

Remote Sensing and Machine Learning Accurately Predict Soil Moisture Dynamics within an In-Service Evapotranspiration Disposal Cell Cover

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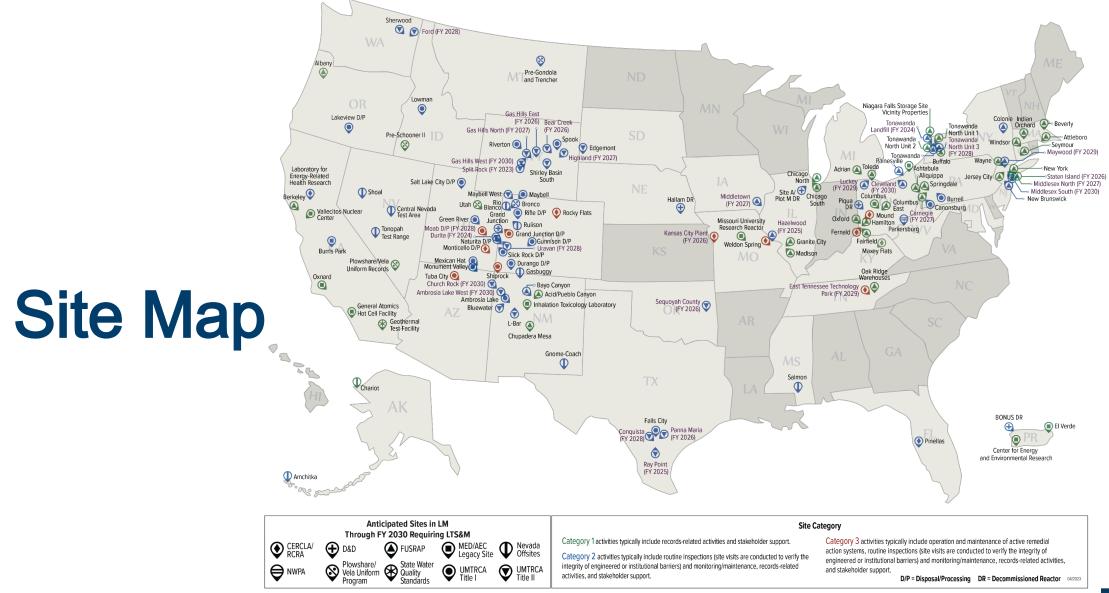


Mission



LM's mission is to fulfill the Department of Energy's postlosure responsibilities and ensure the future protection of human health and the environment.



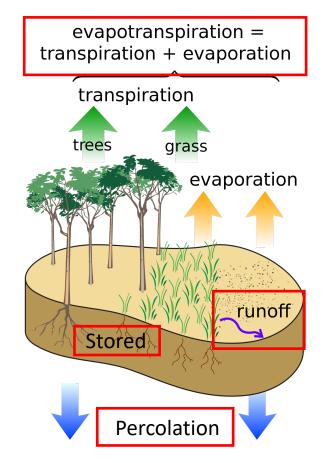




Background



- Conventional Uranium Mill Tailings Radiation Control Act covers include a "low-permeability radon barrier" to:
 - Limit surface flux of radon (NRCl989).
 - Protect groundwater by controlling percolation (DOE 1989).
- Natural ecological and soil-forming processes have changed asbuilt engineering properties of conventional covers.
 - No longer have lowpermeability, and radon diffusion and flux have increased (NRC 2011; NRC 2022)
- Investigating options for future management of selected conventional (unvegetated) covers as evapotranspiration (ET) covers
 - ET limits percolation
 - However, soil structure and drying by plants can increase radon diffusion and flux (NRC 2022)
 - Soil water content is a key performance parameter





Objectives



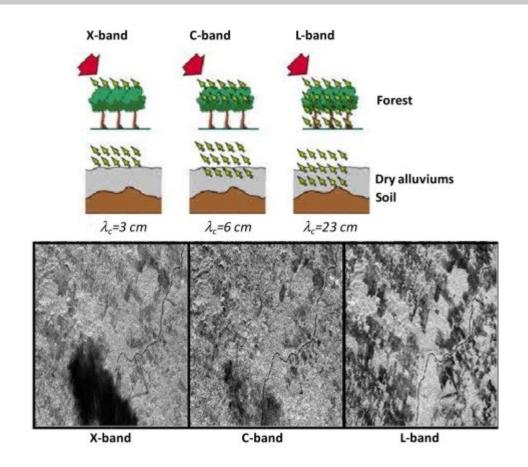
- Assess potential for direct soil moisture (SM) mapping using multifrequency synthetic aperture radar (SAR)
 - SAR sensitivity quantified through physical modeling of radar backscatter
 - P-, L-, C-, and Xbands
- Soil moisture profile mapping using multisource remote sensing
 - Test machine learning models using a fieldscale lysimeter embedded within an in-service disposal cell for model training and validation
 - Freely available satellite SAR and optical remote sensing used as predictors for 3D SM mapping
- Potential tool for maintaining regulatory compliance at ET cover disposal cells



