#### Development, Evaluation and Application of New Soil Moisture Products

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- Wang, Y., J. Mao\*, et al. (2021b) Quantification of human contribution to soil moisture-based terrestrial aridity, Under review.

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## Importance of Soil Moisture (SM)



#### **Rationale for Creating Merged Soil Moisture Products**

Data source	Pros	Cons
In situ observations	More accurate than remote sensing or model products	Represent only small spatial scale and too sparse
Remote sensing observations	Global, relatively high- resolution coverage	Have gaps, only represent the topsoil
Land surface models, reanalysis, and Earth system models	Gap-free, represent multiple soil layers	Subject to modeling biases

Merging multiple-source datasets would overcome the limitations of individual datasets, resulting in long-term, global, gap-free, multi-layer SM products for research purposes.



### **Existing Data Merging Efforts for Soil Moisture**

Data Descriptor Open Access Published: 27 May 2021

#### A long term global daily soil moisture dataset derived from AMSR-E and AMSR2 (2002–2019)

Panpan Yao, Hui Lu ⊠, Jiancheng Shi, Tianjie Zhao, Kun Yang, Michael H. Cosh, Daniel J. Short Gianotti & Dara Entekhabi

Scientific Data8, Article number: 143 (2021)Cite this article1123Accesses2AltmetricMetrics

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Remote Sensing of Environment Volume 203, 15 December 2017, Pages 185-215



ESA CCI Soil Moisture for improved Earth system understanding: State-of-the art and future directions

Wouter Dorigo <sup>a</sup> A <sup>B</sup>, Wolfgang Wagner <sup>a</sup> <sup>B</sup>, Clement Albergel <sup>b</sup> <sup>B</sup>, Franziska Albrecht <sup>c</sup> <sup>B</sup>, Gianpaolo Balsamo <sup>d</sup> <sup>B</sup>, Luca Brocca <sup>e</sup> <sup>B</sup>, Daniel Chung <sup>a</sup> <sup>B</sup>, Martin Ertl <sup>f</sup> <sup>B</sup>, Matthias Forkel <sup>a</sup> <sup>B</sup>, Alexander Gruber <sup>a</sup> <sup>B</sup>, Eva Haas <sup>c</sup> <sup>B</sup>, Paul D. Hamer <sup>g</sup> <sup>B</sup>, Martin Hirschi <sup>h</sup> <sup>B</sup>, Jaakko Ikonen <sup>i</sup> <sup>B</sup>, Richard de Jeu <sup>j</sup> <sup>B</sup>, Richard Kidd <sup>k</sup> <sup>B</sup>, William Lahoz <sup>g</sup> <sup>B</sup>, Yi Y. Liu <sup>1</sup> ... Pascal Lecomte <sup>g</sup>

#### Validation of a New Root-Zone Soil Moisture Product: Soil MERGE

TED TOPICS IN APPLIED EARTH OBSERVATIONS AND REMOTE SENSING, VOL. 12, NO. 9, SEPTEMBER 2019

3351

Kenneth J. Tobin<sup>10</sup>, Wade T. Crow<sup>10</sup>, Member, IEEE, Jianzhi Dong, and Marvin E. Bennett

Data Descriptor Open Access Published: 12 July 2021

## Global soil moisture data derived through machine learning trained with *in-situ* measurements

Sungmin O. 🖂 & Rene Orth

Scientific Data 8, Article number: 170 (2021) | Cite this article

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## **Merging Framework**

Gridded data \ Method	Unweighted averaging	<u>Optimal Linear Combination</u> (using in situ soil moisture observations)	<u>Emergent Constraint</u> (using gridded observed meteorological data)						
<u>O</u> ffline land surface model simulations, <u>R</u> eanalysis, and <u>S</u> atellite	Mean ORS	olc ors	EC ORS						
Earth system models (CMIP5, CMIP6)		—	EC CMIP5, EC CMIP6, EC CMIP5+6						
ORS and Earth system models			EC ALL						

