

Neural network products in land surface data assimilation

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ECMWF – UK Met Office Land DA meeting
Monday 14th December 2020

¹ ECMWF

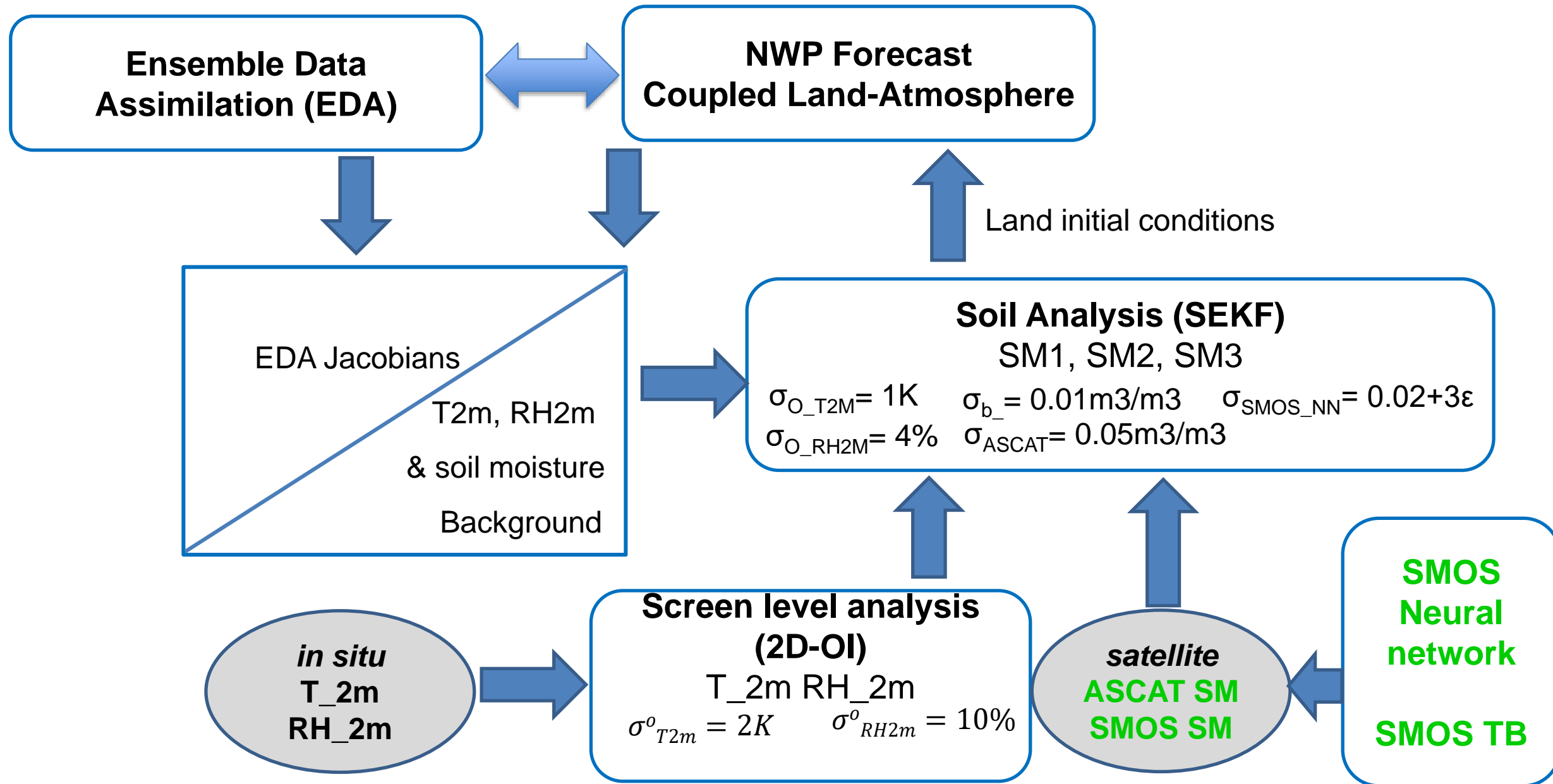
² CESBIO

³ LERMA

Contents

- Introduction and assimilation context
- SMOS neural network products
- ASCAT neural network
- Conclusions and future work

ECMWF Soil Analysis in IFS



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SMOS

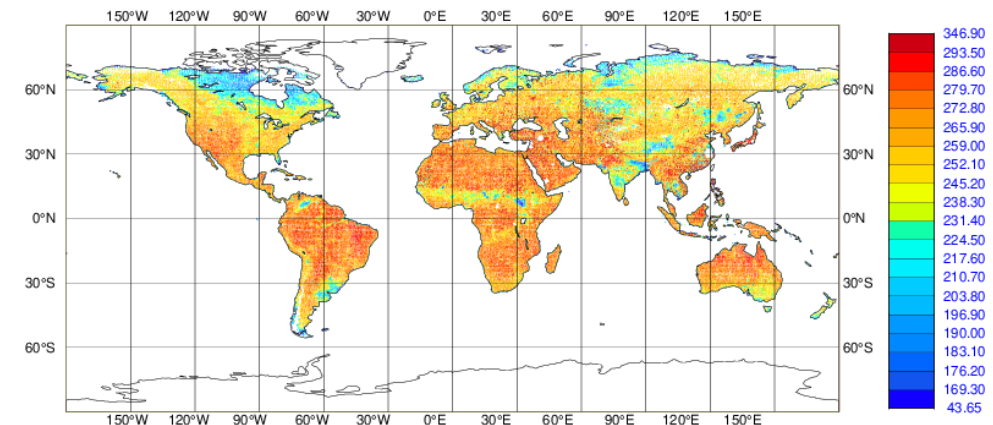
- SMOS is the soil moisture and ocean salinity mission
- ESA Earth Explorer satellite launched in 2009
- Passive microwave instrument measuring L-band (1.4 GHz) brightness temperatures
 - Sensitive to ocean salinity and surface wind speed over ocean
 - Sensitive to soil moisture over land
- ESA run a physically based retrieval to produce L2 products including soil moisture
 - There is a considerable delay due to the timeliness of auxiliary input data
- A neural network approach trained on historical L2 soil moisture can be produced much faster

