

VALIDATION OF SMOS L2 AND L3 SOIL MOISTURE PRODUCTS OVER THE DUERO BASIN AT DIFFERENT SPATIAL SCALES

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ABSTRACT:

An increasing number of permanent soil moisture measurement networks are nowadays providing the means for validating new remotely sensed soil moisture estimates such as those provided by the ESA's Soil Moisture and Ocean Salinity (SMOS) mission. Two types of *in situ* measurement networks can be found: small-scale (100-10000 km²), which provide multiple ground measurements within a single satellite footprint, and large-scale (>10000 km²), which contain a single point observation per satellite footprint. This work presents the results of a comprehensive spatial and temporal validation of a long-term (January, 2010 to June, 2014) dataset of SMOS-derived soil moisture estimates using two *in situ* networks within the Duero basin (Spain). The first one is the Soil Moisture Measurement Stations Network of the University of Salamanca (REMEDHUS), which has been extensively applied for validation of soil moisture remote sensing observations, including SMOS. REMEDHUS can be considered within the small-scale network group (1300 km²). The other network started from an existing meteorological network from the *Castilla y León* region, where soil moisture probes were incorporated in 2012. This network can be considered within the large-scale group (65000 km²). Results from comparison to *in situ* show that the new reprocessed L2 product (v5.51) improves the accuracy of former soil moisture retrievals, making them suitable for developing new L3 products. Validation based on comparisons between dense/sparse networks showed that temporal patterns on soil moisture are well reproduced, whereas spatial patterns are difficult to depict given the different spatial representativeness of ground and satellite observations.

1. INTRODUCTION

Recently, the World Meteorological Organization (WMO), the Global Climate Observing System and the Committee on Earth Observation Satellites included soil moisture as one of the Essential Climate Variables (WMO, 2010), highlighting its importance in meteorology and climatology, taking a central role in the land surface-atmosphere interface, and in the interactions between soil, vegetation and climate forcing. It is also very important for agricultural applications, as it represents the reservoir of the plant available water.

In the last few years, several satellites have been launched dedicated to globally measuring the Earth's surface soil moisture based on passive microwave techniques at the frequency band from 1 to 2 GHz (L-band) (Jackson et al., 2012). The Soil Moisture and Ocean Salinity (SMOS) satellite was launched by the European Space Agency (ESA) and is providing the first global maps of soil moisture since January 2010 (Kerr et al., 2010). The second dedicated mission, the National Aeronautics and Space Administration (NASA)'s Soil Moisture Active Passive (SMAP) mission, was launched in January 2015. It aims at retrieving soil moisture and the freeze/thaw state using the combination of active and passive technologies (Entekhabi et al., 2010). A third instrument, Aquarius, was launched in June 2011 on board the international Aquarius/SAC-D mission by NASA and Argentina's space to primarily measure sea surface salinity. This mission is also providing soil moisture products (Lagerloef et al., 2008).

Remotely sensed soil moisture measurements have the advantage of covering large areas and identify synoptic events that are difficult to characterize using *in situ* measurements, but their averaged and coarse-resolution nature renders validation of these products difficult. A key issue is to bridge the gap between point scale measurements with footprint-scale estimates. Crow et al. (2012) distinguishes between two types of networks, small-scale and large-scale. The small scale networks (between 100 km² and 10000 km²) have the advantage of providing multiple measurements within a single footprint, and therefore allows for the examination of the sub-footprint scale. The large scale networks (>10000 km²), in turn, have the advantage of covering large areas and a larger range of land cover soil types, but they typically lack of multiple measurements per footprint.

In this work, the last version (5.51) of SMOS L2 soil moisture product and a new SMOS BEC L3 v.001 soil moisture product were tested over an area of more than 65000 km² located at the Duero Basin, Spain, for the period January 2010 to June 2014. Two nested *in situ* networks were used: the Soil Moisture Measurement Stations Network of the University of Salamanca (REMEDHUS, small-scale) and the recent Inforiego network (large-scale). Spatio-temporal agreements of the L2 and L3 soil moisture temporal series were analysed through a comparison with four different strategies: point-scale measurements at the two networks, area-averaged, land-use-averaged and soil-texture-averaged data.

