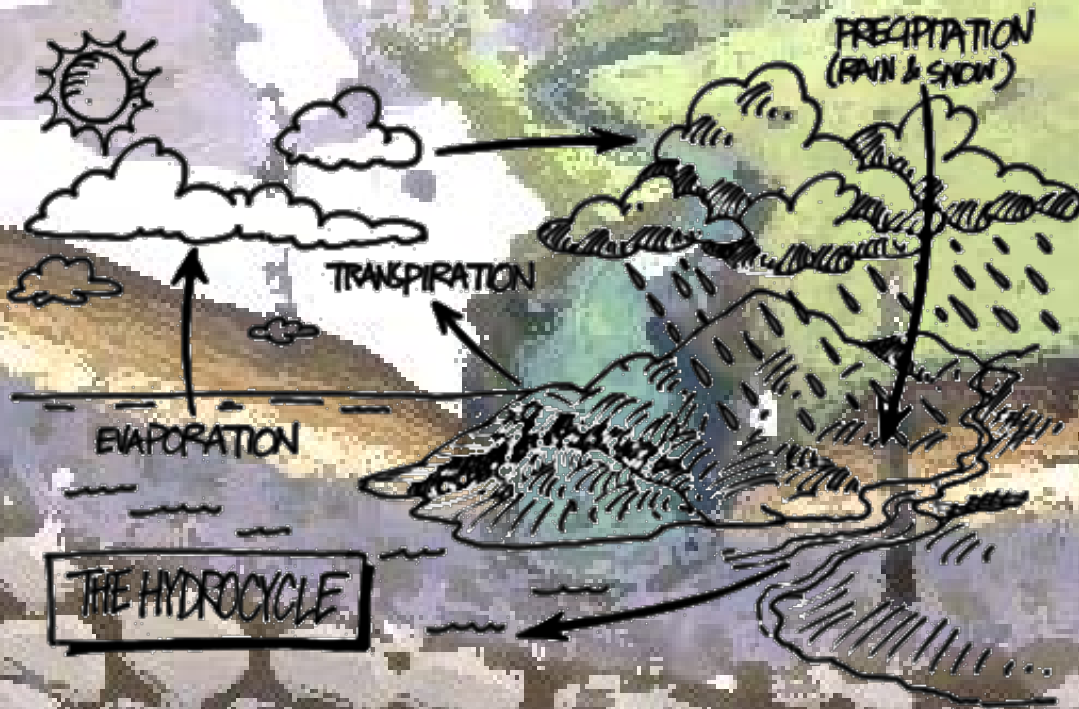
ASCAT: Metop A+B

AMSR2: Asc+Desc, X-band

Benchmark: ERA-Interim

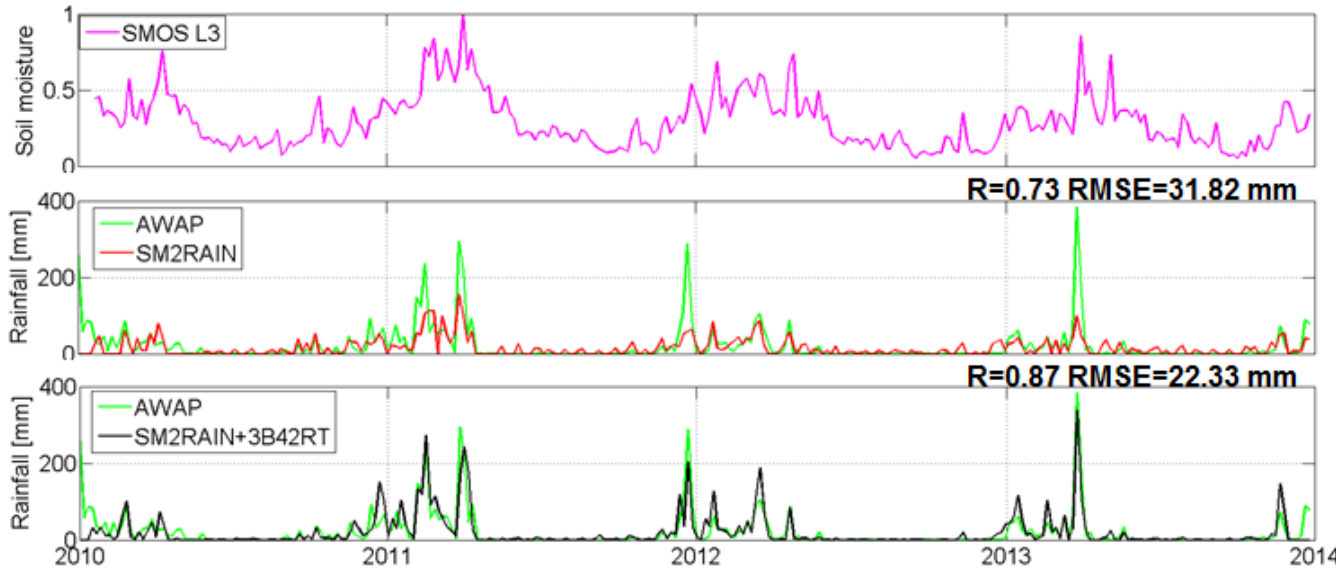
Time step: 1-day

INTEGRATION OF SM2RAIN WITH STATE-OF-THE-ART RAINFALL PRODUCTS



SMOS+TMPA in Australia