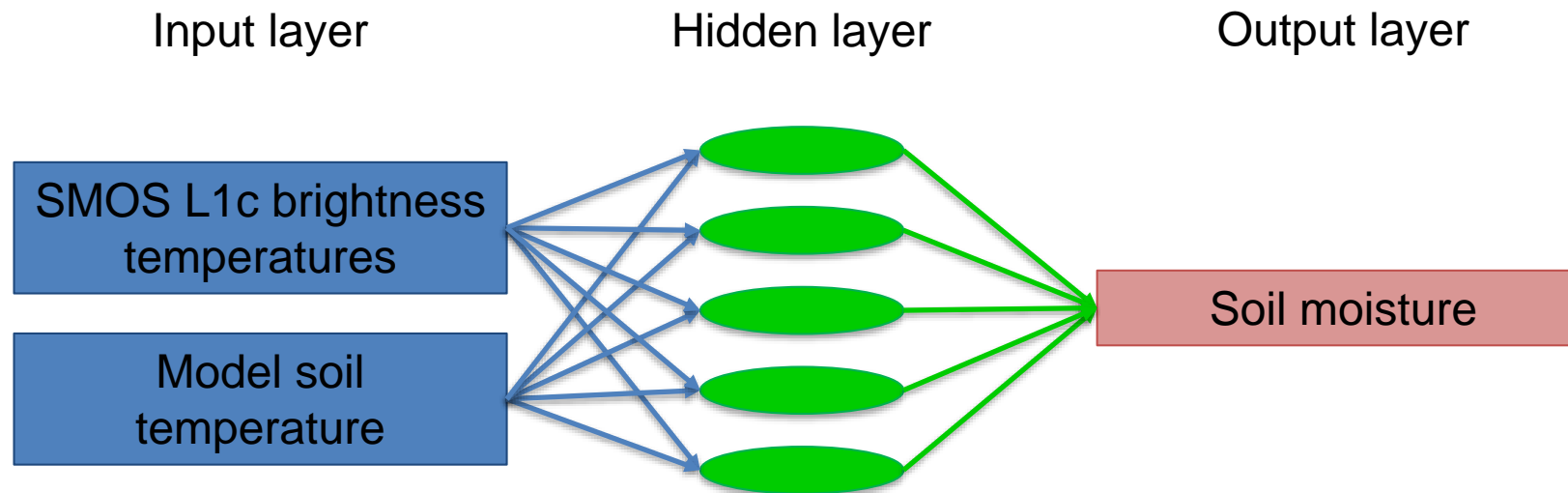


STATISTICS FOR RADIANCES FROM SMOS/SMOS
MEAN OBSERVATION VALUE (RFI SCREENED)
DATA PERIOD = 2020-08-31 21 - 2020-10-02 21
EXP = 0001, CHANNEL = 1 (FOVS: 36-45)

Min: 50.550 Max: 340.000 Mean: 251.898

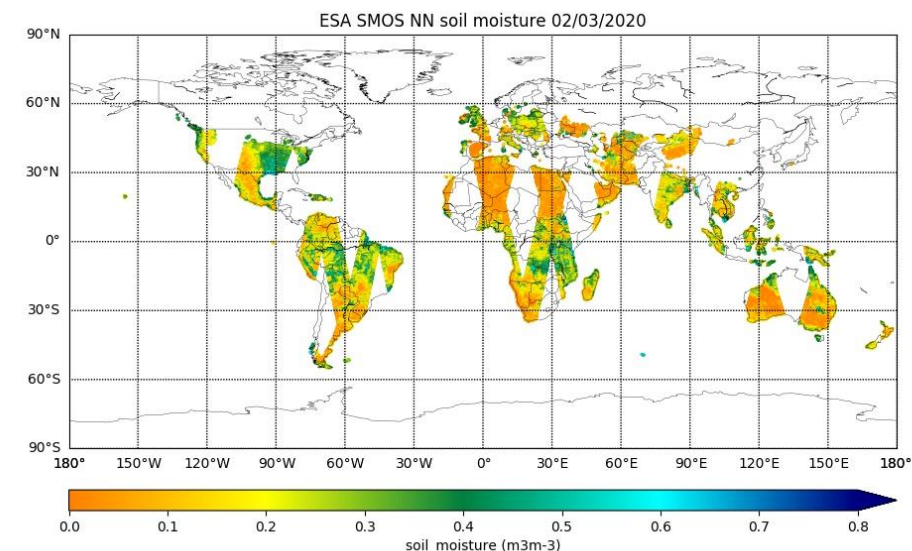
SMOS Brightness temperature (K)





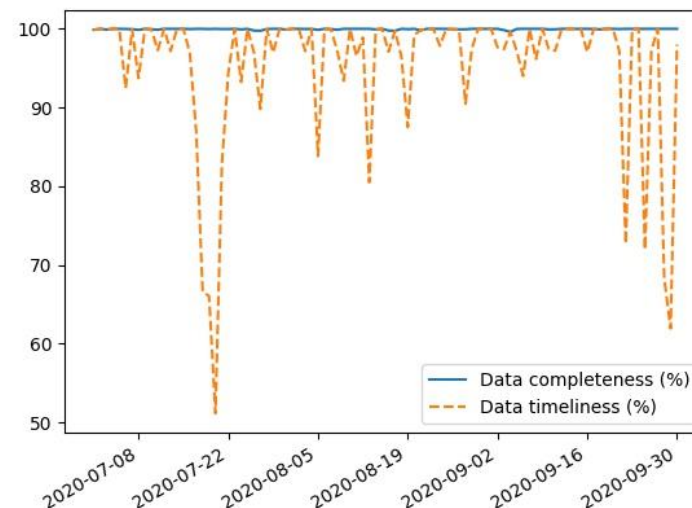
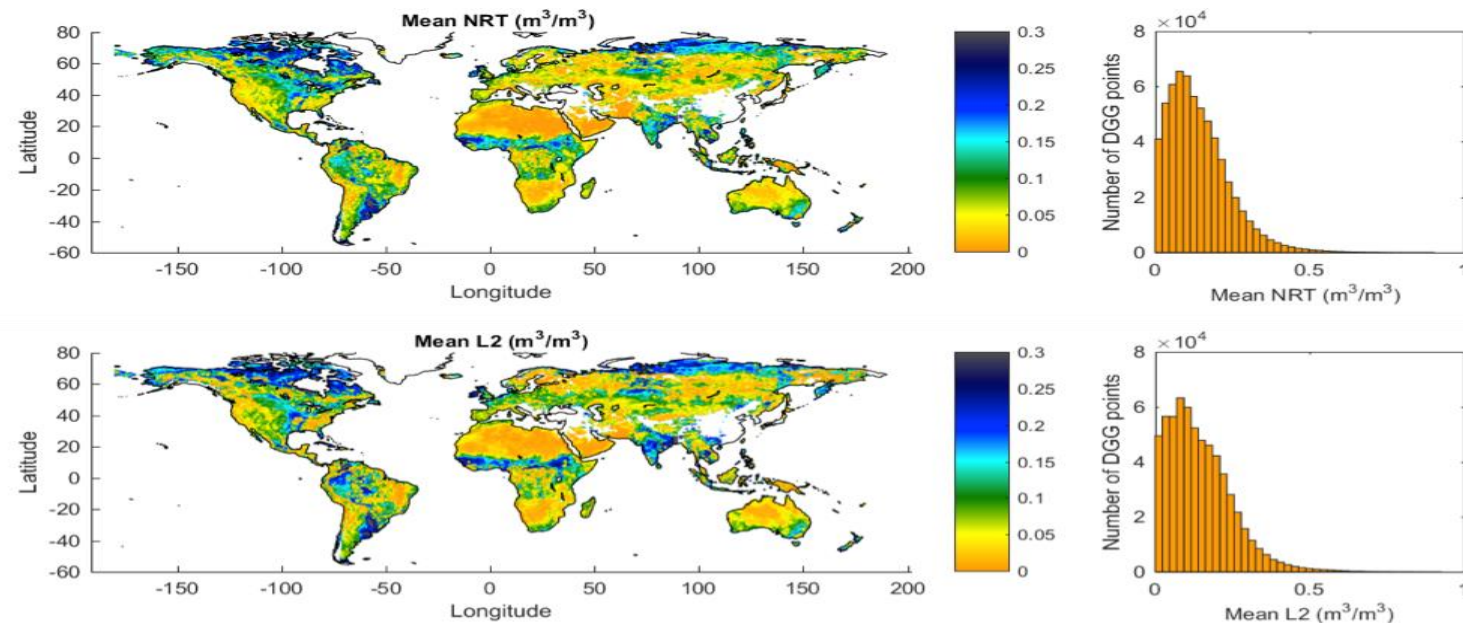
Designed by CESBIO/Estellus, Implemented by ECMWF