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2. IN SITU NETWORKS

2.1 REMEDHUS network

REMEDHUS covers an area over 1300 km² located in the central Duero Basin (41.1° to 41.5°N and 5.1° to 5.7°W in Spain) ranging from 700 to 900 m.a.s.l., nearly flat (Figure 1). The land uses are rainfed cereals, irrigated crops, vineyards and forest-pasture. This region has a continental semi-arid Mediterranean climate, with a mean temperature of 12°C and an average annual precipitation of 385 mm. The network is equipped with 23 automated stations with capacitance probes (Hydra Probes, Stevens Water Monitoring System, Inc.) measuring every hour in the top 5 cm of the soil.

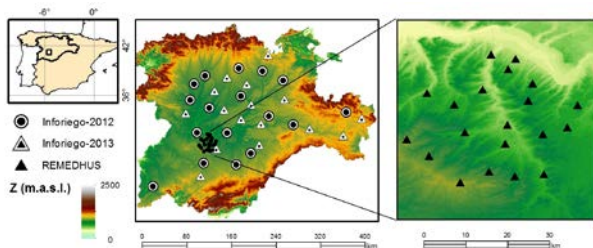


Figure 1. Location of the two networks used in this study.

REMEDHUS has been widely used in the validation of soil moisture products due to the characteristics and homogeneity of the area. REMEDHUS is a member of the International Soil Moisture Network (Dorigo et al., 2011, <https://ismn.geo.tuwien.ac.at/>). A detailed description of this network can be found in previous works (Sánchez-Ruiz et al., 2014).

2.2 Inforiego network

This soil moisture network was set up profiting an existent meteorological network dedicated to irrigation assessment (<http://www.inforiego.org/>). The network is deployed over the Duero River Basin (north-western part of the Iberian Peninsula) and it covers an area over 65000 km² (Figure 1). The area is surrounded by mountains to the north, south and east, with an average height over 800 m.a.s.l. The land uses in this network are similar to those of REMEDHUS: rainfed crops in winter-spring and irrigated crops in summer. The climate of the region is continental semi-arid Mediterranean with an average temperature of 11.7°C and an average annual precipitation of 450 mm.

Inforiego soil moisture network had 17 stations in use from July 2012 to July 2013 (hereafter Inforiego-2012) and other 16 different stations were available between August 2013 and August 2014 (hereafter Inforiego-2013). These 33 stations were selected following the distribution and localization of the main sub-basins of the Duero River Basin and the representativeness of the soil type and land use. As in the REMEDHUS network, Hydra Probes were installed at a 5 cm depth.

3. SMOS DATA

3.1 SMOS L2 soil moisture

The SMOS Soil Moisture Level 2 User Data Product (SMUDP2 file) version 5.51 contains retrieved geophysical parameters (soil moisture and optical thickness) as well as complementary quality parameters. This product is distributed in an Icosahedral

Snyder Equal Area Earth grid (ISEA-4H9) with equally spaced nodes at approximately 15 km, called Discrete Global Grid (DGG). A more detailed description of the L2 algorithm can be found in Kerr et al. (2012). Data of L2 soil moisture v5.51 from January 2010 to June 2014 over the river Duero Basin were used in this study (Table 1). The REMEDHUS area is covered by 11 DGGs, and Inforiego area is covered by approximately 300 DGGs.

