Project ID: 59199



# Contribution ID: 226 Improving of the AMSR-E/NASA soil moisture product in global scale using in-situ soil moisture measurements and fractional vegetation cover datasets Qiuxia Xie<sup>1, 2</sup>, Li Jia<sup>2, \*</sup>, Massimo Menenti<sup>2</sup>

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#### **ABSTRACT**

The AMSR-E/NASA (Advanced Microwave Scanning Radiometer-Earth Observing System/National Aeronautics and Space Administration) daily global-scale soil moisture (SM) product, spanning from 2002 to 2011 with a spatial resolution of 25 km, was provided by the NASA National Snow and Ice Data Center Distributed Active Archive Center (NSIDC DAAC). However, the AMSR-E/NASA SM product exhibited limited sensitivity in capturing intra- and inter-annual variability of SM across many regions. Investigation revealed that inaccurate parameter values  $(A_0 \text{ and } A_1)$  in the AMSR-E/NASA SM retrieval algorithm were pivotal in causing this issue. This study sought to improve the global-scale AMSR-E/NASA SM product. Parameter values  $(A_0 \text{ and } A_1)$  were recalibrated using the 13 in-situ observation networks (totaling 192 sites), and their relationships with fractional vegetation cover (FVC) across four land cover types (sparse vegetation, grassland, cropland, and forest) were analyzed. Inversion models for  $A_0$  and  $A_1$  parameters tailored to each land cover type were constructed, utilizing a global FVC dataset. Subsequently, the improved AMSR-E/NASA SM product was generated employing the refined algorithm proposed by Xie et al. (2019). Evaluation of the product against SM measurements from the 6 in-situ observation networks indicated strong agreement, particularly evident in sparse vegetation areas where a linear relationship between A0 (or A1) parameter values and FVC was observed (e.g.,  $A_1$ =-0.61×FVC+1.21 and  $A_0$ =-0.20×FVC+0.012). Conversely, non-linear relationships were prominent in grassland/cropland/forest areas (e.g., A1 =69.04×(FVC)2-28.49×FVC+5.67 for grass). Furthermore, the improved global-scale AMSR-E/NASA SM product demonstrated superior performance compared to AMSR-E/JAXA (Japan Aerospace Exploration Agency) and AMSR-E/LPRM (Land Parameter Retrieval Model) SM products, exhibiting lower Mean Absolute Error (MAE) and Root Mean Square Error (RMSE) values (i.e., 0.026 cm³/cm³ and 0.032 cm³/cm³, respectively).

#### **METHODS**

AMSR-E/NASA soil moisture retrieval algorithm analysis. The original AMSR-E/NASA SM product inversion algorithm is a multi-frequency, multi-polarization method. The coefficients were determined by using a simplified radiative transfer equation to simulate the minimum difference between the brightness temperature and the AMSR-E satellite sensor observations. In 2003, Njoku et al. developed a linear method for retrieving SM from AMSR-E satellite observations (Njoku et al., 2003). The SM product inversion algorithm uses MPDI value and empirical coefficient to invert SM. The formula is as follows:

$$\begin{split} SM^t - SM^{dry} &= a_0 \cdot g^* + a_1 \cdot \left( MPDI^t_{10.7} - MPDI^{dry}_{10.7} \right) \cdot exp(a_2 \cdot g^*) \\ MPDI_{10.7} &= \left( T_{B(10.7V)} - T_{B(10.7H)} \right) / \left( T_{B(10.7V)} + T_{B(10.7H)} \right) \\ g^* &= \beta_0 + \beta_1 \cdot ln(MPDI^t_{10.7}) \end{split}$$

Where, t represents the time in days;  $SM^t$  is SM with time;  $SM^{dp}$  is the minimum SM value (default is 0.05 cm<sup>3</sup>/cm<sup>3</sup>);  $MPDI^t{}_{10.7}$  is the MPDI value of 10.7 GHz on day t;  $MPDI^*{}_{10.7}$  is the annual minimum MPDI value under dry soil conditions.  $g^*$  is the baseline parameter, estimated by  $MPDI^*{}_{10.7}$  (minimum MPDI at the monthly scale), which can be interpreted as equivalent vegetation water content ( $kg/m^2$ ). V and H indicate vertical and horizontal polarization;  $a_0$ ,  $a_1$ ,  $a_2$ ,  $b_0$  and  $b_1$  parameter values need to be obtained. Jackson et al. determined the  $a_0$ ,  $a_1$ ,  $a_2$ ,  $b_0$  and  $b_1$  parameter values need to be obtained. Jackson et al. determined the  $a_0$ ,  $a_1$ ,  $a_2$ ,  $b_0$  and  $b_1$  parameters in 2012 by taking AMSR-E observations from Chad, Sudan, and the Central African Republic, where surface SM is low, and assuming an average SM of 0.1 cm<sup>3</sup>/cm<sup>3</sup>, thus the  $a_0$ ,  $a_1$ ,  $a_2$ ,  $b_0$  and  $b_1$  parameters are determined.