- Neural Network used to retrieve SMOS L2 SM:
 - Trained on SMOS L2 soil moisture
 - Single hidden layer, 5 neurons
- Product available within 30 minutes of receiving SMOS data (< 4 hours of sensing time)
- Available in NetCDF, since March 2016 on ESA SMOS Online Dissemination service <https://smos-ds-02.eo.esa.int/oads/access>



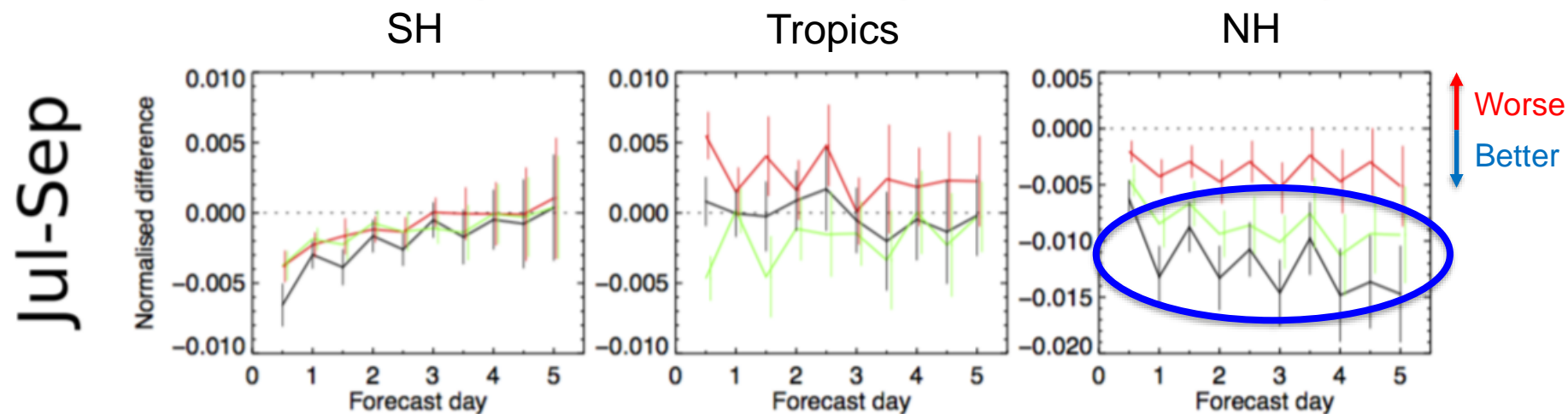
ESA level 2 SMOS NRT Soil Moisture product

- Neural network outputs agree well:
 - With L2 soil moisture
 - With in-situ stations (USCRN, SCAN)
R= 0.71
- Neural network processing run operationally and maintained at ECMWF
 - Completeness and timeliness of dissemination monitored
- Currently working with CESBIO to re-train the NN using newly re-processed SMOS data



SMOS Neural Network SM assimilation in H-TESSSEL

- First experiments assimilating a SMOS neural network product
- Offline assimilation in H-TESSSEL and initialisation of stand-alone atmospheric forecasts (2012)
- Reference H-TESSSEL with no assimilation: Open Loop (OL)
- Impact on two-meter air temperature forecasts (JAS 2012)



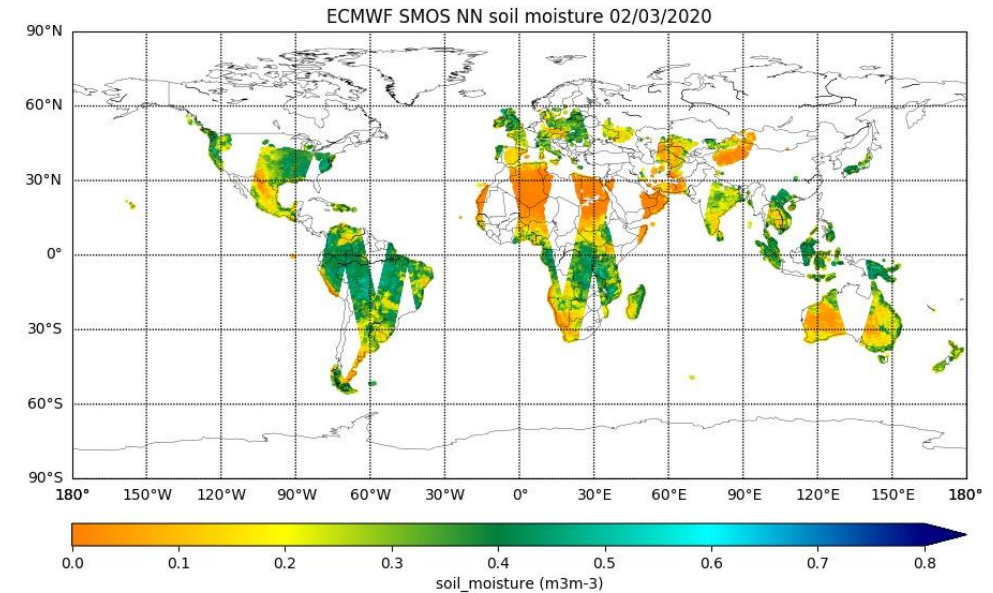
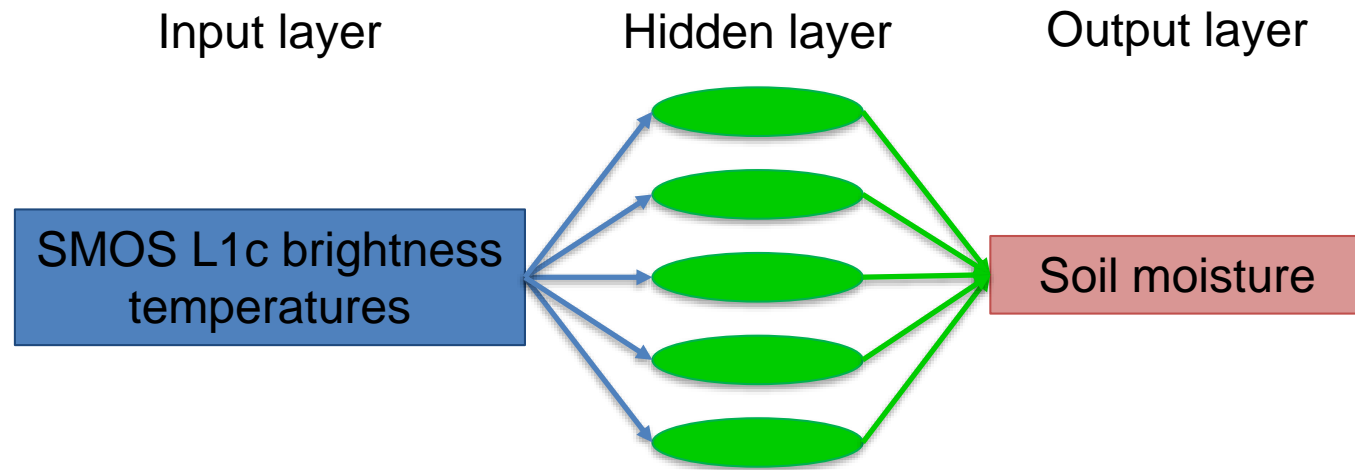
SMOS+SYNOP -OL : —
SMOS only-OL : —
SYNOP only-OL : —

