

Introduction

What are Fiducial Reference Measurements (FRMs) ?

- Output of satellite communities
- Fully characterized & traceable in situ measurements following community agreed guidelines (GEOS/CEOS QA4EO framework)
- EO data easily & openly accessible
- Data with associated Quality Indicator (QI) → to evaluate its fitness for purpose
- Traceable QIs → internationally agreed reference standards (SI if possible)

ESA's FRM activities typically comprise activities:

- Establishing ground-based FRM networks for a particular variable
- Specify the protocols and procedures to establish and use such FRM data
- Validate relevant satellite products against established FRM data

Objectives of FRM4SM

FRM4SM targets addressing all the above mentioned goals through:

- Evolution of the International Soil Moisture Network (ISMN): <https://ismn.earth>
- Evolution of the Quality Assurance for Soil Moisture – free online validation service (QA4SM): <https://qa4eo.eu>
- Development of a set of in situ soil moisture QIs fully describing uncertainty characteristics
- Development of an “FRM Protocols and Procedures” document – building upon community agreed standards
- Improvement of uncertainty understanding in SSM observations – ISMN/ SMOS validation case studies

→ More info on <https://project-frm4sm.geo.tuwien.ac.at/>

Project questions & identified issues

- (1) What makes “fiducial reference data” fiducial?
- (2) Is the creation of a globally-representative FRM subset already feasible for SSM?
- (3) What are the current limitations of in situ observations that limit fiduciality?
- (4) What is needed to create a full traceability chain from in situ point measurements to the satellite footprint scale?

FRMs ought to be fully characterized and traceable ground measurements to support satellite Cal/Val

- Issue 1:** Most in situ data providers do not provide any uncertainty information → no calibration standard
- Issue 2:** Missing standards to calculate in situ soil moisture uncertainty budget → output FRM4SM data
- Issue 3:** Upscaling to the satellite scale typically breaks traceability
- Issue 4:** In situ networks have a strong spatial bias and thus cannot fully capture satellite uncertainty variations
- Issue 5:** FRMs could differ for individual satellite missions

Uncertainty components & challenges

FRM Qualification for ISMN sensors in top layer:

- Spatial Representativeness Quality Indicator (QI)
- Using Triple Collocation (TC) = error estimates for time series

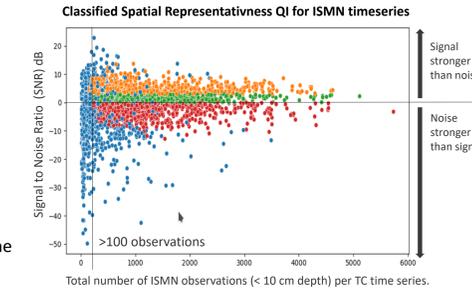
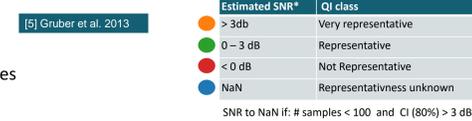
Methodology:

1. Triple Collocation TC (CCI passive, ERA5 land and ISMN)
2. Signal to Noise Ratio (SNR)

$$Tc = \text{in situ error} + \text{representativeness error}$$

$$SNR(dB) = 10 \log_{10} \frac{P_s}{P_N} \cdot \frac{Ps \dots \text{Signal}}{PN \dots \text{Error}}$$

3. Bootstrapping → Confidence Interval 80%
3. Classification of spatial Representativeness QIs for ISMN time series



Sensor life: Time series buddy check possible?

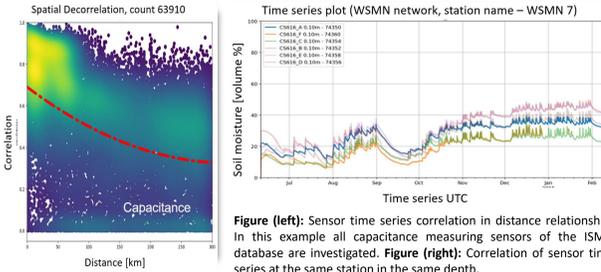


Figure (left): Sensor time series correlation in distance relationship. In this example all capacitance measuring sensors of the ISMN database are investigated. Figure (right): Correlation of sensor time series at the same station in the same depth.

