

Trends and case studies for the H SAF ASCAT root-zone soil moisture data records

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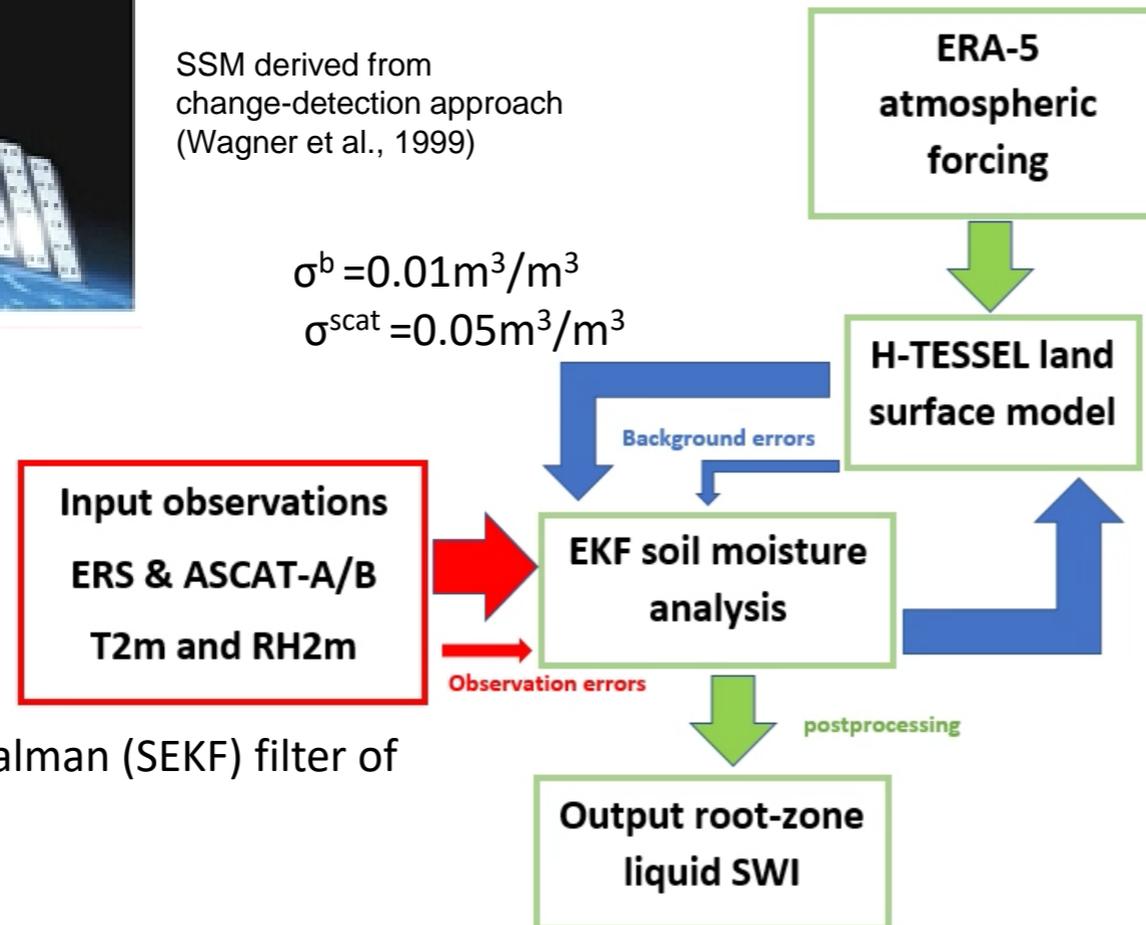
1. Overview of RZSM-DR2019-10km



SSM derived from
change-detection approach
(Wagner et al., 1999)

$$\sigma^b = 0.01 \text{m}^3/\text{m}^3$$

$$\sigma^{\text{scat}} = 0.05 \text{m}^3/\text{m}^3$$



Simplified EKF analysis

$$x^a(t_i) = x^b(t_i) + K_i [y^o(t_i) - \mathcal{H}_i(x^b)],$$

$$K_i = [B^{-1} + H_i^T R^{-1} H_i]^{-1} H_i^T R^{-1},$$

$$H_{m,i} = \frac{\mathcal{H}_{m,i}(x^b + \delta x_n^b) - \mathcal{H}_{m,i}(x^b)}{\delta x_n}.$$

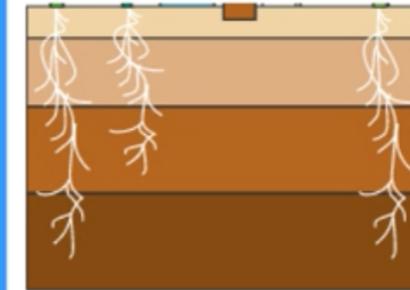
SM analysed over first 3 layers
in H-TESEL:

Layer 1: 0-7 cm

Layer 2: 7-28 cm

Layer 3: 28-100 cm

Layer 4 (not analysed): 100-289
cm



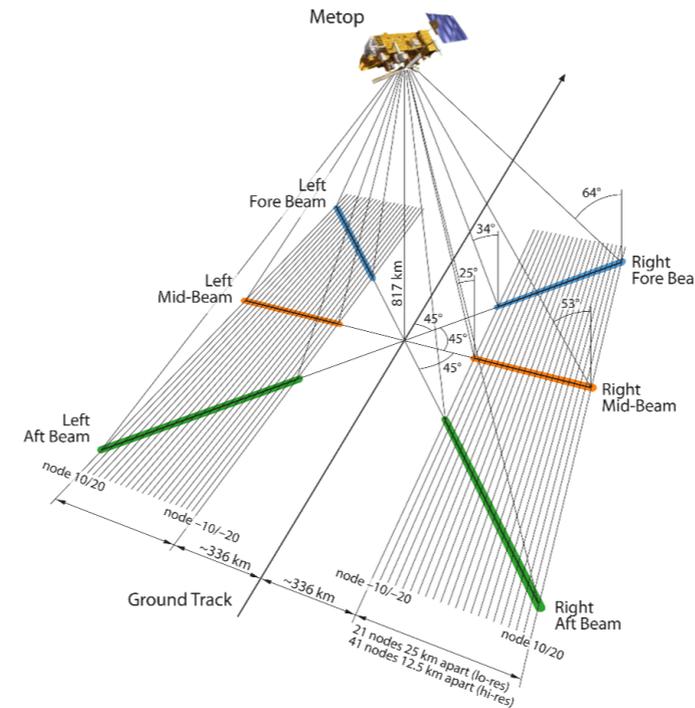
Simplified Extended Kalman (SEKF) filter of
de Rosnay et al., 2013

