

Global Downscaling of Remotely-Sensed Soil Moisture using Neural Networks

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Supplementary Materials

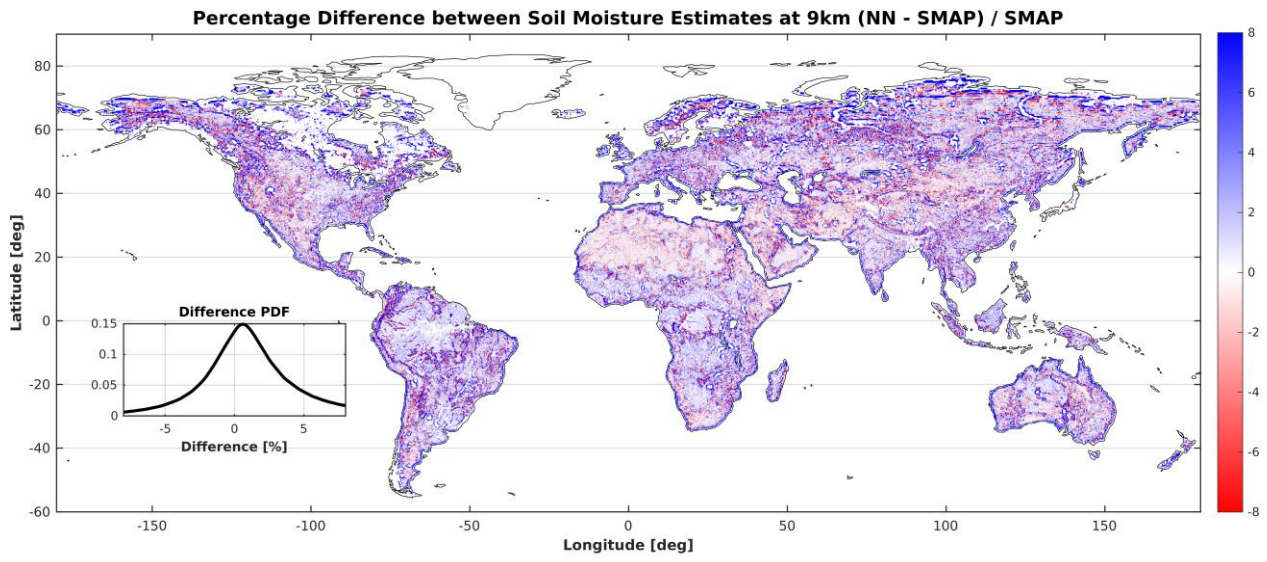


Figure S1- Percentage bias between the SMAP 9km soil moisture estimates and NN 9km estimates. White regions indicate no data.