Two filter criteria were used. The first filter was a threshold over the Data Quality Index (DQX, which represents the uncertainty of the retrieval) of 0.04 (volumetric soil moisture units, m³m⁻³) was used to select the best-quality SMOS retrievals. This threshold restricts the retrieval to the theoretical accuracy of the SMOS targeted accuracy and is usual in validation experiments. The second filter is based on the probability of occurrence of Radio Frequency Interferences (RFIs), which is updated in a new flag (RFL_Prob flag) in this version for each grid node. Users can use this indicator as a predictor of the RFI corruption, ranging from 0 (no RFI probability) to 1 (expected corrupted data). For this research, the use of several thresholds was explored. When a filter of 0.4 was applied, the number of data was reduced by 3.8% for ascending orbits and 3.1% for descending orbits. Using a more restrictive filter of 0.3, the data reduction was 5.8% for ascending and 6.4% for descending orbits. Thus, prior to further analysis, three levels of RFI filtering were tested: 0.4, 0.3 and none. There were no meaningful differences in the validation results, except in the Root Mean Square Difference, which was reduced by 0.01%. Thus, only DQX filtering was applied.

Soil moisture datasets	Number of data	
	Asc.	Des.
REMEDHUS network	1642	1642
Inforiego-2012 network	363	363
Inforiego-2013 network	334	334
SMOS L2 (REMEDHUS)	835	881
SMOS L2 (Inforiego-2012)	229	248
SMOS L2 (Inforiego-2013)	234	233
SMOS BEC L3 (REMEDHUS)	804	851
SMOS BEC L3 (Inforiego-2012)	226	251
SMOS BEC L3 (Inforiego-2013)	233	235

Table 1. Number of data for each dataset

3.2 SMOS BEC L3 soil moisture product

This product is provided by the SMOS Barcelona Expert Centre (<http://cp34-bec.cmima.csic.es/>). SMOS BEC L3 v.001 soil moisture daily maps are computed by quality-filtering and reprojecting SMOS L2 SMUDP files from ISEA 4H9 to 25 km EASE-ML grid. Soil Moisture data are discarded when the probability of RFI is high, the quality of the retrieval is poor (Data quality Index (DQX) greater than 0.07), the soil moisture value is negative or outside of the extended range, or the retrieval has failed. The filtered values are then DQX-weighted averaged into an EASE-ML 25 km grid. Ascending and descending orbits are processed separately. These products are created in a variety of generation rates and averaging periods: 1 and 3 days-generated daily-, 9 days-generated every 3 days-, monthly, seasonal (quarterly) and annual. The fields given by grid cell are: Soil Moisture value, DQX, variance of the averaged value, and number of L2 soil moisture estimates used in the computation.

The REMEDHUS area is covered by 4 L3 pixels, and Inforiego is covered by approximately 100 L3 pixels.

4. METHODOLOGY

L2 and L3 soil moisture datasets were compared with the *in situ* data at the two spatial scales provided by the REMEDHUS and Inforiego networks. Spatial and temporal comparisons were performed.

For the temporal comparisons, the entire time series of the satellite soil moisture estimates were compared with ground measurements that were previously collocated in time with the SMOS overpasses. At point-scale measurements at each station with its collocated DGG (L2) and pixel (L3) were compared. Also, the *in situ* data from both networks were averaged using area-average, land-use-average, soil-texture-average. SMOS ascending and descending passes were distinguished in order to assess their individual performances. Regarding the land use, four categories were taken into account (rainfed cereals, vineyard, forest-pasture and irrigated) and for texture, four soil categories were considered: fine (including silty clay loam, clay loam, clay and loam textures), medium (including sandy loam and clay loam textures), coarse (including loamy sand texture) and very coarse (including sand texture). The distribution of the soil texture compositions of the 56 stations used showed a slightly high sand content at most of the stations (Figure 2).