→ Proof of concept of SMOS NN
assimilation for NWP initialisation

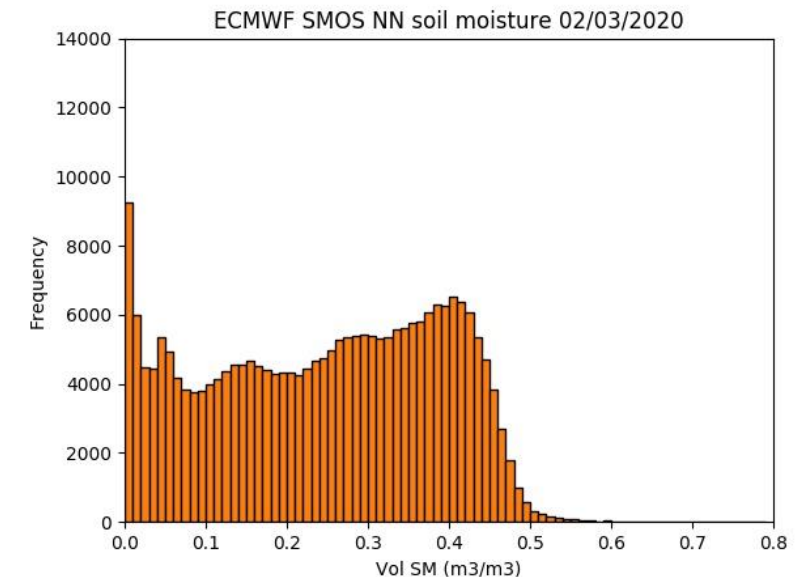
Rodriguez-Fernandez et al, Remote sensing, 2019

ECMWF level 2 SMOS Neural Network soil moisture (2018)

For data assimilation purposes



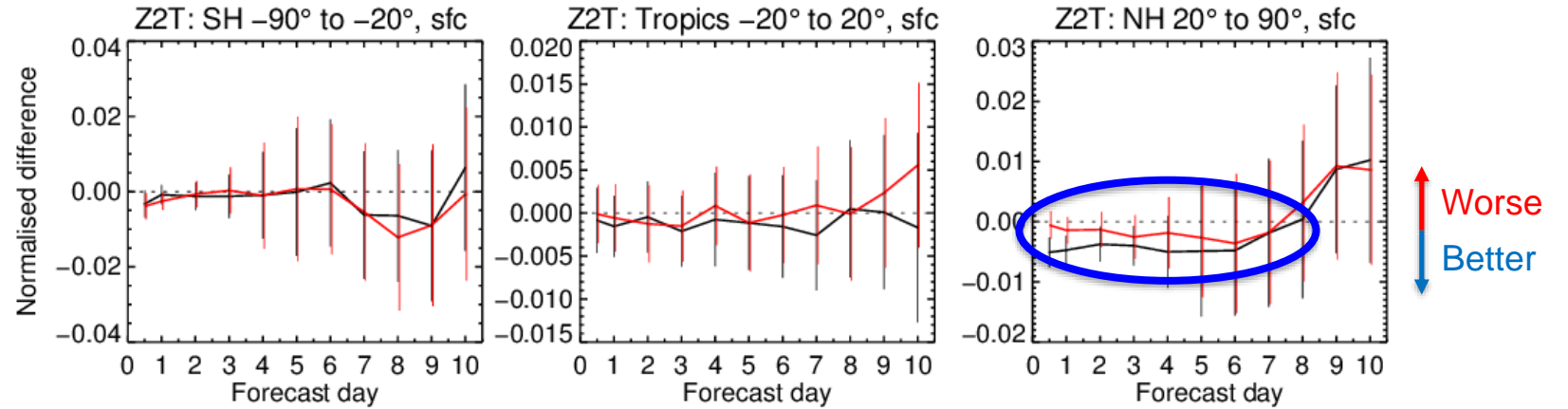
- Neural Network trained on ECMWF model soil moisture:
 - SMOS soil moisture with no bias compared to ECMWF model soil moisture
 - NRT availability
- Suitable for data assimilation
- Assimilated into the ECMWF SEKF land-surface analysis since June 2019