- Global root-zone liquid soil wetness index at 10 km sampling
- Assimilates reprocessed scatterometer-derived surface SM and screen-level T2m/RH2m
- Offline surface model forced by ERA5
- Available daily at 00 UTC and extended annually (RZSM-DR-EXT-10km, 2019-2020)

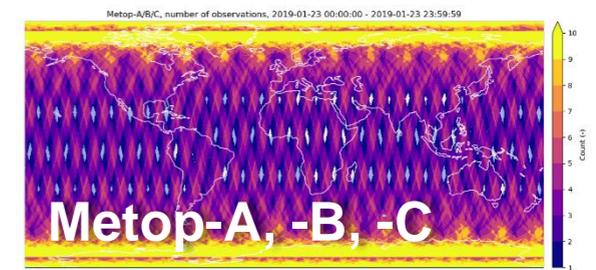
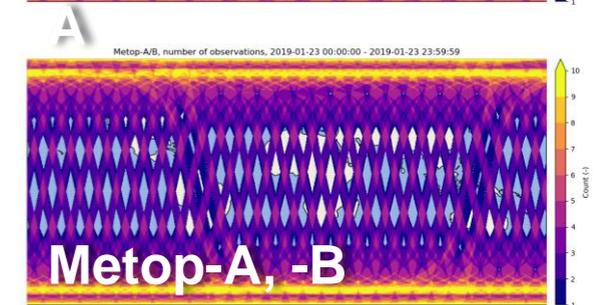
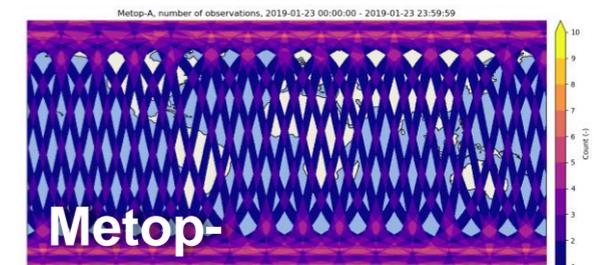
2. Advanced Scatterometer (ASCAT) on-board Metop

- Sensor characteristics

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



Spatial coverage in 24 h

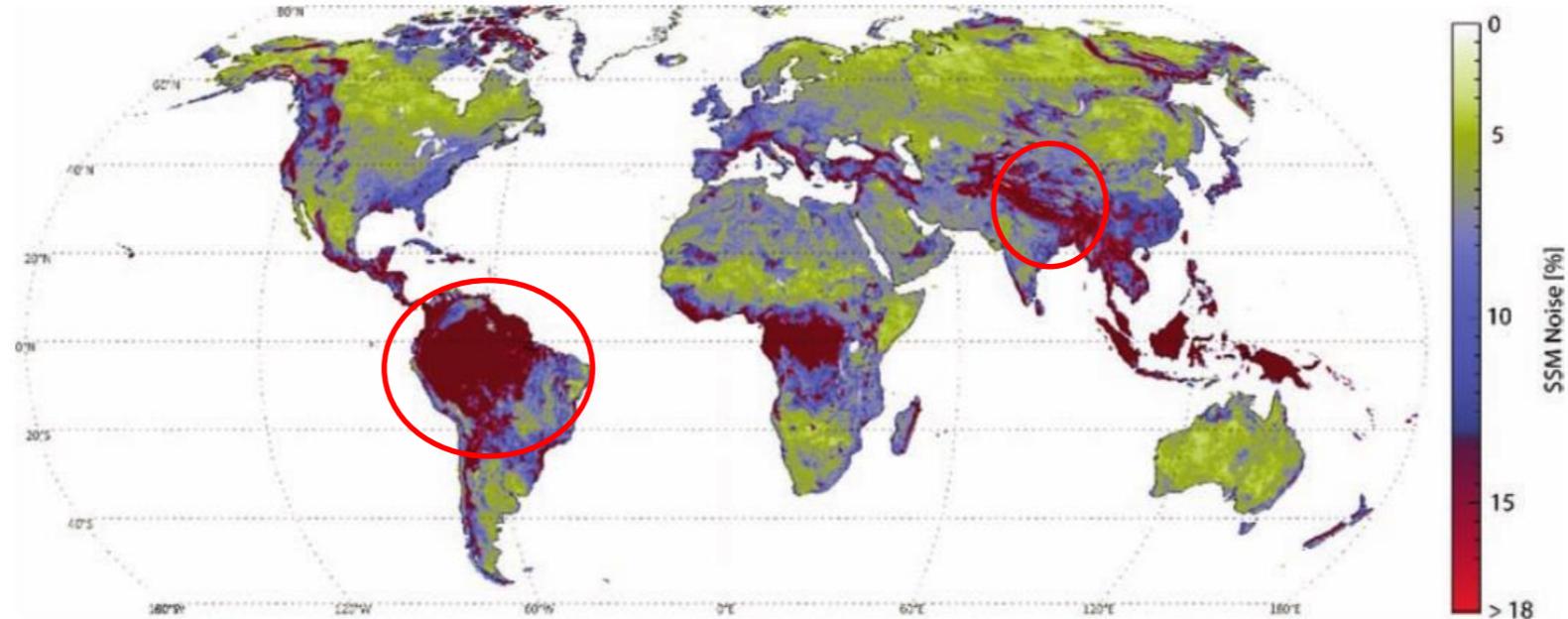


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RESEARCH UNIT
REMOTE SENSING



Figa-Saldana, et al., The advanced scatterometer (ASCAT) on the meteorological operational (MetOp) platform: A follow on for European wind scatterometers, Canadian Journal of Remote Sensing, 28(3), 404-412 (2002). <http://dx.doi.org/10.5589/m02-035>

ASCAT observation accuracy



Estimate of noise (%) in ASCAT-derived observations. From Figure 6 of Wagner et al (2013). Based upon the methods presented in Naiemi et al. (2009).

- Most areas have a high signal-to-noise ratio. But observations in highly vegetated regions and mountainous regions are noisy.