Investigating neighbouring sensors

- Installation of three sensors per depth for comparison
- Not easy to understand drift effects
- Comparison of sensors in close proximity – extremely critical (see figure left)
- Sensor life time quite good for at least 10 years (ISMN community)
- Re-calibration of sensors not really done – disturbance of soil

Installation & Environmental factors

→ Sensor loss or resulting poor data quality through problem with contact soil – sensor :

- Incorrect installation (e.g., inexperience, ect.)
- Animals activity (e.g., mice, deer, snails, moles, worms, etc.)
- Harsh environmental conditions (e.g., high latitudes - ice shifts, soil shifts, etc.)
- Heavy weather events (e.g., extreme rain events can flood the sensor, extreme drying events can crack the soil, etc.)
- Agricultural activity (e.g., ploughing the fields, harvesting, etc.)

→ FRM super sites possible across other EO variables

Calibration of soil moisture sensors

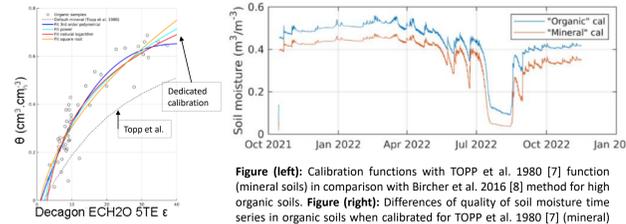


Figure (left): Calibration functions with TOPP et al. 1980 [7] function (mineral soils) in comparison with Bircher et al. 2016 [8] method for high organic soils. Figure (right): Differences of quality of soil moisture time series in organic soils when calibrated for TOPP et al. 1980 [7] (mineral) versus Bircher et al. 2016 [8] (organic soils).

No calibration standard to this date

- Many functions can be applied e.g., TOPP, CRIM, Mironov, etc.
- Usually only for mineral soils sometimes high organic soils
- Measurement highly soil dependent
- Lab calibration usually done with thermo-gravimetric method

[8] Topp et al. 1980 ; [9] Bircher et al. 2016 [10] Roth et al. 1990, [11] Mironov et al. 2004

SI Traceability

- Diversity of different measuring techniques
- Uncertainty budget individual per measurement technique

Challenges for traceability chain

- Scaling differences (in situ versus satellite)
- No uncertainty/calibration info within ISMN

Establishing traceability

Traceability – uncertainty budget:

- 1) Identification of influences on in situ soil moisture measurements
- 2) SI traceability for all soil moisture measurement techniques
- 3) Generating an uncertainty budget per time series

Traceability – data towards FAIR principles (tracking dataset changes):

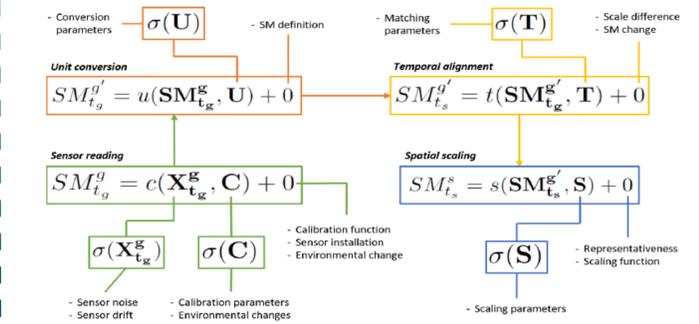
- 1) Versioning of data
- 2) Persistent Identification (PID) for data (e.g., DOI, hdl, etc.)