#### **RESULTS**

**Spatiotemporal variations of AMSR-E soil moisture products.** By comparing the global AMSR-E/JAXA and AMSR-E/LPRM SM products with the AMSR-E/NASA SM products on August 1, 2010 (Figure, 2 and 3)

Parameter optimization of AMSR-E/NASA soil moisture retrieval algorithm. B ased on the four land cover types of farmlands, grassland, forest area and sparse vegetation area, the relationship between  $A_0$  and  $A_1$  parameter values of the four land cover types and the monthly minimum MPDI data of 10.7GHz and vegetation coverage was analyzed respectively (Figure 4).

Evaluation for the improved AMSR-E/NASA SM product. Long-term measured data of six sites (Widgiewa site of OZNET observation network, Cullingral site of SASMAS observation network, NST\_09 site of Maqu observation network, HYDROL-NET\_PERUGIA observation network site, ARAPAHO\_RIDGE on the SNOTEL network and Edinburg\_17\_NNE on the USCRN network) located in five typical land cover types, namely irrigated farmland, forest/grassland mixed, grassland, forest area and raon-fed farmland, were selected for comparison and analysis with the improved AMSR-E, AMSR-E/NASA, AMSR-E/JAXA, AMSR-E/LPRM SM products (Figure 5 and Table 5).

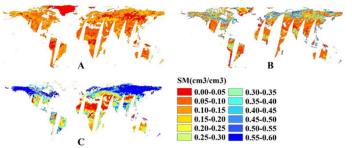


Figure 2.Spatial distribution patterns of AMSR-E (A: NASA; B: JAXA; C: LPRM) SM products.

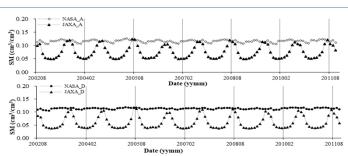


Figure 3. Daily GCF during 2011~2018 of global SMOS, FY3-B, ASCAT, ESA-CCI, SMAP, 1st and 2nd merged SSM (2015~2018 only).

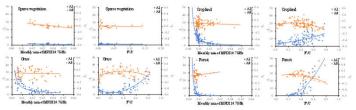
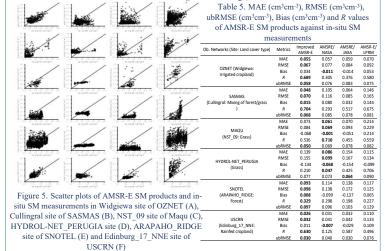


Figure 4. Scatter plots between A<sub>1</sub> (or A<sub>0</sub>) values and monthly min of MPDI in 10.7GHz data (or fractional vegetation coverage).



## CONCLUSIONS

The main conclusions of this study are as follows:

- The spatial and temporal distribution of AMSR-E/NASA, AMSE-E/JAXA and AMSR-E/LPRM SM
  products are significantly different. In particular, AMSR-E/NASA SM product values have small
  changes in time and space, and narrow inter-annual and inter-annual dynamic ranges, which can
  hardly reflect the characteristics of inter-annual and seasonal changes of SM.
- 2. Globally, A<sub>0</sub> and A<sub>1</sub> parameter values are not constant, but change with vegetation change, and there is an obvious nonlinear correlation between them and vegetation coverage. A<sub>0</sub> and A<sub>1</sub> parameter values vary from 0 to 35 and from -0.2 to 0.4, respectively. Compared with AMSR-E/NASA SM products, the improved AMSR-E SM products have a wider spatiotemporal variation range and are more consistent with the spatial distribution pattern of vegetation coverage.
- 3. The improved AMSR-E, AMSR-E/NASA, AMSR-E/JAXA and AMSR-E/LPRM SM products were evaluated using the measured data of 6 observation network sites. The results showed that the improved AMSR-E had better performance and was more consistent with the measured SM data. Compared with the AMSR-E/NASA SM products, the improved AMSR-E had a wider dynamic variation range and could better reflect the characteristics of annual and interannual variation of SM technology.

### **ACKNOWLEDGEMENT**

This study received support from the Natural Science Foundation (Grant No. 42301367), the Open Fund of State Key Laboratory of Remote Sensing Science (Grant No. OFSLRSS202326), the Natural Science Foundation of Shandong Province (Grant No. ZR2022QD138) and the Youth Innovation Team Project of Higher School in Shandong Province (Grant No.2023KJ121).

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