3. H-TESSSEL land surface model

- Land surface models (LSMs) provide continuous and spatially complete estimates of root-zone soil moisture.
- Dependent on accuracy of model and atmospheric forcing, notably precipitation and radiative forcing;
- LSMs require parameterizations (e.g. soil texture, vegetation type), which are not always accurate;

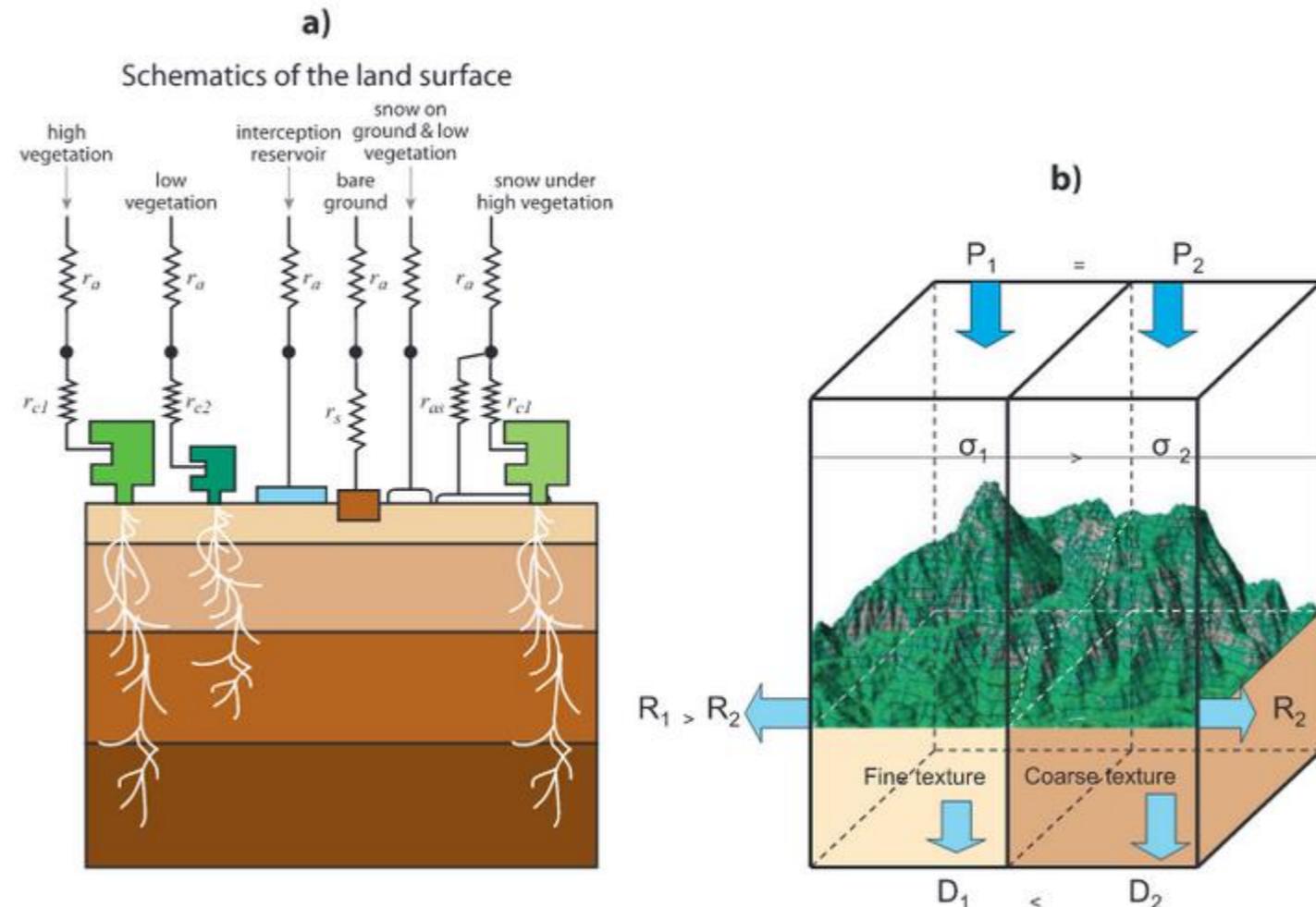


Figure 1: Schematic representation of the structure of (a) TESSEL land-surface scheme and (b) spatial structure added for H-TESSSEL. From Balsamo *et al* (2009).

4. SEKF data assimilation

Simplified EKF analysis

$$\mathbf{x}^a(t_i) = \mathbf{x}^b(t_i) + \mathbf{K}_i [\mathbf{y}^o(t_i) - \mathcal{H}_i(\mathbf{x}^b)],$$

$$\mathbf{K}_i = [\mathbf{B}^{-1} + \mathbf{H}_i^T \mathbf{R}^{-1} \mathbf{H}_i]^{-1} \mathbf{H}_i^T \mathbf{R}^{-1},$$

$$H_{mm,i} = \frac{\mathcal{H}_{m,i}(\mathbf{x}^b + \delta \mathbf{x}_n^b) - \mathcal{H}_{m,i}(\mathbf{x}^b)}{\delta x_n}.$$

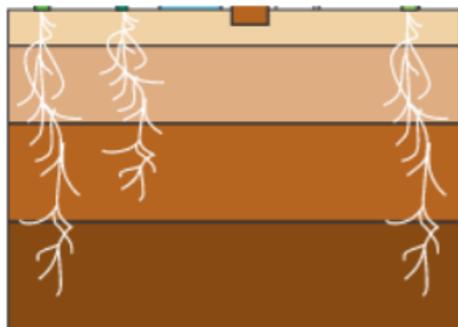
SM analysed over first 3 layers in H-TESEL:

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Layer 4 (not analysed): 100-289
cm



- SEKF based on de Rosnay et al, (2013);
- **B** is diagonal, with background-error standard deviation $0.01 \text{ m}^3\text{m}^{-3}$ for each layer;
- **R** is diagonal, with observation-error standard deviation $0.05 \text{ m}^3\text{m}^{-3}$ for ASCAT-derived SSM, 1 K for 2 m temperature and 4% for relative humidity.