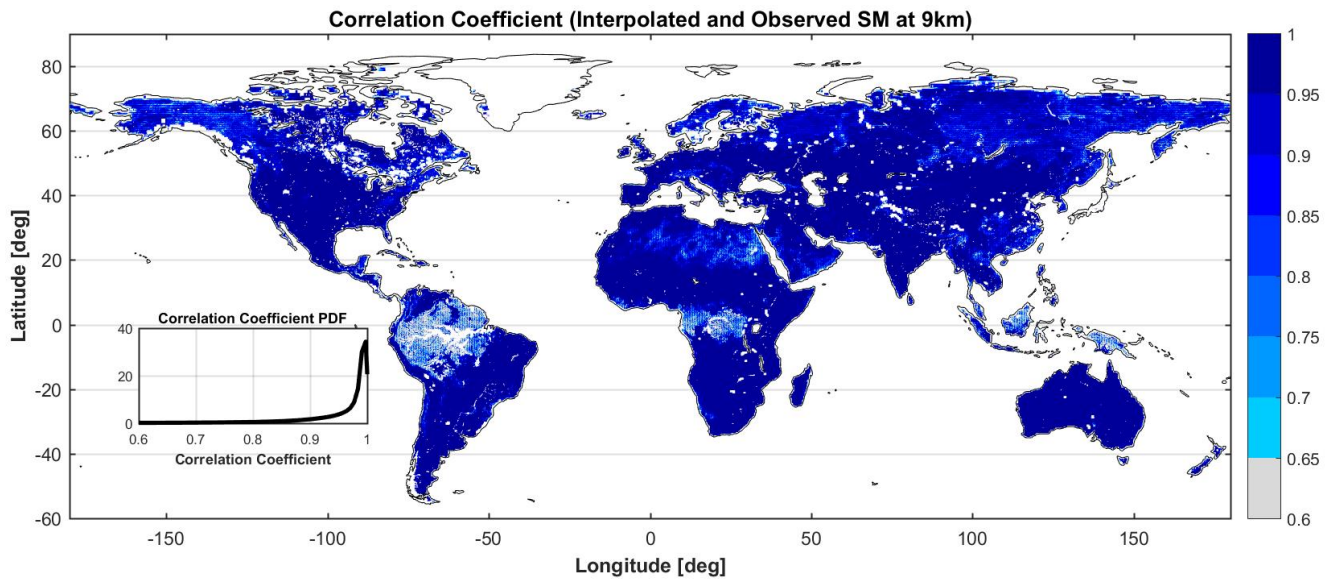


Figure S2- Correlation coefficient (R^2) between SMAP observed soil moisture at 9km and Interpolated soil moisture at 9km. White regions indicate no data.

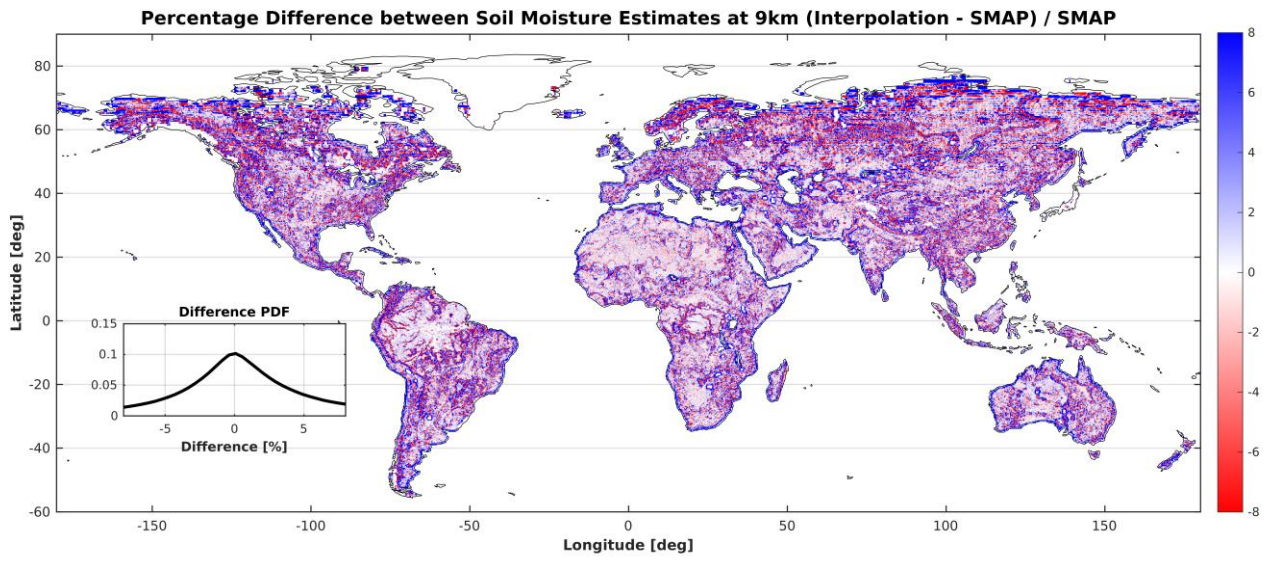


Figure S3- Percentage bias between the SMAP 9km soil moisture estimates and Interpolated 9km estimates. White regions indicate no data.