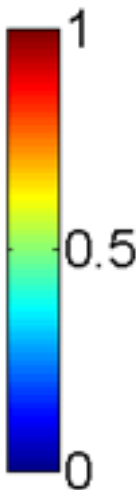
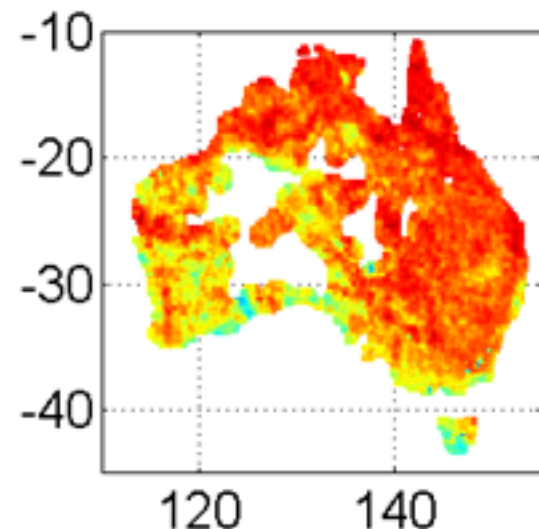
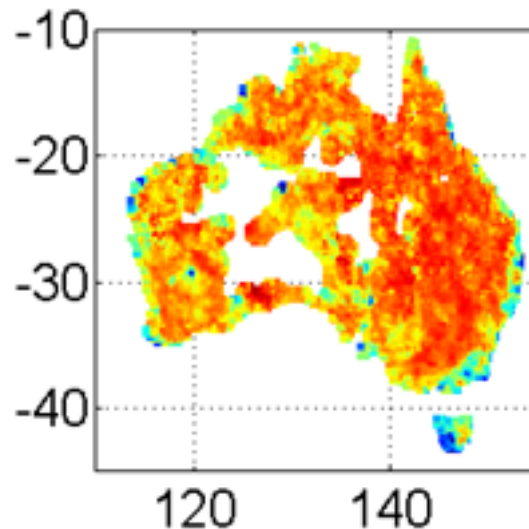
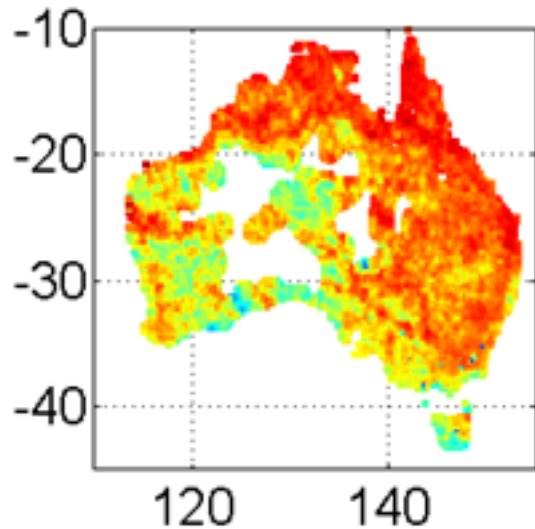
- Period 2010-2013
- AWAP rainfall as benchmark
- Calibration with 3B42
- 5-day cumulated