#### 7 hybrid SM products based on 3 merging methods:

- ✓ Coverage: global, 1970–2016;
- ✓ Spatial resolution: 0.5°;
- ✓ *Temporal resolution: monthly;*
- ✓ Vertical layers: 0-10cm, 10-30cm, 30-50cm, 50-100cm;



#### **Evaluation Datasets**

Dataset	Туре	Period	Depth (cm)	Resolution	Coverage	Reference		
SMOS L3 RE04 MIR_CLF3MA, MIR_CLF3MD	Satellite	2010–2020	Surface (0–5)	~25 km	Global with missing values	(Al Bitar et al., 2017)		
SMOS L4 SCIE MIR_CLM4RD	Reanalysis	2010–2020	0–100	~25 km	Global with missing values	(Al Bitar et al., 2013)		
GLEAM v3.3a	Reanalysis	1980–2018	0–100	0.25°	Global	(Martens et al., 2017)		
SMERGE v2	Reanalysis	1979–2019	0–40	0.125°	Contiguous United States	(Tobin et al., 2017)		
SoMo.ml	Machine learning upscaled from in situ observations	2000–2019	0–10, 10–30, 30– 50	0.25°	Global	(O and Orth, 2020)		



#### Performance of the Merged SM Products (Site Eval.)



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#### Performance of the Merged SM Products (Global Eval.)

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0-100cm



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#### Performance of the Merged SM Products (Frequency Eval.)



Power

#### Performance of the Merged SM Products (Drought Eval.)



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#### Performance of the Merged SM Products (Sensitivity Eval.)



11

**Dominant driving factors** 

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Yaoping Wang, Jiafu Mao<sup>\*</sup>, Forrest M. Hoffman, Céline J. W. Bonfils, Hervé Douville, Mingzhou Jin, Peter E. Thornton, Daniel M. Ricciuto, Xiaoying Shi, Haishan Chen, Stan D. Wullschleger, Shilong Piao, and Yongjiu Dai, *Quantification of Human Contribution to Soil Moisture-based terrestrial aridity*, Under review, 2021.

of the year contain detectable signal, indicating the presence of seasonal variability;  $\checkmark$  The signals are attributable to anthropogenic forcings (ALL/ANT), especially greenhouse gases (GHG);  $\checkmark$  Provide a comprehensive basis for drought risk reduction

strategies and

activities;



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Evapotranspiration/GLEAM																					
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#### **Application II: ILAMB Hydrology Benchmarking**

- ILAMB evaluates model results by comparing with global-, regional-, and site-scale data
- The current set of variables and datasets (blue text) are shown below



Missing Data or Error ILAMB 2.5 (0705c73e07947221604bfdda0004e1999dbcb4ac)

CMIP6 models (Hoffman et al., in prep)

#### **Application II: <u>ILAMB SM Benchmarking</u>**

14



### Summary of the Merged SM Products

- Achieved the goal of creating long-term, gap-free, multi-layer SM products (<u>https://doi.org/10.6084/m9.figshare.13661312.v1</u>);
- The merged SM products showed reasonable performances, and were broadly within the estimates reported by previous SM evaluations;
- Three "offline-based" SM products (mean ORS, OLC ORS, and EC ORS) were generally shown to perform better than those "online-based" ESM products;
- Opening doors to new applications;



## **Next Steps**

#### Further Application and Development

- Analyze the impacts of long-term soil moisture changes on above- and belowground C dynamics;
- Provide the initial and boundary conditions for atmospheric models;
- Assemble more in situ SM datasets and implement other advanced fusion algorithms;

#### > New Ecohydrology Working Group

- ✓ Leverage existing AmeriFlux/SOC-RUBISCO;
- Understand and benchmark the global SM dynamics using multi-source and multi-scale datasets;
- ✓ Improve existing SM databases and benchmarking methods;
- Find innovative ways to use benchmarking results to improve model parameterization, predictions, and projections.



# Thanks for Your Attention! Questions and Comments?

