

High resolution soil moisture drought monitoring over Luxembourg

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Objectives

- Intercompare different high and low resolution soil moisture datasets to in situ data
- Compute drought indices from soil moisture datasets to monitor severe droughts over the past decade
- Compare spatial patterns to geology and land use and temporal patterns to precipitation variability
- Determine differences and offsets between soil moisture droughts and impacts on vegetation

Research Question

How do high resolution drought monitoring strategies compare to those on lower resolutions?





Data

Satellite datasets

CCI active soil moisture data Spatial resolution: 0.25° ERS and ASCAT Input:



- C-band Van der Sat soil moisture data Spatial resolution: 100 m AMSR2, AMSRE Input: (van der Sat algorithm)
- **MULESME** soil moisture data Spatial resolution: 60 m Sentinel-1 Input: (MULESME algorithm)
- **NDVI** vegetation data Spatial resolution: 10 m Sentinel-2 Input:



- and ASTA



Methods Data preparation

An exclusion map for insensitive backscatter pixels was derived using Zhao et al., 2020. Over these pixels, the MULESME algorithm does not retrieve soil moisture.

Data X (NDVI, soil moisture) were normalized using:

$$z = \frac{X - \bar{X}}{\sigma_{v}}$$

With a reference period between 2015 and 2022.

SPI1 was computed for each station according to McKee et al., 1993, and spatial maps were created using Voronoi polygons.

In situ datasets Precipitation and temperature from LIST

- Soil moisture from LIST; ASTA; Blume & Weiler; van Hateren et al.; Foets et al.

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Results







- soil moisture datasets
- temporally and per dataset
- moisture estimates



The accuracy of satellite soil moisture data depends on land cover & soil

Drought conditions as identified in SPI1 are reflected in most satellite

- The **spatial correlation** between satellite soil moisture and SPI1 varies

- High resolution soil moisture datasets show more **spatial variation** in drought conditions than low resolution datasets

Geology (van der Sat) and land use (Copernicus) influence the soil

Zhao et al., 'Deriving Exclusion Maps from C-Band Sar Time-Series: An Additiona Information Layer for Sar-Based Flood Extent Mapping', ISPRS Annals of the Photogrammetry, Remote Sensing and Spatial Information Sciences V-1–2020 (**2020**): 395–400, http://dx.doi.org/10.5194/isprs-annals-V-1-2020-395-2020 Foets et al., 'Temporal and Spatial Variability of Terrestrial Diatoms at the Catchment Scale: Controls on Communities', PeerJ 8 (2020): e8296, https://doi.org/10.7717/peerj.8296 van Hateren et al., 'On the potential of Sentinel-1 for sub-field scale soil moisture monitoring', International Journal of Applied Earth Observation and

Geoinformation (**under review**) **McKee et al.**, 'The Relationship of Drought Frequency and Duration to Time Scales', in Eighth Conference on Applied Climatology (1993). Blume & Weiler, Soil moisture data resulting from the CAOS project (2022)







			2020						2022					
7	8	9	4	5	6	7	8	9	4	5	6	7	8	9
0	-0.05	0.12	0.16	0.62	0.18	0.23	-0.35	-0.14	-	-	-	-	-	-
5	0.01	0.15	0.29	0.60	0.08	0.39	0.37	0.43	-0.10	0.25	0.50	0.62	0.41	-0.45
1	-0.31	-0.15	0.30	-0.37	0.03	-0.24	0.42	0.03	-0.36	-0.45	-0.24	0.04	0.43	0.47
4	0.28	0.24	0.53	0.62	0.12	-0.03	-0.19	0.33	-0.28	-0.11	0.39	0.38	0.58	0.12



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