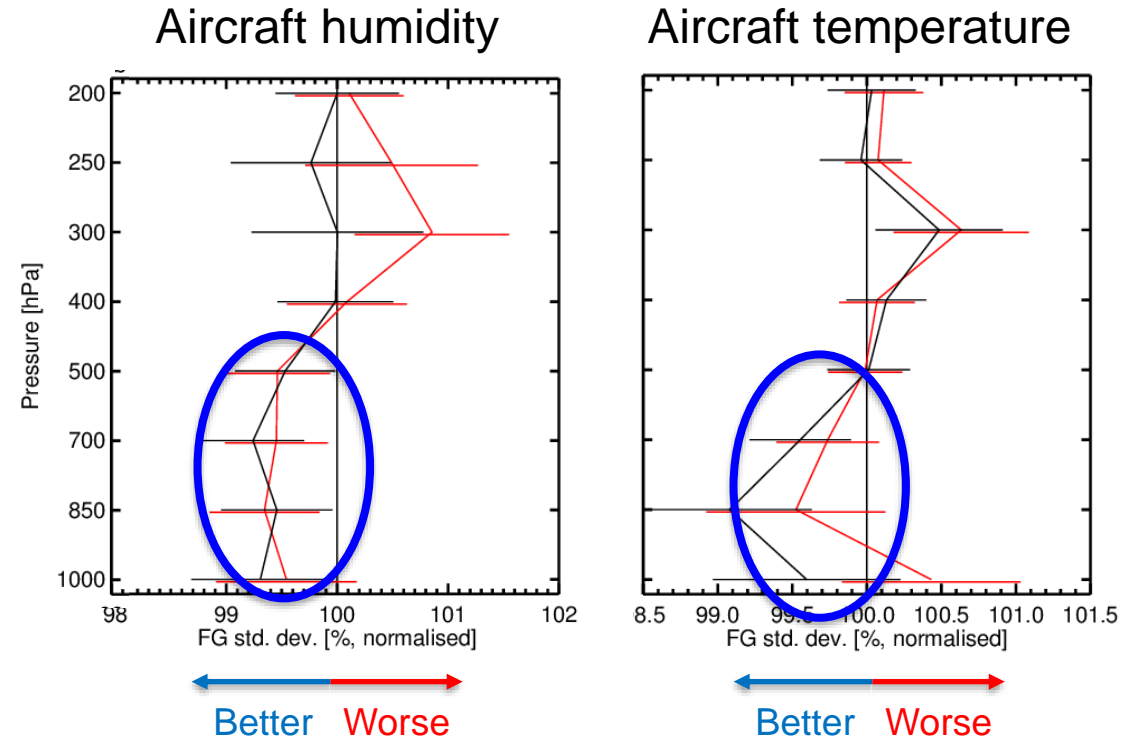
SMOS NN DA impact

1-Jun-2017 to 31-Aug-2017 from 164 to 183 samples. Verified against own-analysis.

Confidence range 95% with AR(2) inflation and Sidak correction for 8 independent tests



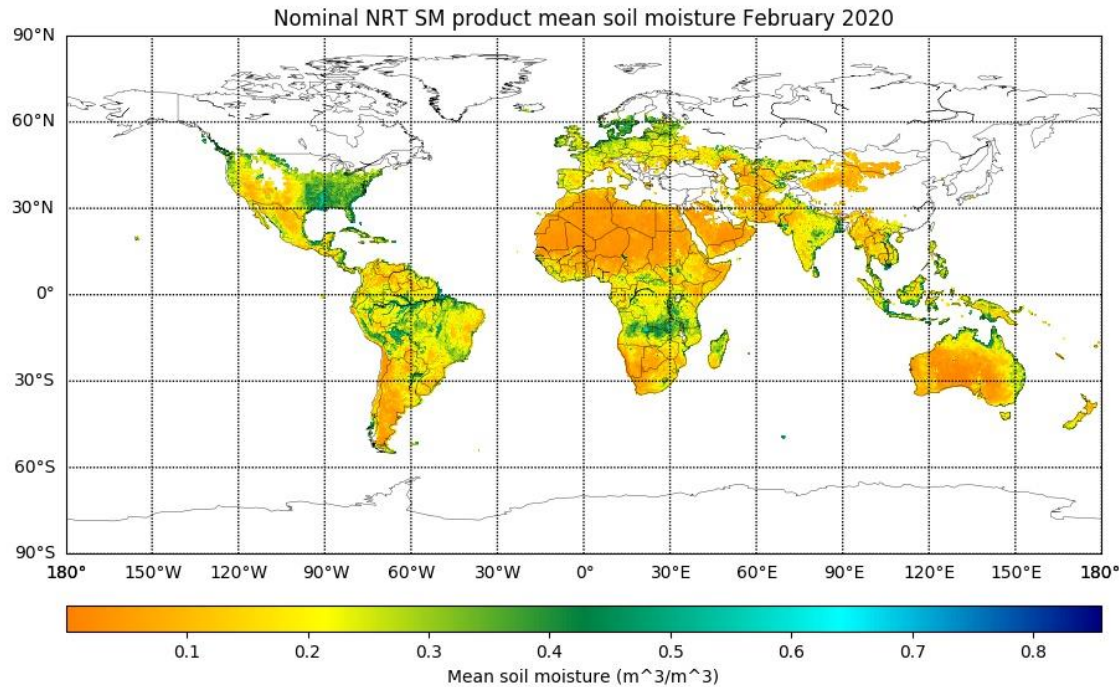
- Results from assimilating soil moisture from SMOS in coupled land-atmosphere forecasting system:
 - Neutral/slightly positive impact on T2m in the Northern hemisphere
 - Improved first-guess fits to aircraft humidity and temperature in lower troposphere



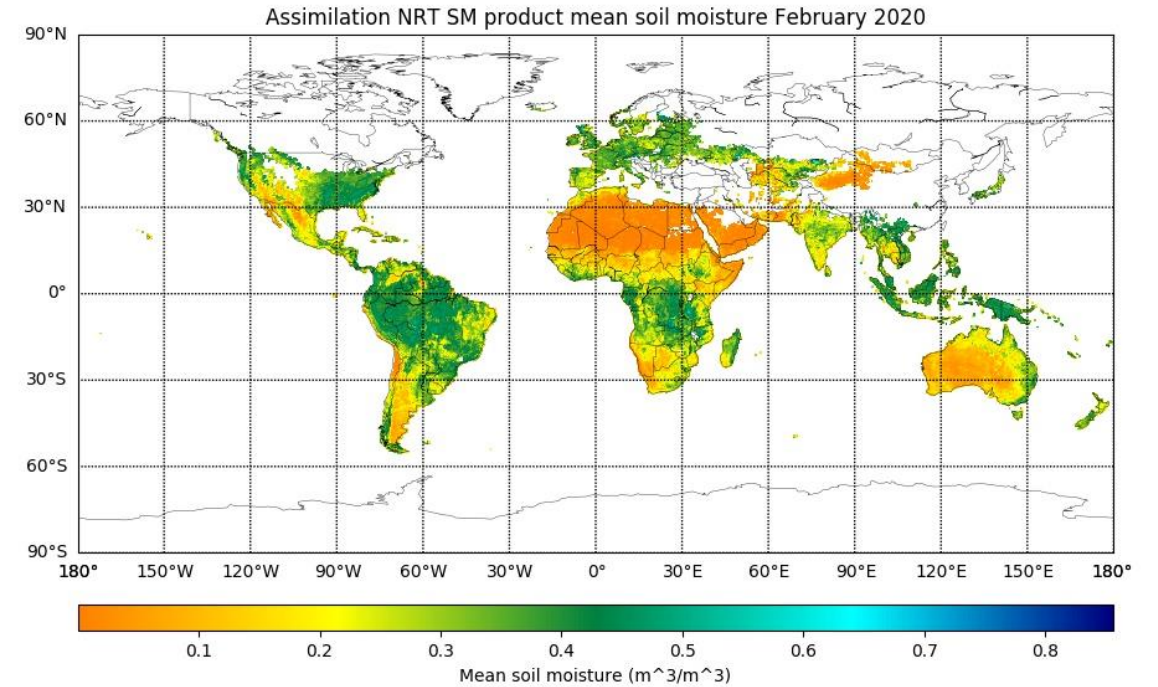
SMOS NN nominal and assimilation soil moisture comparisons