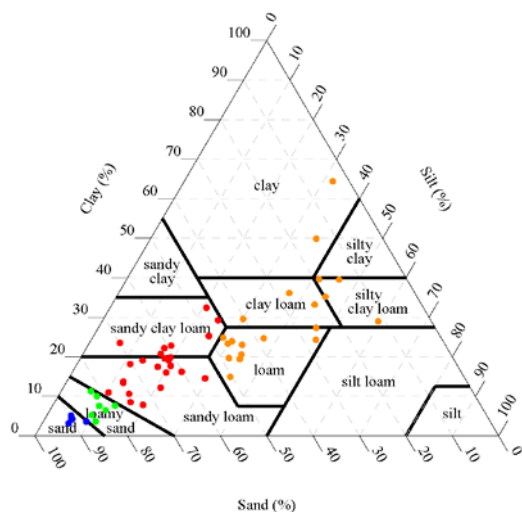


Figure 2. Soil texture distribution of the 56 stations used. Orange, red, green and blue colors represent fine, medium, coarse and very coarse texture, respectively.

Regarding the spatial validation, spatially collocated SMOS and *in situ* observations for each day were compared. A threshold of 16 concurrent observations was established as the minimum to compute statistics for a given day. This threshold tried to preserve a robust comparison while taking into account the number of the stations in each network.

The statistical metrics chosen for evaluate the comparison were the correlation coefficient (R of Pearson), the RMSD and the centered RMSD (cRMSD), the bias and the standard deviation (std). Additionally, Taylor diagram (Taylor, 2001) was used for the temporal validation at point-scale. This diagram relates the metrics of R, std and cRMSD.

5. RESULTS AND DISCUSSION

5.1 Temporal validation

The soil moisture time series in REMEDHUS (Figure 3) and Inforiego (not shown) show that the SMOS data set series have a larger dynamic range than the *in situ* data. For REMEDHUS, *in situ* measurements have a mean value of 0.141 with a minimum of 0.052 in summer and a maximum of 0.280 in winter, with a std of 0.054. The SMOS L2 ascending data ranged from 0.003 to 0.356, with a std of 0.075, and for descending data from 0.009 to 0.439, and the std values are similar for both orbits. The difference between both products is that L2 have larger dynamic range than L3. The SMOS dataset followed the temporal dynamics of the REMEDHUS dataset, although a slight underestimation can be observed, particularly in the dry periods. This dry bias is consistent with recent studies of calibration/validation on SMOS; however, the underestimation observed in this study is considerably smaller than that obtained with previous SMOS L2 versions (Dall'Amico et al., 2012; Sánchez et al., 2012). SMOS showed a quicker reactivity to rainfall events and dry-downs than the ground observations. Both effects, underestimation and reactivity, were already detected in previous experiments on validations in the same area (Sánchez et al., 2012).

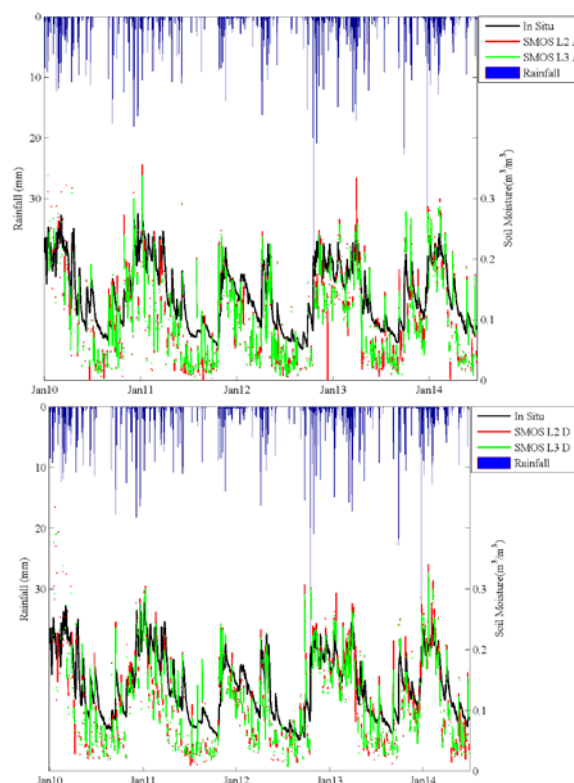


Figure 3. Evolution of the soil moisture *in situ* measurements (average of stations) and SMOS products (average of DGGs/Pixels), along with rainfall, during the study period (January 2010-June 2014) for the REMEDHUS network. Top, ascending orbit, bottom, descending orbit.