SEKF data assimilation

Simplified EKF analysis

$$\mathbf{x}^a(t_i) = \mathbf{x}^b(t_i) + \mathbf{K}_i [\mathbf{y}^o(t_i) - \mathcal{H}_i(\mathbf{x}^b)],$$

$$\mathbf{K}_i = [\mathbf{B}^{-1} + \mathbf{H}_i^T \mathbf{R}^{-1} \mathbf{H}_i]^{-1} \mathbf{H}_i^T \mathbf{R}^{-1},$$

$$H_{nm,i} = \frac{\mathcal{H}_{m,i}(\mathbf{x}^b + \delta x_n) - \mathcal{H}_{m,i}(\mathbf{x}^b)}{\delta x_n}.$$

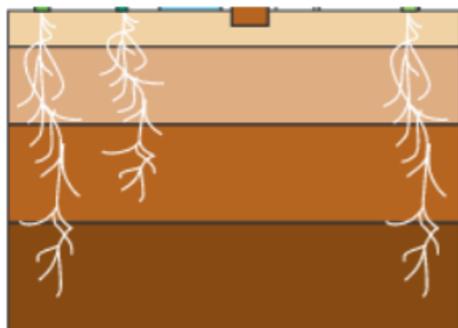
SM analysed over first 3 layers
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Layer 1: 0-7 cm

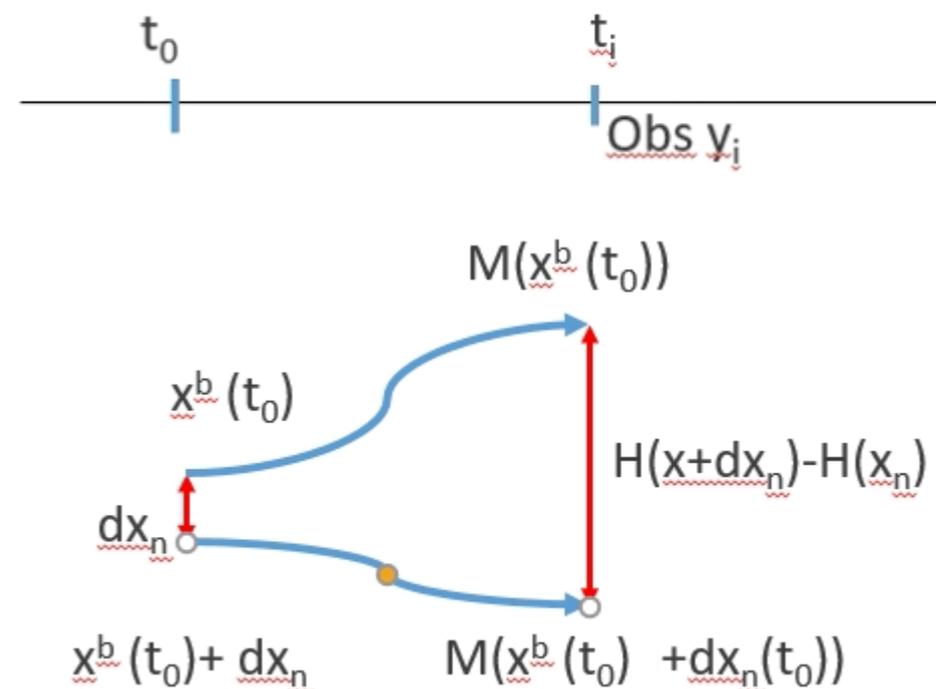
Layer 2: 7-28 cm

Layer 3: 28-100 cm

Layer 4 (not analysed): 100-289
 cm



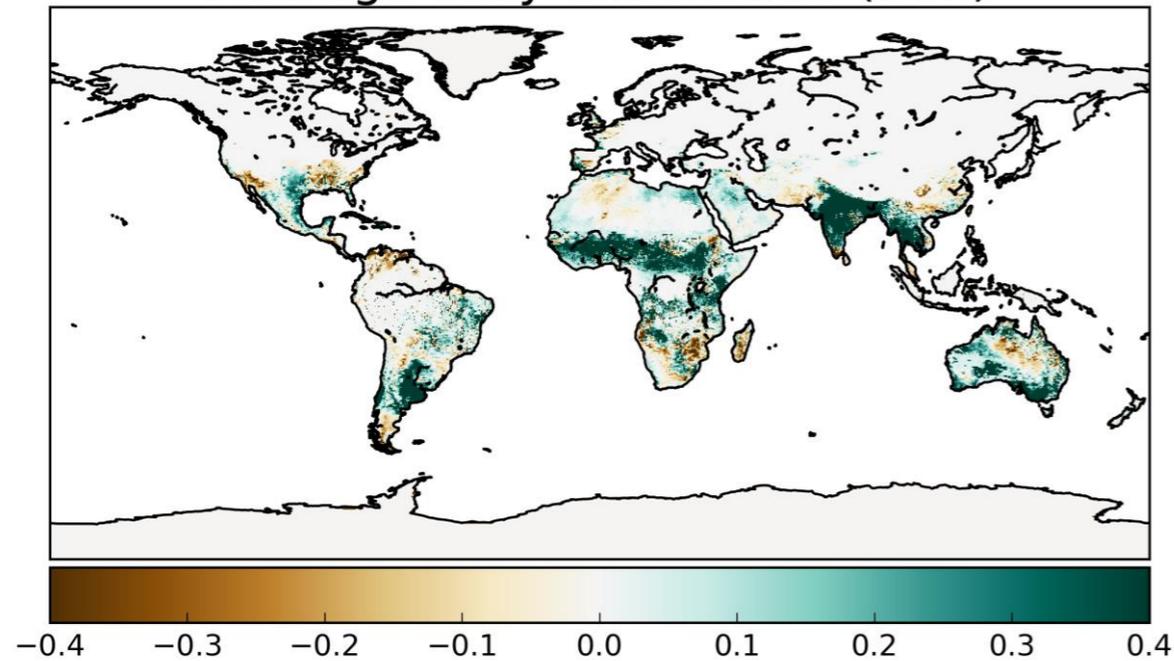
- Jacobian elements H_{nm} for analysis variable n and observation m calculated using finite differences:



Average daily RZSM increments

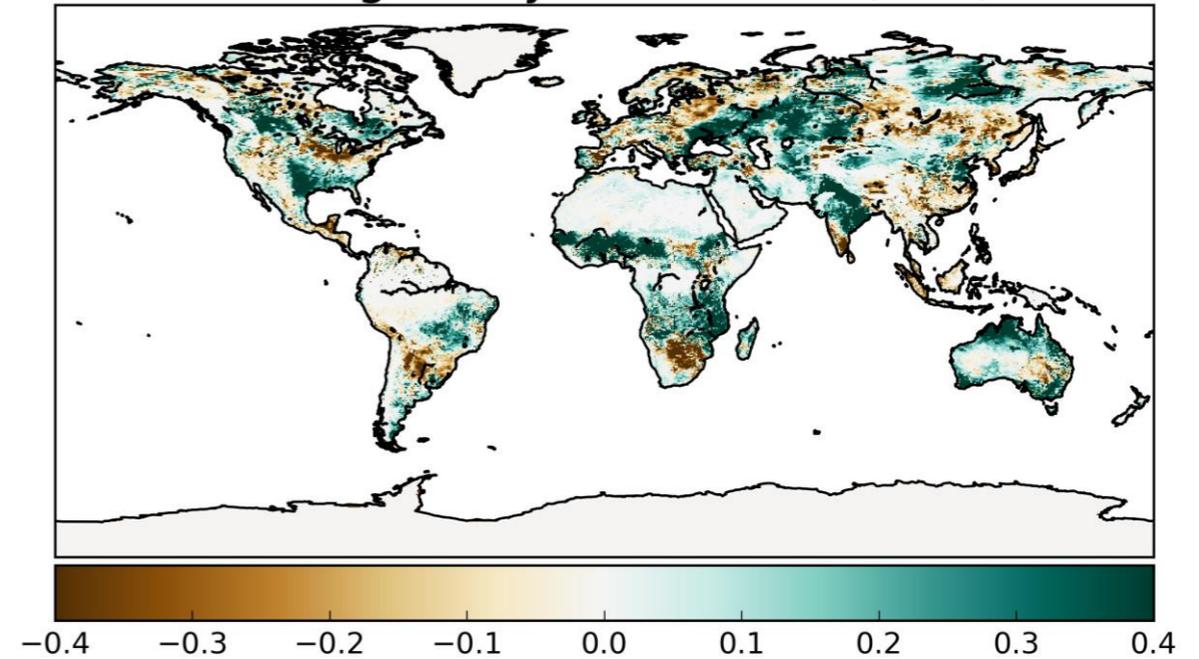
January 2009:

Average daily increments (mm)



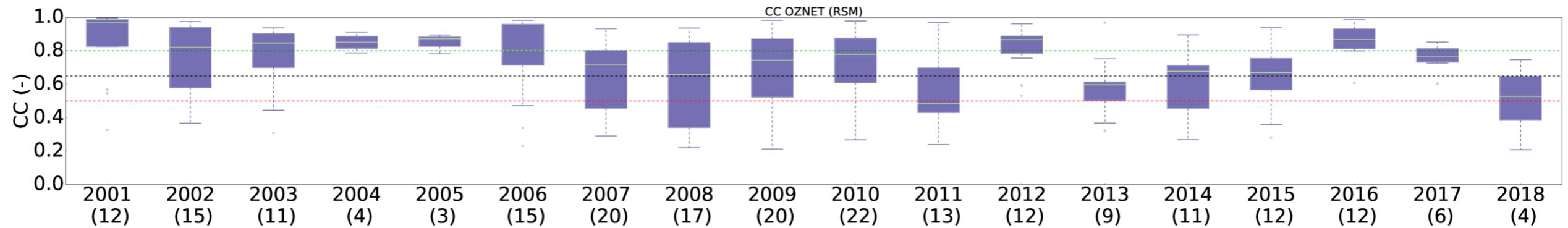
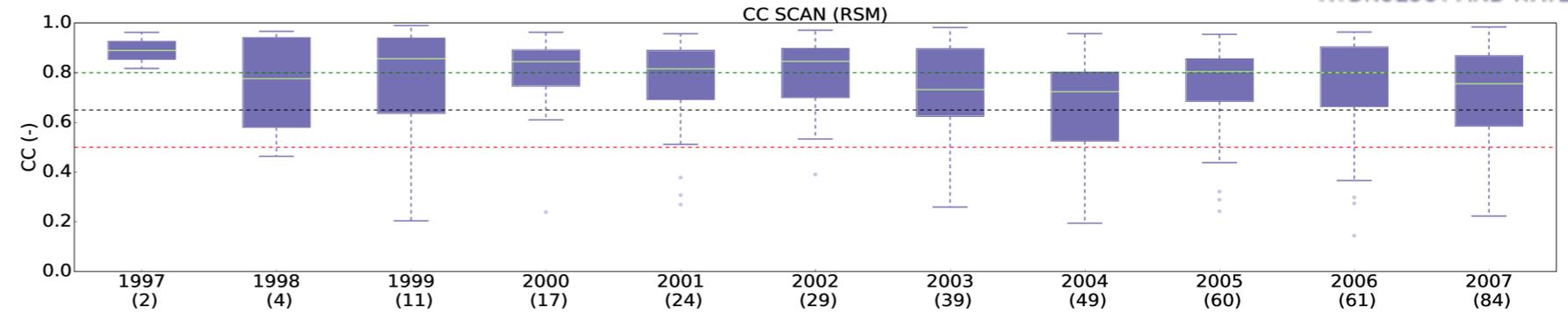
June 2009:

Average daily increments (mm)

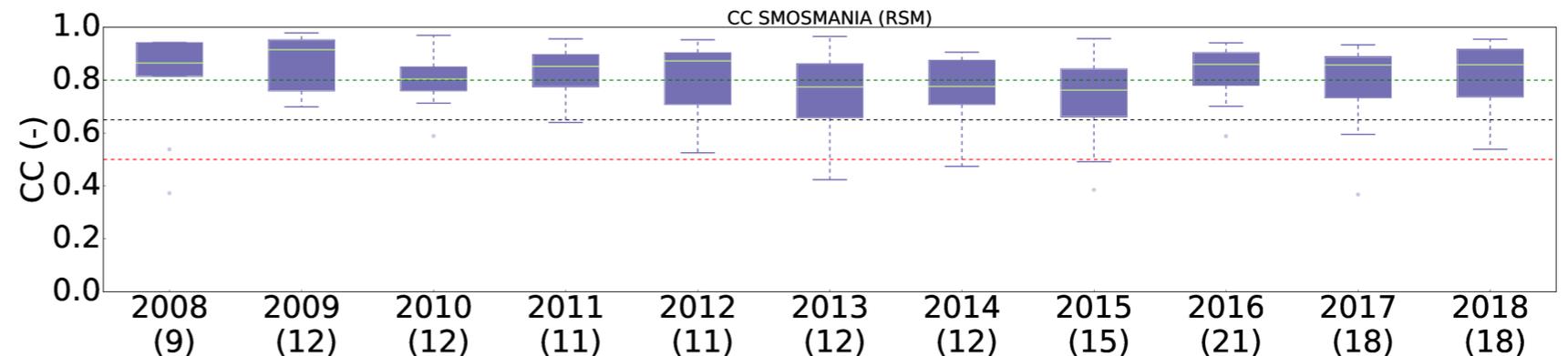


5. Data record validation

- RZSM validated annually with in situ data from networks in four countries:
 - SMOSMANIA (France)
 - USCRN/SCAN/SNOTEL (US)
 - REMEDHUS (Spain)