- Monthly means - February 2020
- Same observation inputs but trained on datasets with different characteristics
 - Can be used to diagnose model biases and inconsistencies between SMOS L2 and ECMWF model soil moisture

Nominal



Assimilation

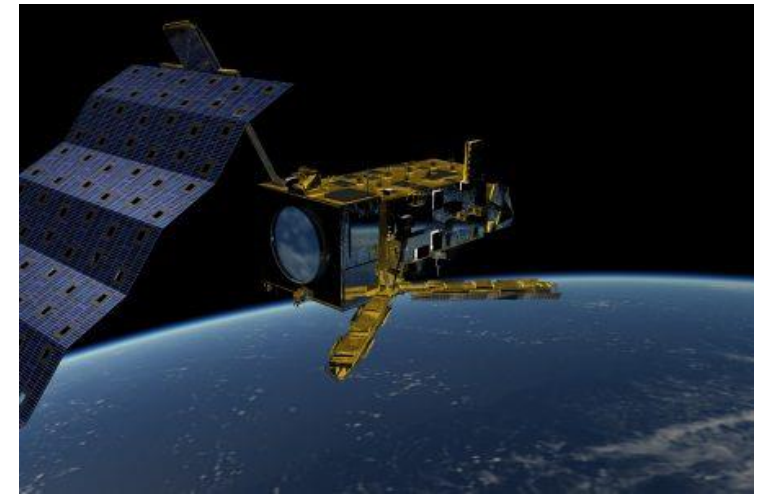


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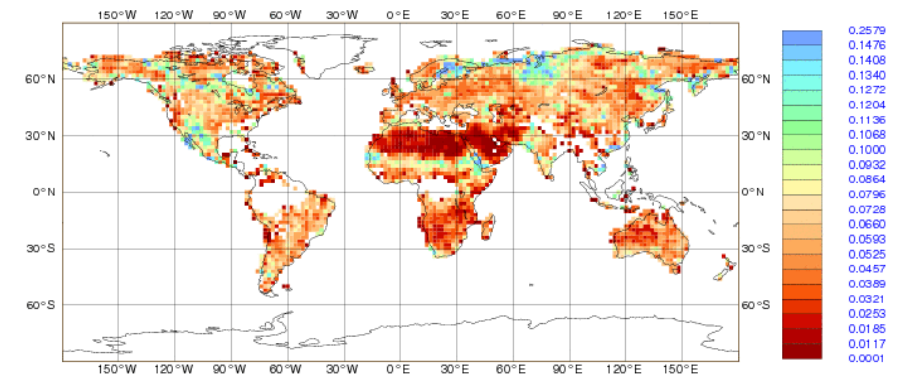
ASCAT

- ASCAT is the Advanced Scatterometer
- On MetOp-A (2006-), MetOp-B (2012-), MetOp-C (2018-)
- EUMETSAT operational satellites
- Active microwave instrument measuring C-band (5.6GHz) backscattering coefficient
 - Sensitive to surface wind-speed over ocean
 - Sensitive to root-zone soil moisture over land
- Level 2 soil moisture product assimilated into SEKF



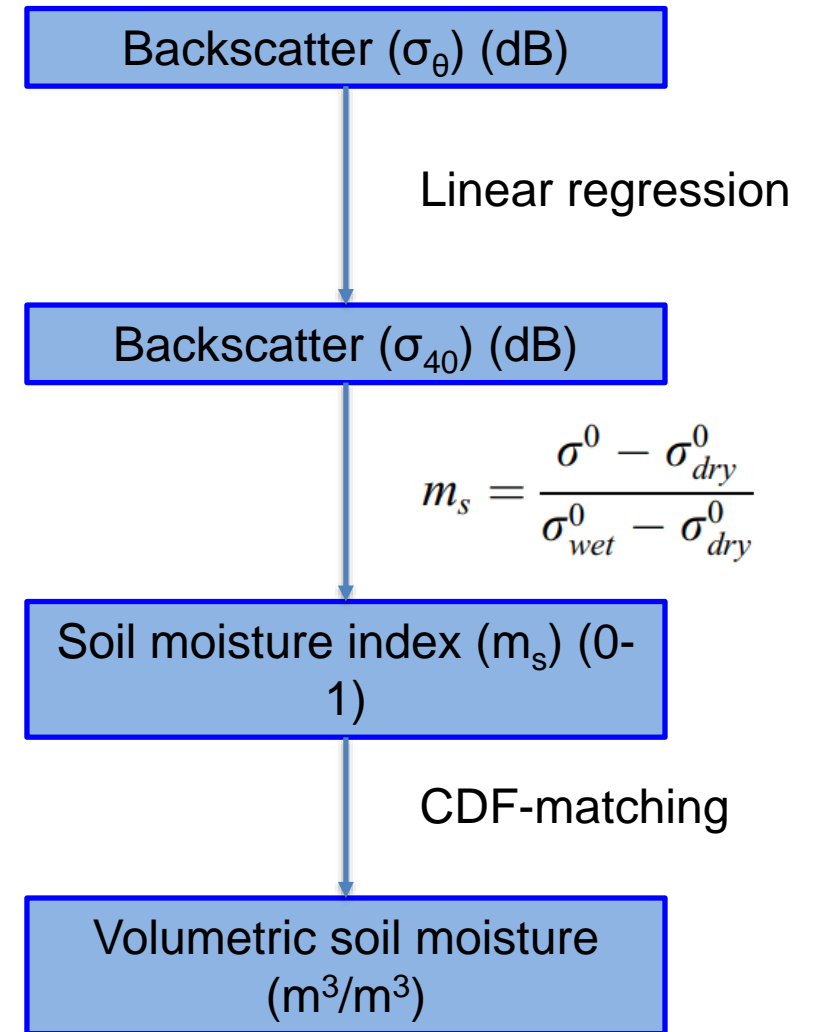
STATISTICS FOR SOIL MOISTURE FROM METOP-B/ASCAT
STDV OF FIRST GUESS DEPARTURE [M3/M3] (ALL)
DATA PERIOD = 2013-08-27 21 - 2013-09-28 09
EXP = 0001, CHANNEL = 1
Min: 0.000 Max: 0.258 Mean: 0.059

ASCAT soil moisture (m^3m^{-3})