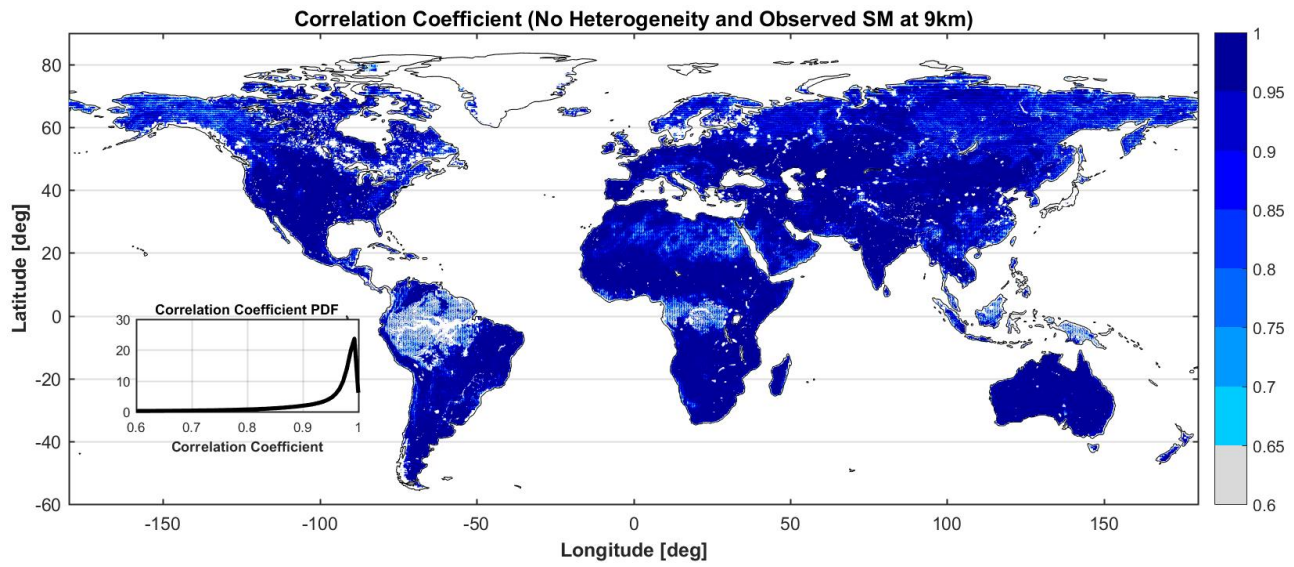


Figure S4- Correlation coefficient (R^2) between SMAP observed soil moisture at 9km and No Heterogeneity soil moisture estimates at 9km. White regions indicate no data.