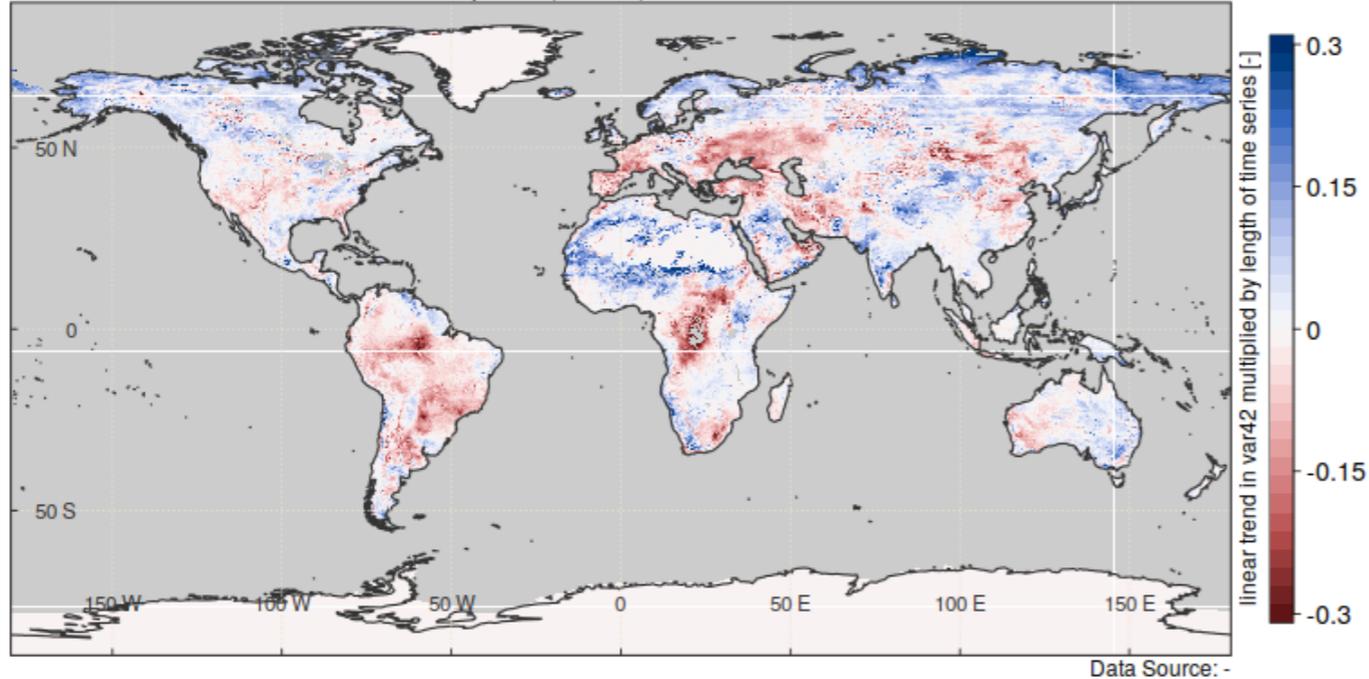


- Good overall performance, with $CC > 0.65$ over most stations

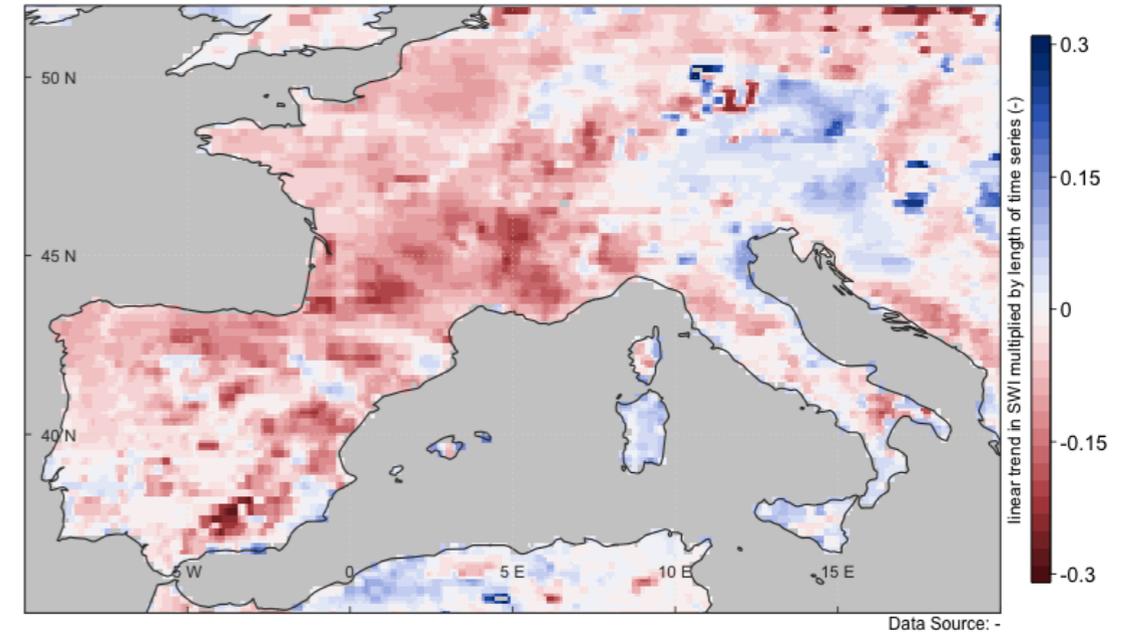


6. Soil moisture trends (1992-2020)

Linear Trend Liquid Soil Wetness Index (28-100 cm depth)
Mean of September, October, November 1992 to 2020



