

Soil moisture retrieval from SMOS observations using neural networks

Summary Report of project SMOS+“Neural Networks” (SMOS+NN)

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