The reprocessed v5.1 reduces outliers and leads to a more accurate estimation respect to older versions (Albergel et al., 2012; Dall'Amico et al., 2012; Jackson et al., 2012; Sánchez et al., 2012; Dente et al., 2012), with values in area-averaged comparison of $R \sim 0.80$; the correlation is higher for the

Inforiego networks (Table 2). The SMOS cRMSD reached the expected accuracy of $0.04 \text{ m}^3\text{m}^{-3}$, and the bias was always positive (i.e., dry bias); the value is higher for both Inforiego networks (Table 2) than for REMEDHUS. This effect can be explained by the length of the time series, which is shorter for Inforiego, resulting in more accentuated seasonal effects that are smoothed in the longer time series of REMEDHUS.

		L2		L3	
		Asc.	Des.	Asc.	Des.
REMEDHUS	R	0.80	0.81	0.80	0.79
	bias	0.034	0.023	0.037	0.027
	RMSD	0.057	0.051	0.056	0.051
	cRMSD	0.045	0.045	0.042	0.043
Info. 2012	R	0.84	0.88	0.84	0.87
	bias	0.073	0.042	0.078	0.047
	RMSD	0.088	0.064	0.093	0.069
	cRMSD	0.050	0.048	0.050	0.050
Info. 2013	R	0.85	0.85	0.87	0.80
	bias	0.059	0.050	0.063	0.038
	RMSD	0.071	0.073	0.073	0.061
	cRMSD	0.040	0.053	0.038	0.047
All together	R	0.78	0.83	0.79	0.82
	bias	0.034	0.014	0.034	0.014
	RMSD	0.056	0.040	0.055	0.040
	cRMSD	0.044	0.037	0.043	0.037

Table 2. Results of the comparison between the spatially averaged series of soil moisture for the three networks in each usable time series.

From the time series comparison, it can be stated that the SMOS soil moisture products were generally underestimated. This underestimation appears to be larger for the ascending products than for the descending (Table 2) and for shorter time series. Regarding the products and orbits, there was no difference between the L2 and L3 performances, and the ascending and descending series were similar in terms of R and cRMSD, with a bias noticeably smaller for the descending series (Table 2). When computing the comparison of all 56 stations (REMEDHUS + two Inforiego), the statistics slightly improved. Increasing the number of stations and integrating the scales led to a smaller bias.