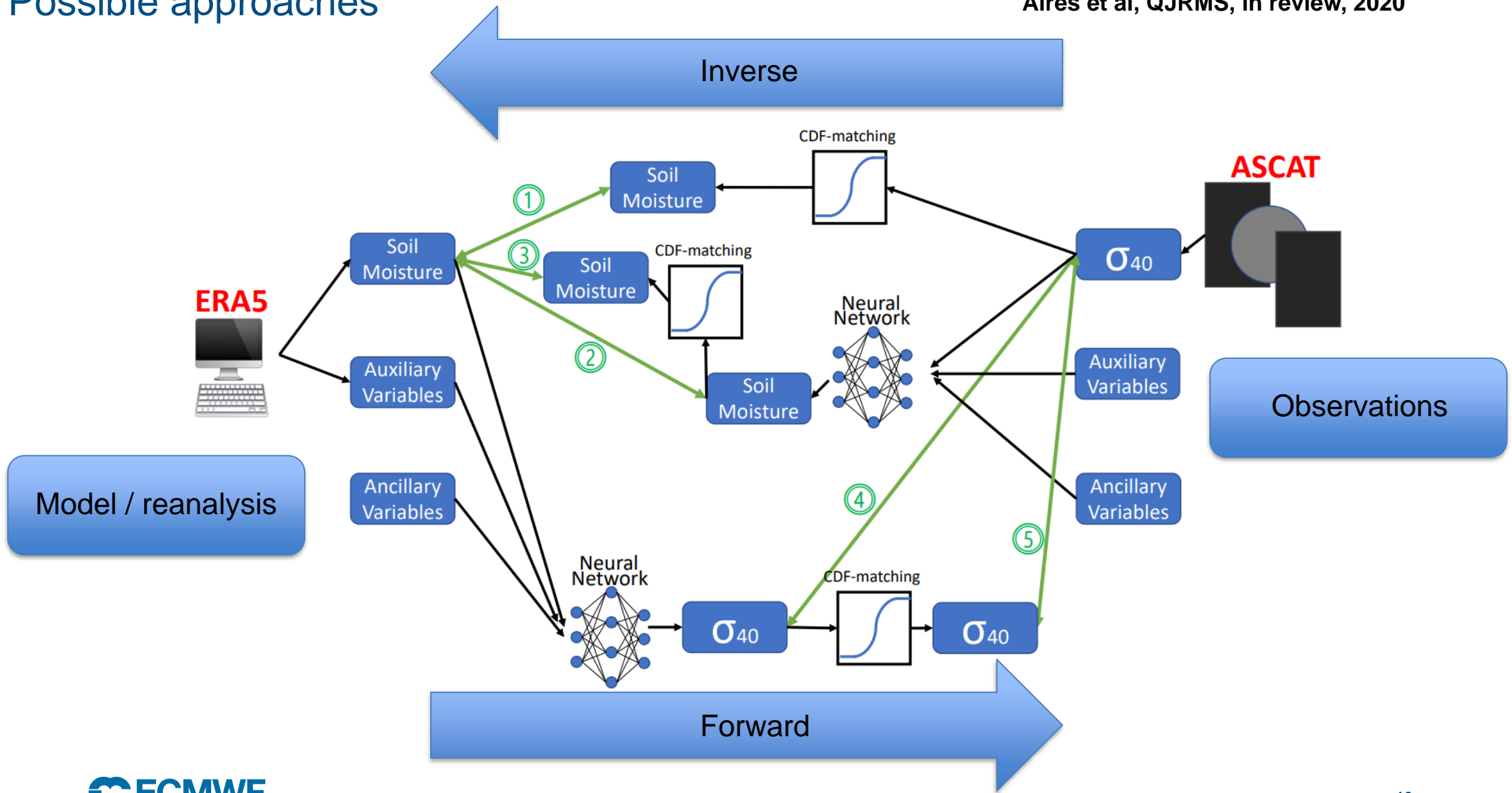
From backscatter to soil moisture

- Modern approach is to assimilate L1 measurements (backscatter) where possible:
 - Requires fast and accurate observation operator, not currently available
- Alternative approach is to retrieve soil moisture (L2 product) and assimilate this:
 - Wagner approach plus CDF-matching
 - Neural network
 - Neural network plus CDF-matching



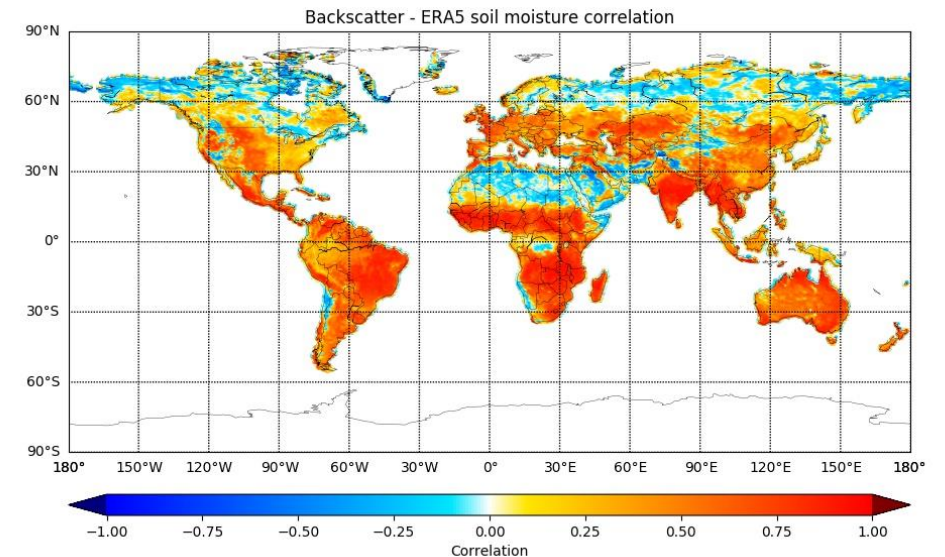
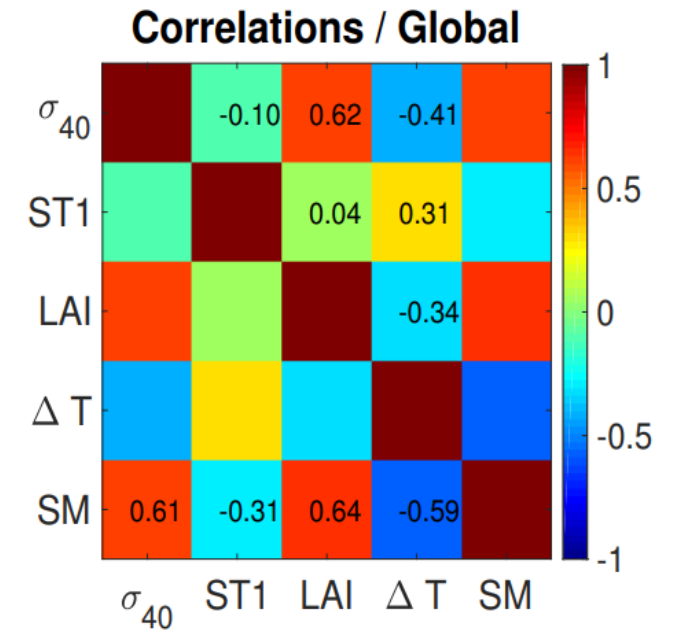
Possible approaches