Autumn trend in SWI layer 3 (28-100 cm), 1992-2020

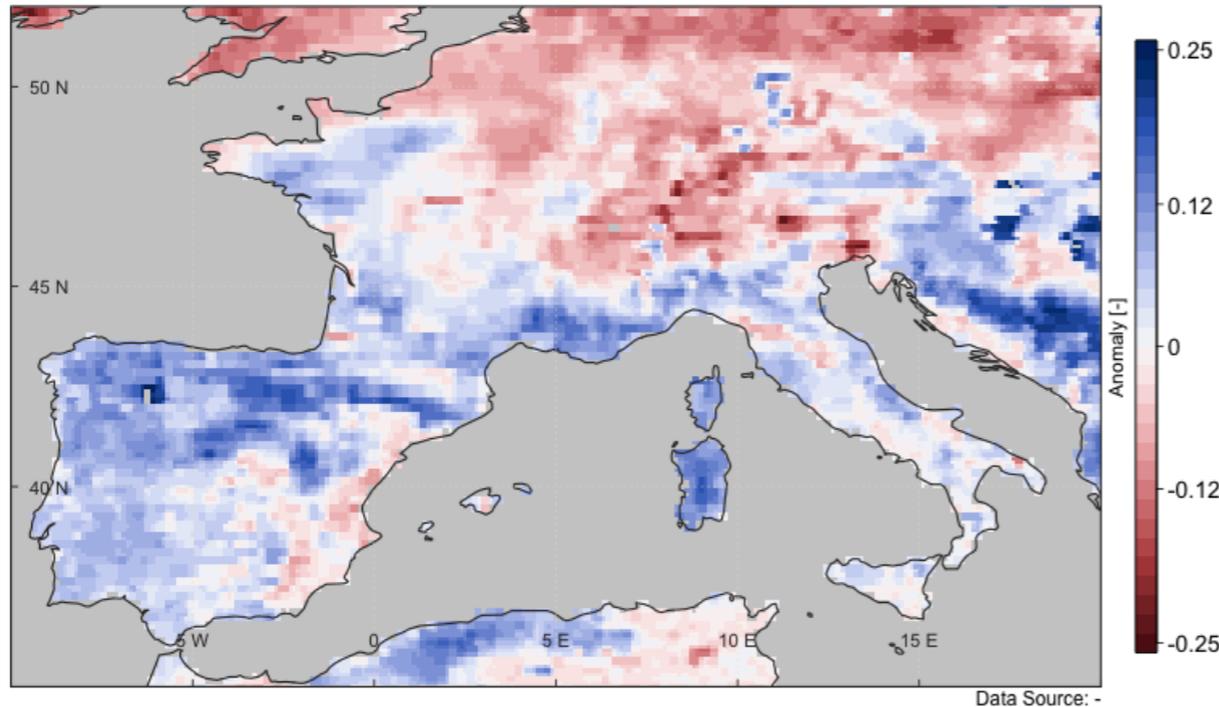


Trends calculated using CMSAF toolbox software (Kothe et al., 2019)

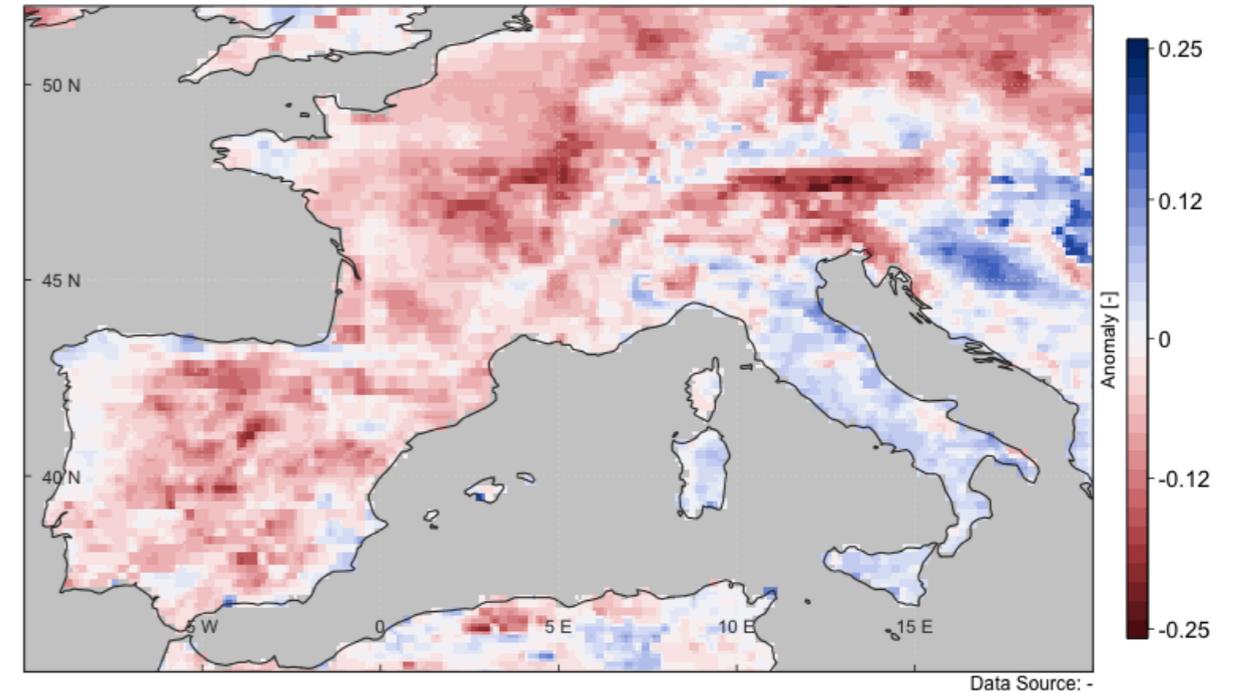
- Soil moisture has decreased by up to 30% in midlatitude autumn months, especially Europe
- Winter/spring trends are less significant
- Possible climate change mechanisms:
 - (1) Hotter summers drying soil
 - (2) Relative humidity reducing due to land warming faster than sea (Simmons et al., 2010)