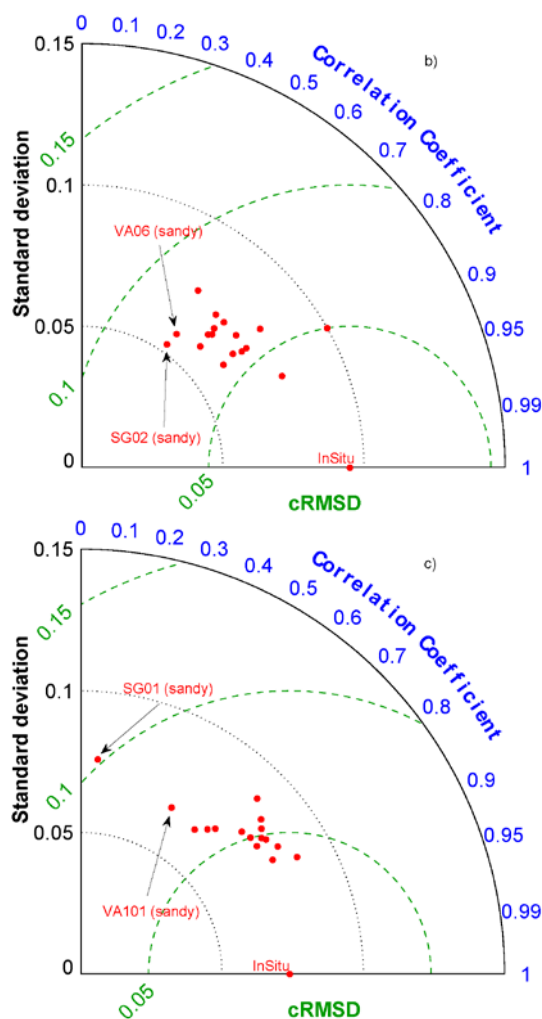
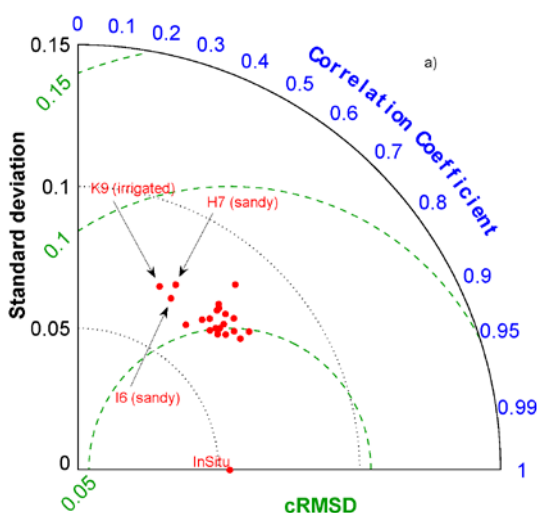


Figure 3. Results of the comparison between each station in: a) the REMEDHUS network, b) the Inforiego-2012 network, c) the Inforiego-2013 network and its corresponding L2 ascending DGG. Only stations with results discussed in the text are labeled.

Using Taylor diagrams (Figure 3, only L2 ascending data was shown) it can be seen R ranging from 0.41 to 0.8, and a cRMSD from 0.045 to 0.145 for the L2/L3 and ascending/descending data respectively, in REMEDHUS (Figure 3a). Similar metrics were obtained for the Inforiego networks. The worst results were obtained for K9, H7 and I6. These poor results in H7 and I6 may be attributed to the very high sand content at these stations (85% and 90%, respectively), which is consistent with a particular behavior of these soils that cannot be captured by the average values of SMOS (Panciera, 2009; Crow et al., 2012). Soils with higher sand content exhibited persistently drier soil moisture conditions than soils with finer textures. The K9 station is located in an irrigated area, where the soil moisture pattern is controlled by repetitive irrigations events. Inforiego-2012 (Figure 3b) and Inforiego-2013 (Figure 3c) had better results than REMEDHUS, with $R \sim 0.91$, and cRMSD ranged between 0.036 and 0.102. Also, the *in situ* cRMSD is smaller than that obtained with REMEDHUS. The scatter of the cRMSD values of the Inforiego stations is higher than REMEDHUS (Figure 3), and the worst results were obtained for the sand stations (SG02, VA06, SG01 and VA101, H7 and I6). In 10 out of 23 REMEDHUS stations and in 7 out of 33 Inforiego stations, the bias was negative, indicating

underestimations (not shown). Overall, the bias for each station in both networks ranged between -0.084 and 0.215.

In summary, the results of the comparisons at the point scale were less satisfactory than those taking into account the regional average of each network, in agreement with other works. According to Bircher et al. (2012), point measurements do not necessarily support coarse-resolution satellite observations due to the heterogeneity of the soil and the vegetation cover. Several strategies of data upscaling are required to more effectively translate the information derived from sparse point-scale ground-based sensors to satellite footprint resolutions (Crow et al., 2012).

The results of the comparisons regarding to soil use and texture, were in agreement with the results obtained for the area-averaged and point-scale analyses, showing a generally better match for the Inforiego networks than for REMEDHUS. No differences between orbits were detected, with values of $R > 0.80$ in all cases excepting for the irrigated land use ($R \sim 0.46$) and very coarse textures ($R \sim 0.60$). In contrast, for the higher soil moisture contents (forest-pasture and fine textures), the bias is the highest (bias > 0.180). Rainfed land use and medium soil textures, which are common in the area, exhibited the best results in terms of R (> 0.82), bias (< 0.05) and cRMSD (< 0.05).