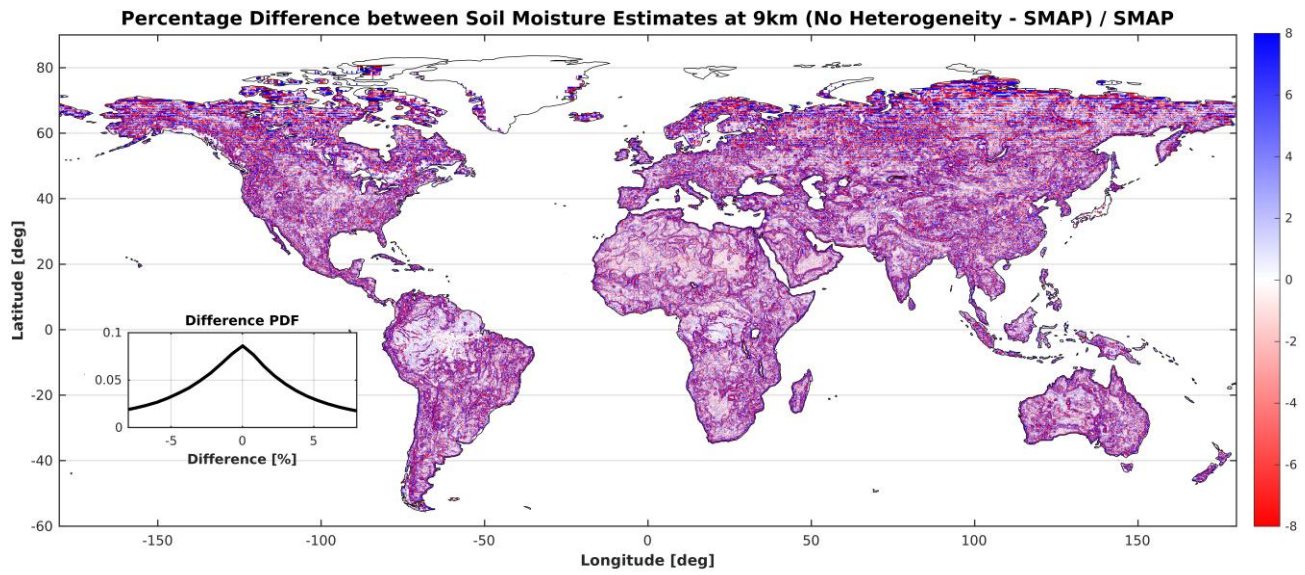


Figure S5- Percentage bias between the SMAP 9km soil moisture estimates and No Heterogeneity 9km estimates. White regions indicate no data.

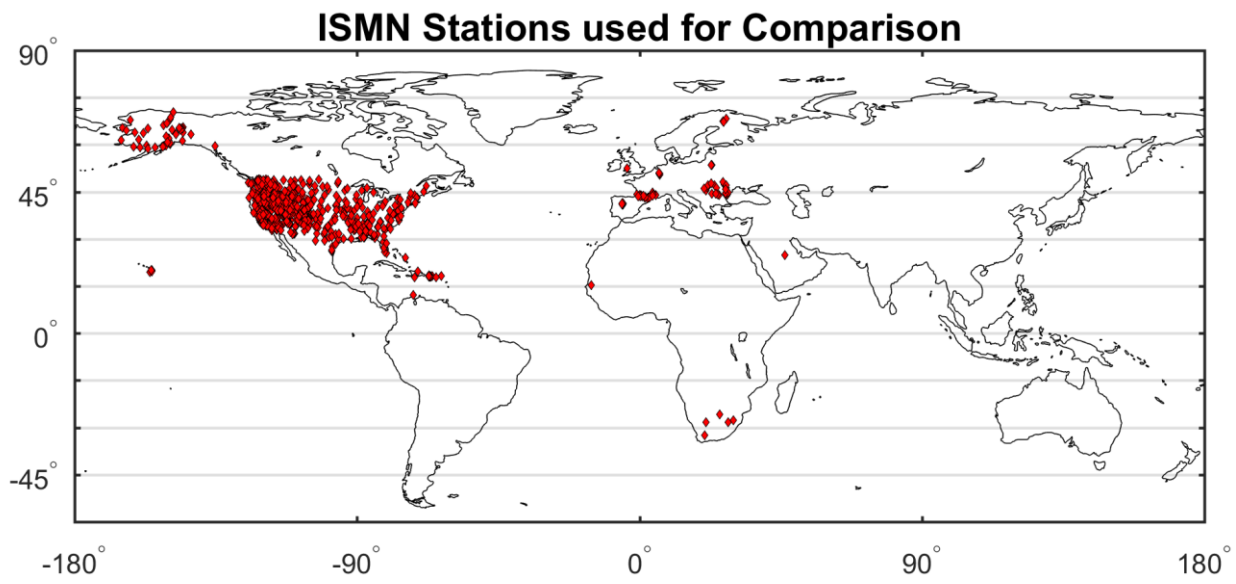


Figure S6- Location of each ISMN station used for comparison against downscaled soil moisture estimates