3B42RT
median $R=0.714$

SM2R_{SMOS}
median $R=0.733$

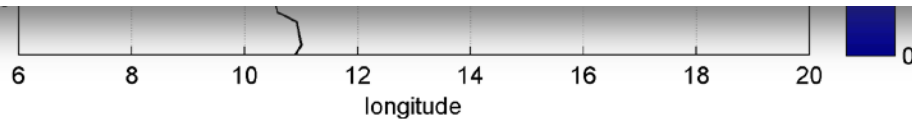
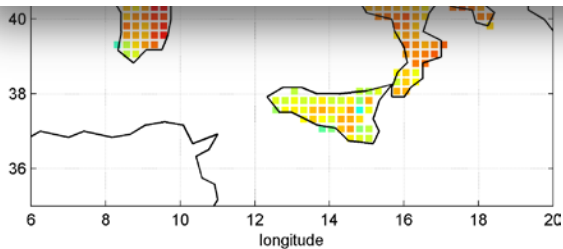
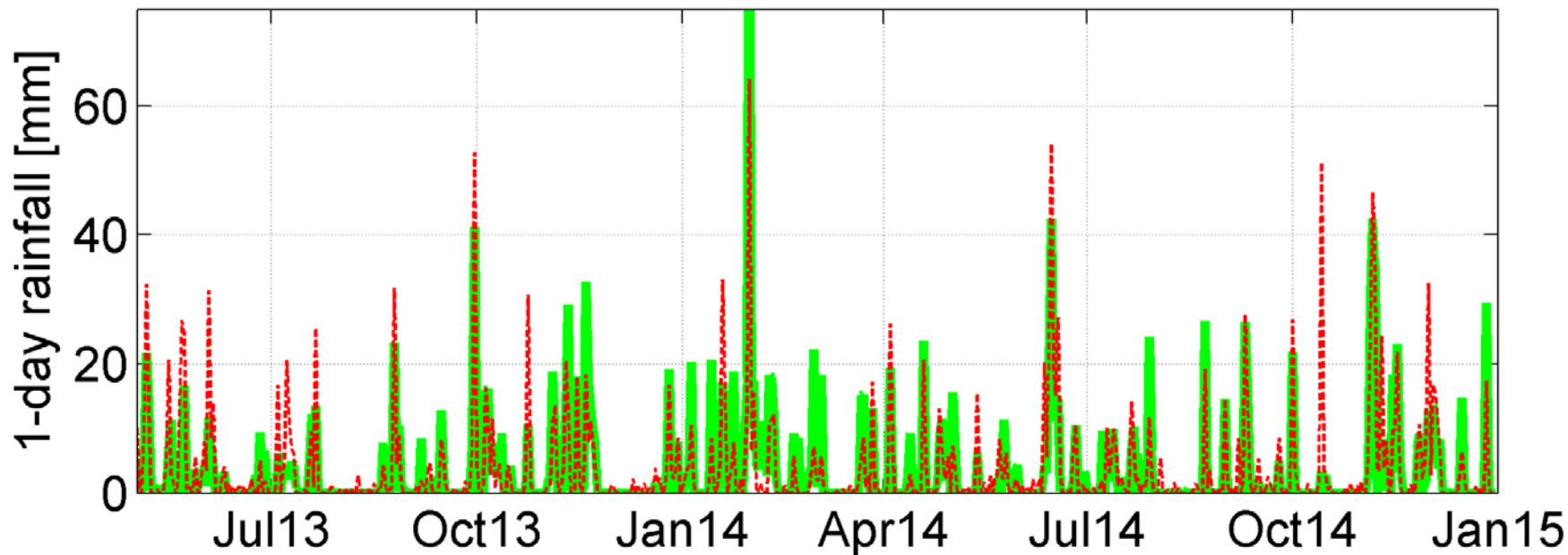
SM2R_{SMOS} + 3B42RT
median $R=0.757$