5.2 Spatial validation

The daily validation showed poor results for both networks (Table 3). Of all the dates studied, only 13.9% of the comparisons have a p -value < 0.05 . Negative and positive correlations were found for all products and networks. The variability between the satellite products (15 km for L2 and 25 km for L3) and the *in situ* measurements (point scale) is extremely large. Given this spatial difference their direct comparison is debatable.

			Days		cRMSD		RMSD	
			Days	Days $p < 0.05$	Max	Min	Max	min
L2	REMEDI.	Asc.	463	60	0.201	0.040	0.212	0.040
		Des.	461	64	0.196	0.040	0.198	0.040
	Info-12	Asc.	23	1	0.053	-	0.055	-
		Des.	28	0	-	-	-	-
	Info-13	Asc.	5	1	0.058	-	0.072	-
		Des.	3	0	-	-	-	-
L3	REMEDI.	Asc.	584	99	0.189	0.044	0.213	0.051
		Des.	598	82	0.200	0.042	0.214	0.042
	Info-12	Asc.	22	2	0.090	0.053	0.129	0.056
		Des.	37	1	0.055	-	0.061	-
	Info-13	Asc.	1	0	-	-	-	-
		Des.	3	0	-	-	-	-

Table 3. Results of the daily validation for the days were the correlation was significant (p -value < 0.05) given a threshold of 16 concurrent pairwise observations for calculation.

To explore the comparisons at an improved spatial matching, a new spatial validation was performed (not shown) using a SMOS-derived product at a 1 km spatial resolution (Piles et al., 2014). Still, poor results were obtained. No significant correlation was found at the 95% confidence level for the daily correlation at a 1 km resolution at any network.

6. CONCLUSIONS

A comprehensive validation of SMOS L2 and L3 soil moisture products over the Duero basin, Spain, was presented by comparison with *in situ* measurements at a small- and a large-scale network. Compared with former SMOS validations, this version of the SMOS L2 product (v5.51) exhibits better results in terms of temporal comparison with *in situ* measurements, which demonstrates its potential for developing new L3 products at different temporal resolutions and regular grids, as the ones presented in this work. Indeed, there was not a significant difference between the statistical scores obtained with L2 and L3 data. However, a general dry bias was observed in both the SMOS L2 and the SMOS L3 products. This effect appears to be larger for the shorter time series and for the ascending overpasses. The areal-averaged results were confirmed by the comparisons at the point scale, except at stations with particular conditions (very sandy soil and irrigated plots) were in the comparison obtained worst results. For temporal correlations, the results obtained for the large-scale network were slightly better than for the small-scale site, indicating that enlarging the spatial extent of the network resulted in a better matching. In this line, when all networks were jointly computed, better results were obtained.

A detailed analysis of soil types and land coverages was also performed. Regarding to soil texture and land use, better results were achieved, except for irrigated land use and very-coarse-textured soils. This can be explained by the high sand content, - which coincides with vineyard use- leading to a quick soil moisture decrease in the top soil layer, therefore hindering a good estimation. Conversely, a lower bias in the vineyard land use and very coarse soils was found, maybe due to their typically low soil moisture content. The main type of soils in the area (medium coarse) and land use (rainfed crops) showed the best results for all the statistics used. In this case, the data representativeness and favourable results are correlated.

Regarding the spatial validation, only a few dates show a significant correlation for both networks. It seemed almost impossible (and likely useless) to depict the soil moisture spatial distribution at all stations at a given instantaneous time. This result may be attributed to the contrast between *in situ* measurements, which provide point-scale measurements under very different conditions at each location, and the large footprint of the radiometer observations.

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