Aires et al, QJRMS, in review, 2020



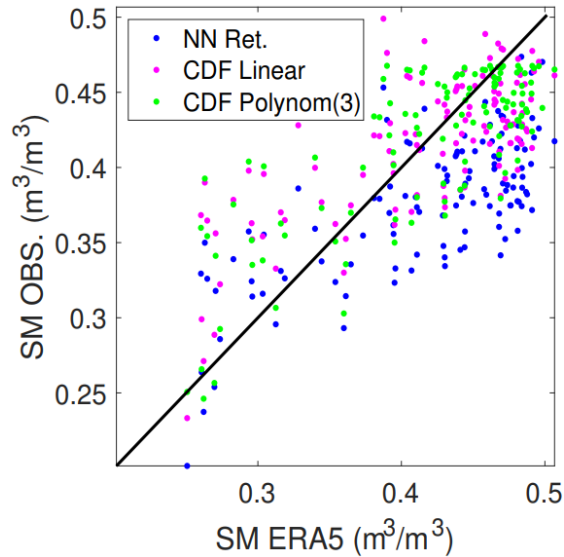
ASCAT neural network setup

- Inputs (features):
 - ASCAT backscatter at 40 degree incidence angle
 - ERA5 soil temperature in layer 1 (0-7cm)
 - ERA5 leaf area index
 - ERA5 magnitude of the diurnal cycle of 2 metre temperature
- Target:
 - ERA5 volumetric soil moisture in layer 1 (0-7cm)
- Single hidden layer with 10 neurons
- Sampling/quality control:
 - 2016-2019 (60% training, 20% testing, 20% validation)
 - Land-sea mask > 0.95
 - Avoid frozen surfaces, wetland, mountainous areas

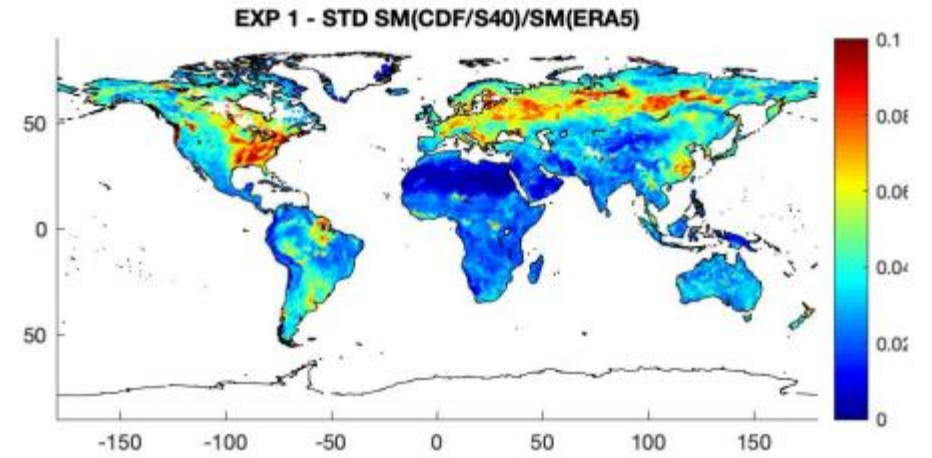


Preliminary validation results

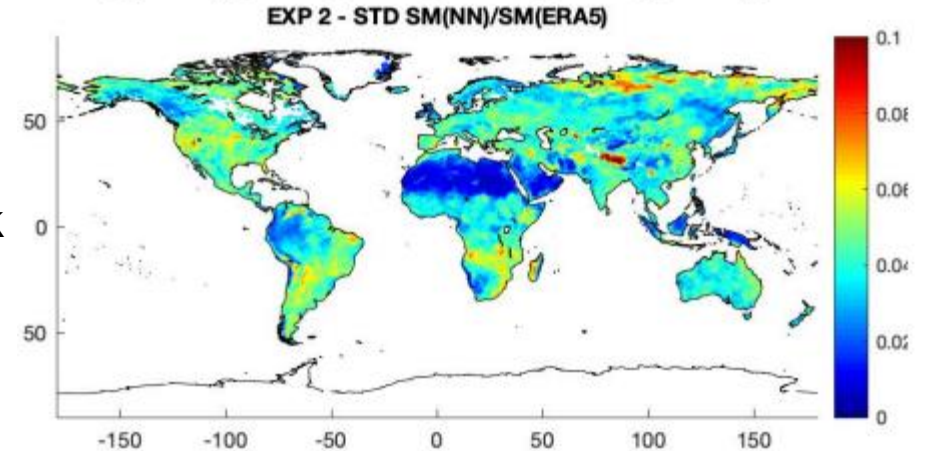
- Pure CDF-matching works best:
 - Over tropics, Southern extra-tropics
 - Smallest biases
- Neural network better:
 - Over Northern extra-tropics
- NN + CDF-matching reduces biases compared to NN only



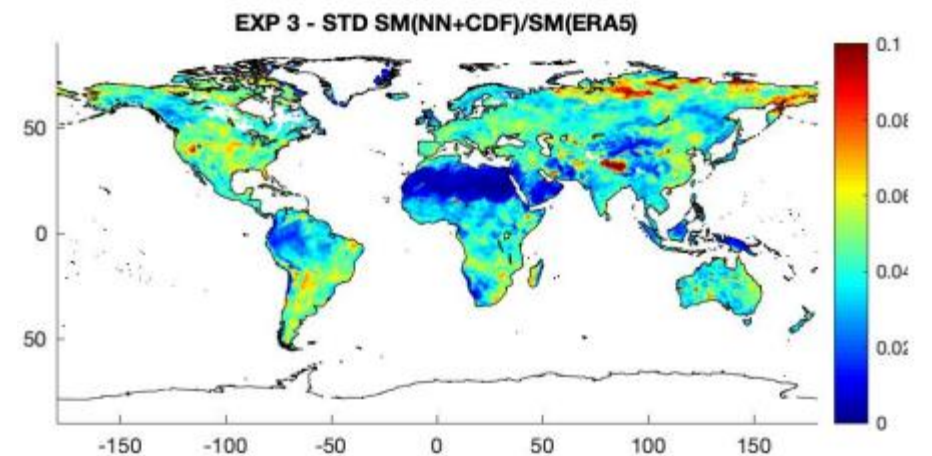
1: CDF-matching



2: Neural network



3: NN + CDF



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Conclusions and future work

- SMOS neural network:
 - Nominal neural network product provides timely soil moisture information
 - Assimilation of neural network product improves T2m and short-range near-surface humidity forecasts
 - Re-training using latest re-calibrated SMOS data
- ASCAT neural network:
 - Aires et. al. (2020), QJRMS, in review:
 - Promising initial ASCAT neural network results for inversion
 - Comparison between NN, CDF-matching and combined methods
 - HSAF visiting scientist mission by F.Aires in 2021:
 - Further development and evaluation
 - Assimilation experiments
 - New soil moisture product planned for HSAF CDOP-4
 - Development of an observation operator for direct backscatter assimilation