ASCAT-AB, median R=0.681

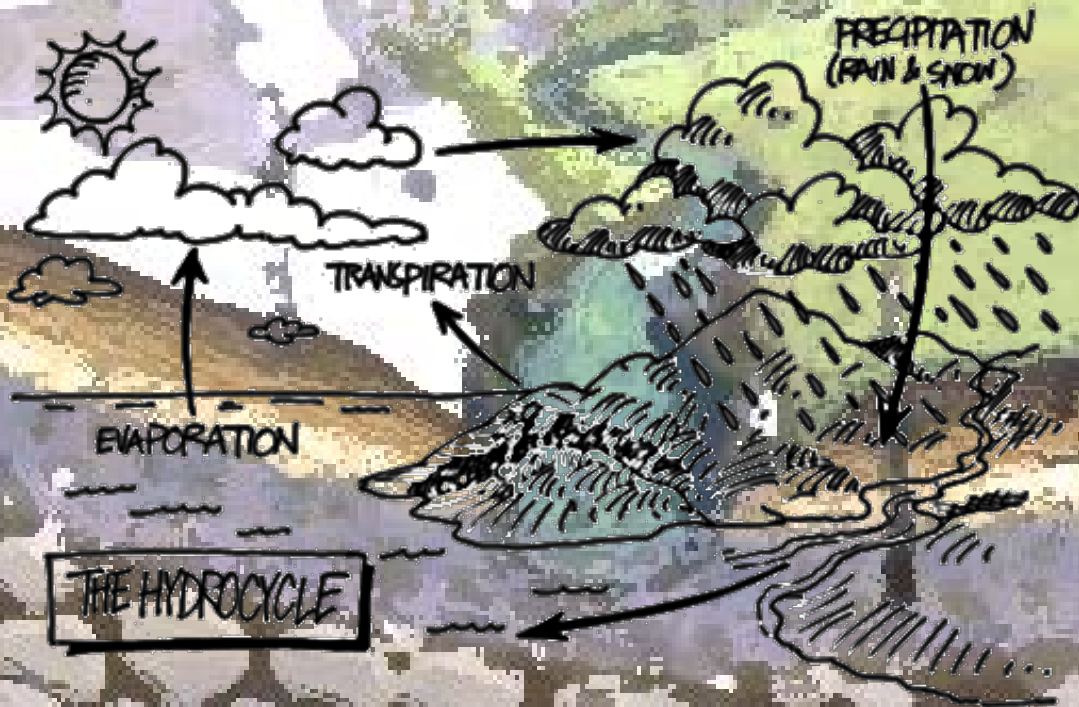


CDRD-PNPR-plus-ASCAT-AB - 12.5,42.5-1- R=0.81727



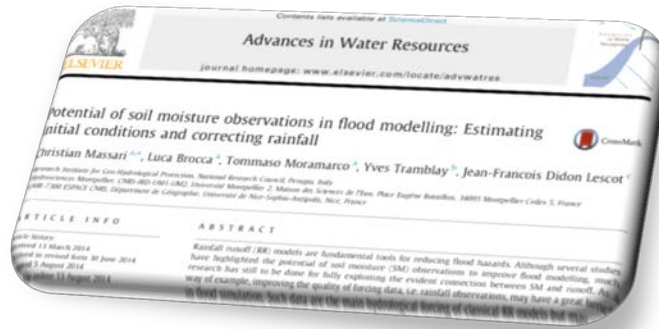
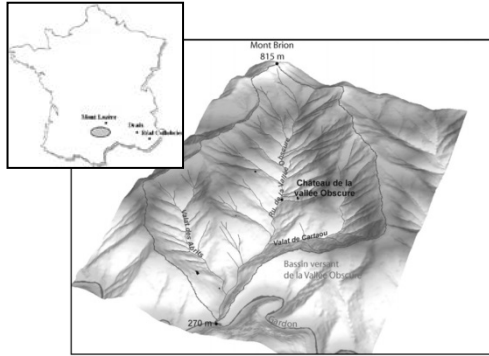
Time step: 1-day