Recent Soil moisture anomalies

July 2018 SWI layer 3 anomaly w.r.t. 1992-2019 mean



July 2019 layer 3 anomaly w.r.t. 1992-2019 mean

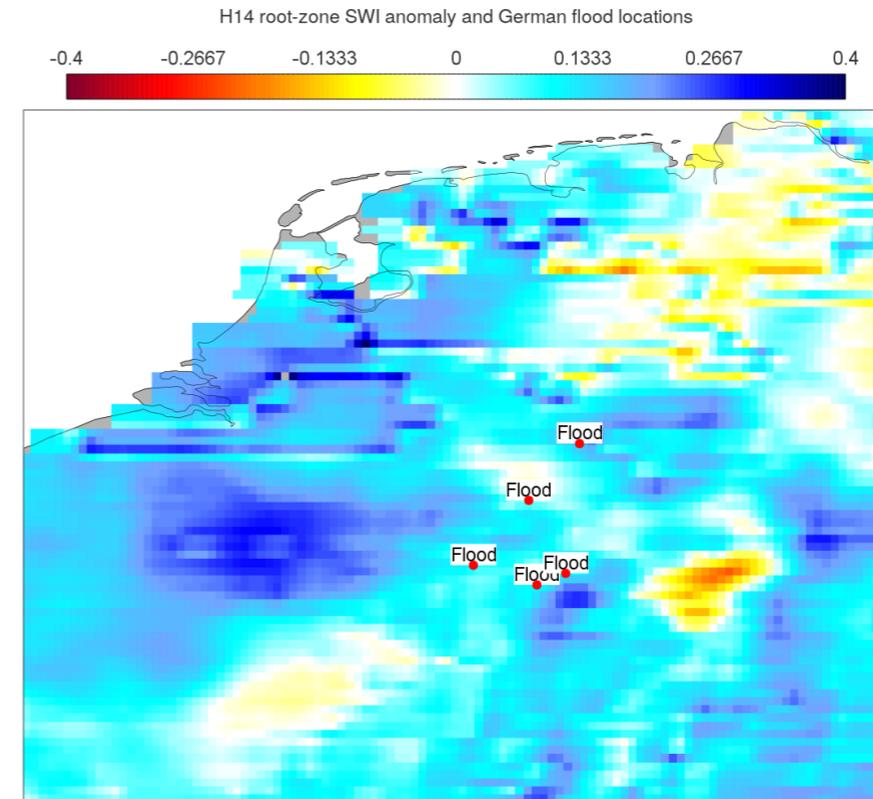
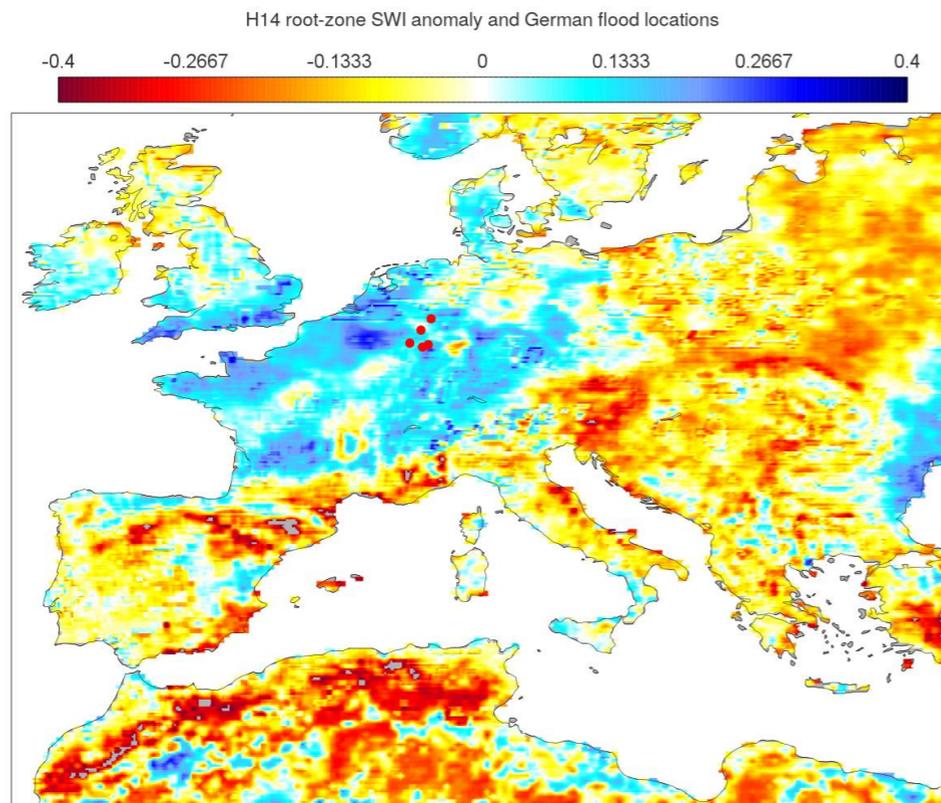


Anomalies calculated using CMSAF toolbox software (Kothe et al., 2019)

- Contrasting soil moisture conditions in July 2018 associated with dry (wet) weather over northern (southern) Europe
- Widespread dry conditions over July 2019 over Europe, when several temperature records were broken

7. German floods in July 2021

- Near-real-time daily root-zone SM product (SM-DAS-2) can be used to calculate anomalies relative to RZSM-DR2019-10km climatology with a latency of 12-36 hours
- A positive SWI anomaly was detected on 14/7/2021 at 00 UTC , approximately 24 hours before the main rainfall event that caused the flooding in northwest Germany



8. Summary

- Global H SAF root-zone soil wetness index data record (RZSM-DR2019-10km, 1992-2018) and Offline extension (RZSM-DR-EXT-10km, 2019-2020) at 10 km sampling
- Scatterometer-derived SSM and SLV data assimilation in offline LDAS forced by ERA5
- SEKF data assimilation provides dynamic link between assimilated obs and soil moisture layers
- Data record well correlated with in situ data (US, France and Spain) with average $CC > 0.65$
- Drying of root-zone SM over Europe (1992-2020) in summer/autumn probably linked to hotter summers and a reduction in relative humidity
- Near-real-time root-zone SWI detected positive SWI anomalies relative to data record climatology prior to flooding in Germany during July 2021

9. Future work

- In 2023, new RZSM data record to be released (1992-2022) with
 - recalibrated scatterometer bias correction
 - 12-hour assimilation windows (consistent with new near-real-time product).
 - Dynamic SEKF Jacobians from ensemble of data assimilations (EDA)
- EUMETSAT 2nd generation (EPS-SG) scatterometer derived surface soil moisture data to be assimilated in future data records – higher spatial resolution (12.5 km) than current ASCAT (25 km)



User information

- **To access H SAF data**, first register with H SAF:
<http://hsaf.meteoam.it/user-registration.php> to obtain username and password
- **User documentation/training:**
 - PUM, ATBD and validation reports for RZSM-DR2019-10km:
<https://confluence.ecmwf.int/display/LDAS/H+SAF>
 - Online user training course: <http://hsaf.meteoam.it/training-courses.php>
 - EUMETRAIN event week 4th – 8th November 2019: <http://eumetrain.org>

References

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