$$SM_{t_g}^g = c(X_{t_g}^g, C) + 0 \quad \text{SM(ground scale, ground sampling time, sensor units) Sensor reading}$$

$$SM_{t_g}^g = u(SM_{t_g}^g, U) + 0 \quad \text{SM(ground scale, ground sampling time, satellite units) Unit conversion}$$

$$SM_{t_g}^g = t(SM_{t_g}^g, T) + 0 \quad \text{SM(ground scale, satellite overpass time, satellite units) Temporal alignment}$$

$$SM_{t_g}^g = s(SM_{t_g}^g, S) + 0 \quad \text{SM(satellite scale, satellite overpass time, satellite units) Spatial scaling}$$



Effect	Type	Correlated	Magnitude	Confidence
Sensor noise	R	N		
Sensor drift	S	N		
Calibration parameters	S	P		
Calibration function	S	P		
Sensor installation	S	P		
Environmental factors	R+S	P		
Conversion parameters	S	P		
SM definition	S	P		
Matching parameters	S	P		
Scale difference	R+S	P		
SM decorrelation	R+S	P		
Scaling parameters	R+S	P		
Spatial representativeness	R+S	P		
Scaling function	R+S	P		

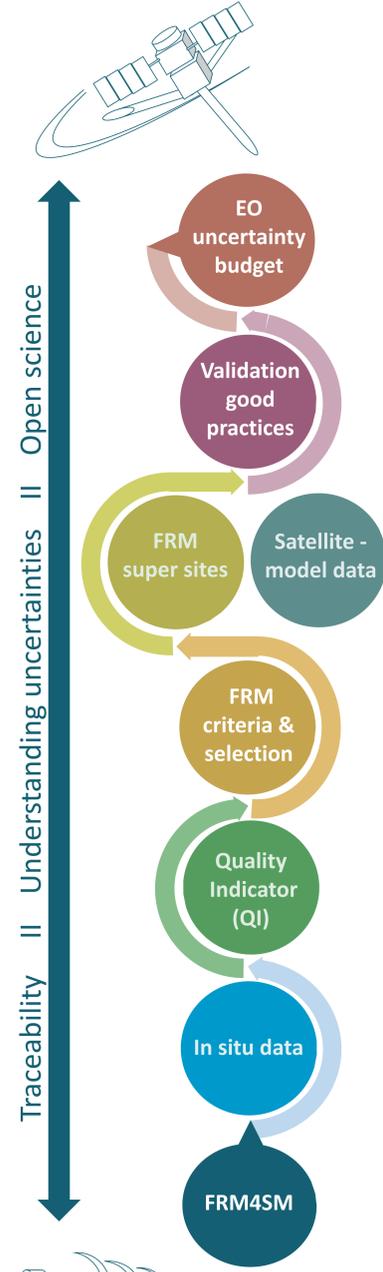
Effect and Figure (top): The equation shows the different currently identified uncertainty sources which are then included in the traceability Diagram figure below. The + 0 = no uncertainty attached (assumption).

Table (left): This table shows the effects table for all currently identified uncertainty factors and tries to list the differences of these errors: R = Random, S = Systematic; Correlated: Y = Yes, N = No, P = Potential; 1 = Estimates only; 2: Some analysis performed to evaluate; 3: Rigorous analysis performed. Magnitudes are given in m3 / m3.

Key message / future direction

- Important to find all uncertainty influences
- Calculation of uncertainty budget
 - Important to understand current limitations
 - Finding a way to quantify uncertainty associated with
 - (lacking) calibration
 - Installation & environmental factors
 - Working more interdisciplinary
- Clear understanding what is really needed for soil moisture satellite validation
 - How does a best possible data quality look like (= FRM4SM subset of data)
 - Finding common ground and differences per satellite mission
 - FRM(4SM) super sites
- Building upon and towards community agreed standards
 - Data sharing towards open source and FAIR principles
 - Best practices and recommendations for best possible FRM4SM data

Traceability || Understanding uncertainties || Open science



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