SM2RAIN FOR IMPROVING FLOOD MODELLING



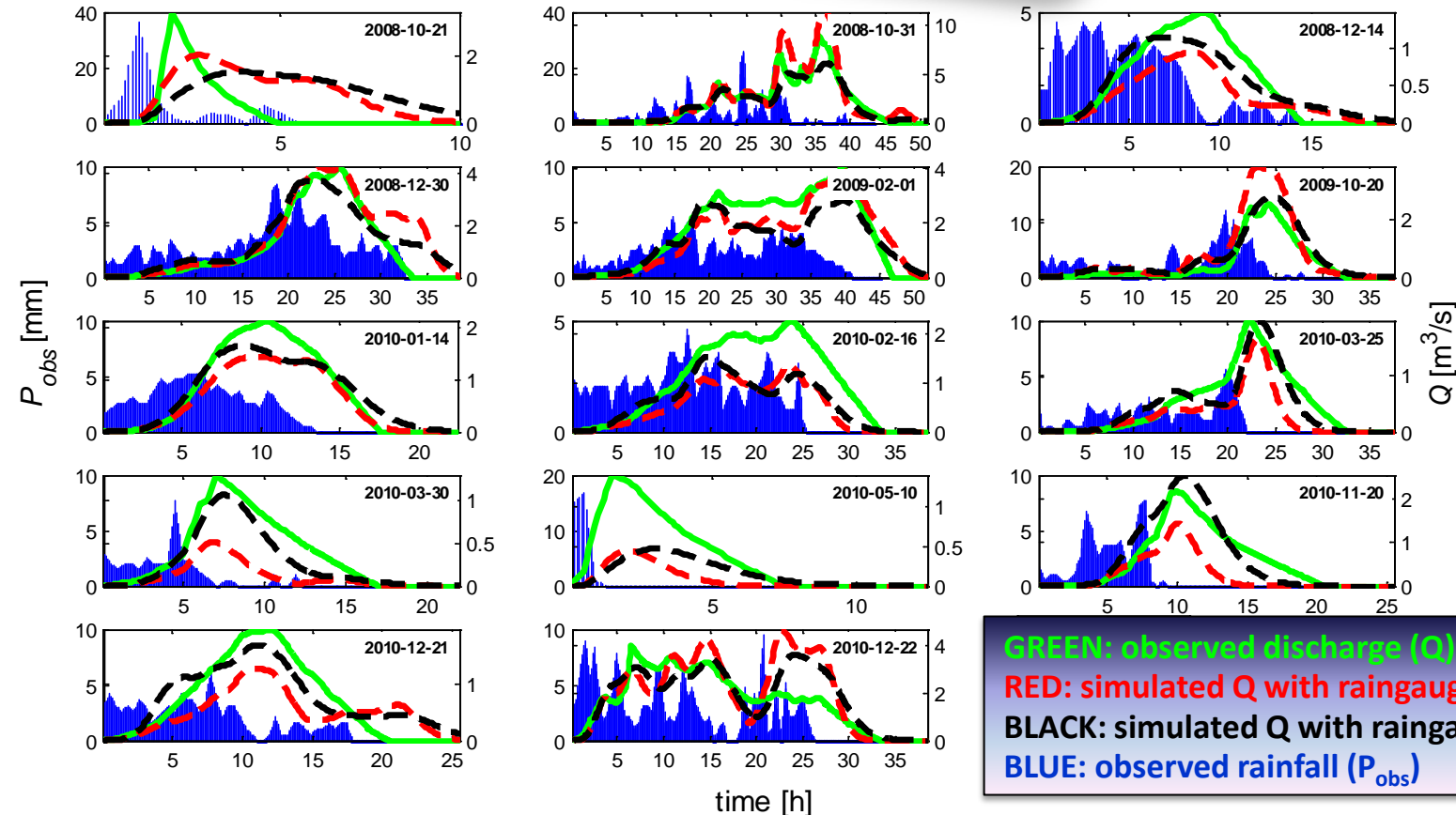
VALESCURE CATCHMENT (France)

MOISST 2015
Stillwater
2nd June 2015
Brocca Luca



IN SITU OBSERVATIONS

We obtained an increase in performance from NS=0.49 to NS=0.83 when SM2RAIN-derived rainfall is used for correcting rainfall observations during the flood event



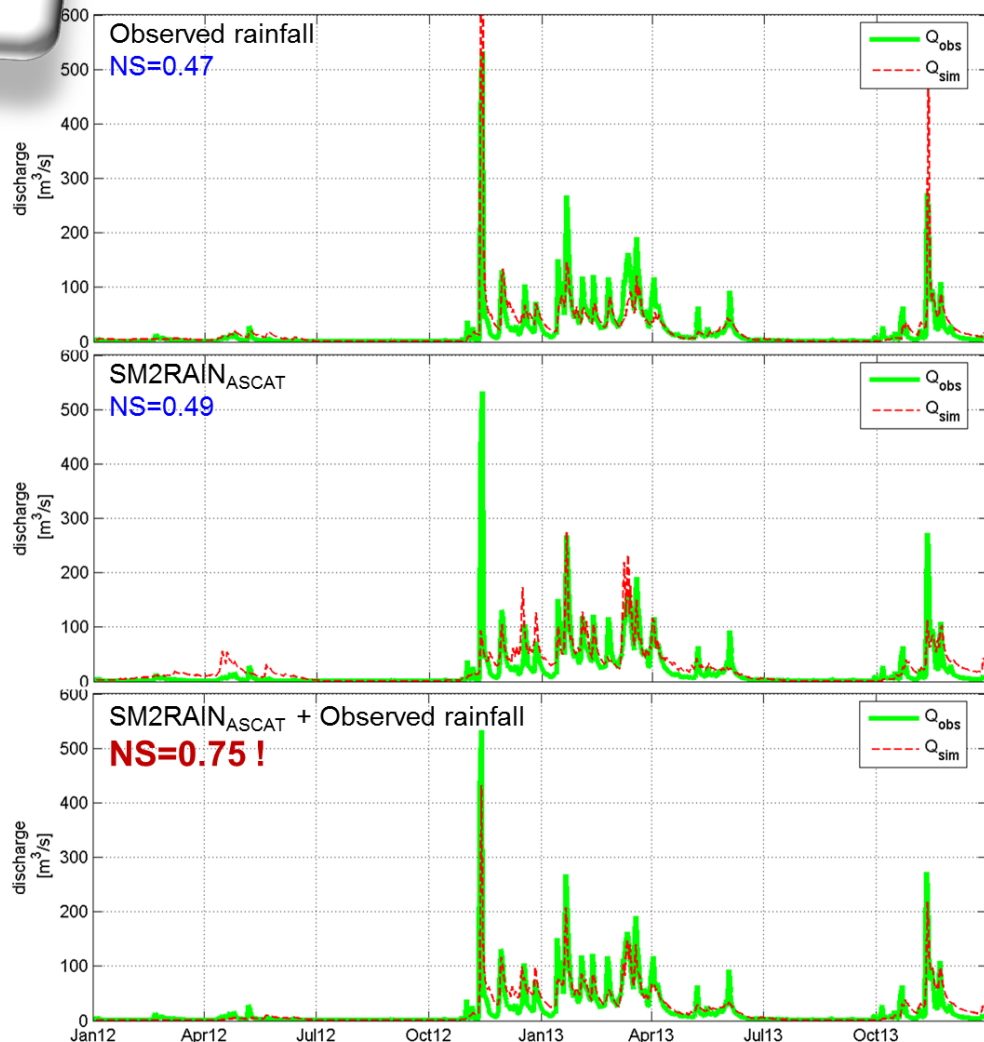
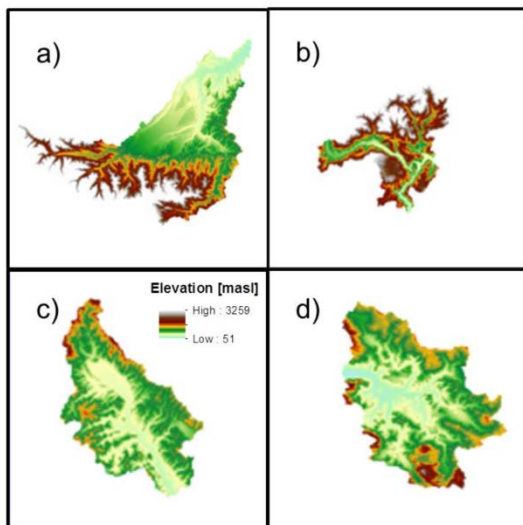
Enclosed you find the following manuscript for
Journal of Applied Earth Observation and Geoin

Submitted for publication

RAINFALL-RUNOFF MODELLING BY USING SM2RAIN-DERIVED AND STATE-OF-THE-ART SATELLITE RAINFALL PRODUCTS OVER ITALY

by L. Ciabatta, L. Brocca, C. Massari, T. Moramarco, S. Gabellani, S. Puca, W. Wagner.

SATELLITE OBSERVATIONS



Runoff simulation by using as input

1. Observed rainfall
2. SM2RAINASCAT
3. SM2RAINASCAT + Observed Rainfall

CONCLUSIONS

- ✚ **Compelling evidence that it is possible to estimate rainfall from satellite soil moisture data**
- ✚ **The soil moisture derived products are found to **estimate accurately the accumulated rainfall** addressing the difficulties of estimating light rainfall from state-of-the-art products**
- ✚ **Merging multiple satellite soil moisture products allows to obtain **good performance also at daily time scale****
- ✚ **The integration of the SM-derived products **with state-of-the-art products** provides a **significant improvement, very useful for hydrological applications****

FOR FURTHER INFORMATION

URL: <http://hydrology.irpi.cnr.it/people/l.brocca>

URL IRPI: <http://hydrology.irpi.cnr.it>

This presentation is available for download at:

<http://hydrology.irpi.cnr.it/repository/public/presentations/2015/moisst-2015-l.-brocca>