SAR

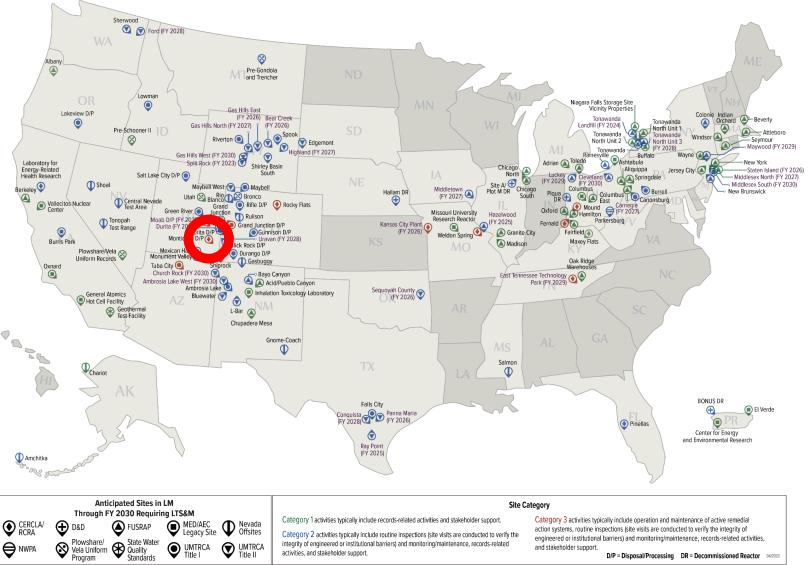


- Active imaging sensor
 - Emits pulses (bands) of microwave energy
 - Backscatter is used to identify/map features/materials on or below the earth's surface
 - ↑ frequency, ↓penetration
- Common bands
 - L and P (lower frequency)
 - C and X (higher frequency)
- Synthetic uses motion to achieve a finer spatial resolution
 - Mounted on moving platform (aircraft, satellite)
 - Sentinel-1A/B in this study





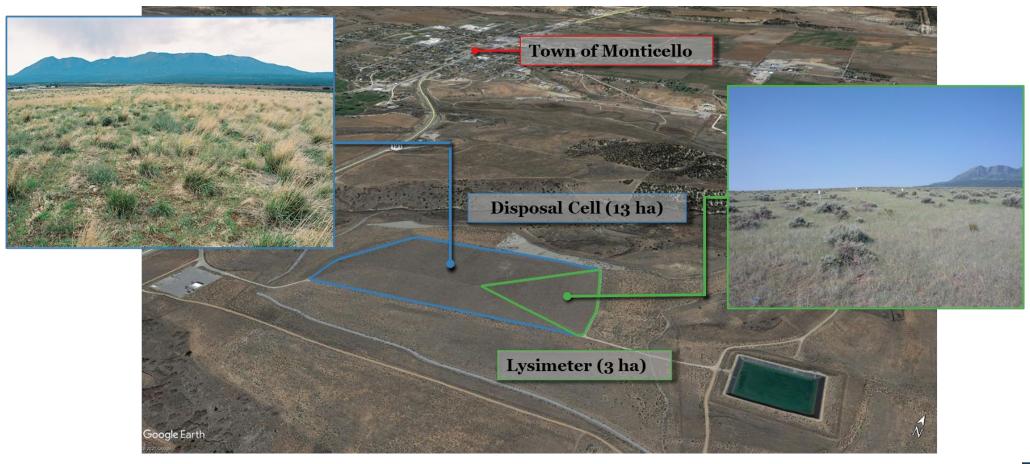
Study Site





Monticello, Utah, Disposal Site







Monticello, Utah, Disposal Site Lysimeter

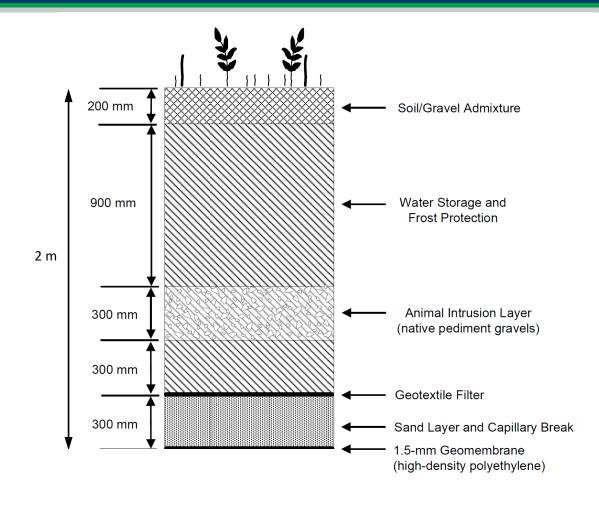


- Lysimetry
 - Allows us to quantify the components of the local water balance, where:
 - ET = precipitation percolation runoff Δ water storage.
- Unique opportunity:
 - World's largest drainage lysimeter (observable by space satellite)
 - 20+ years of data



Vertical Profile of Cover/Lysimeter

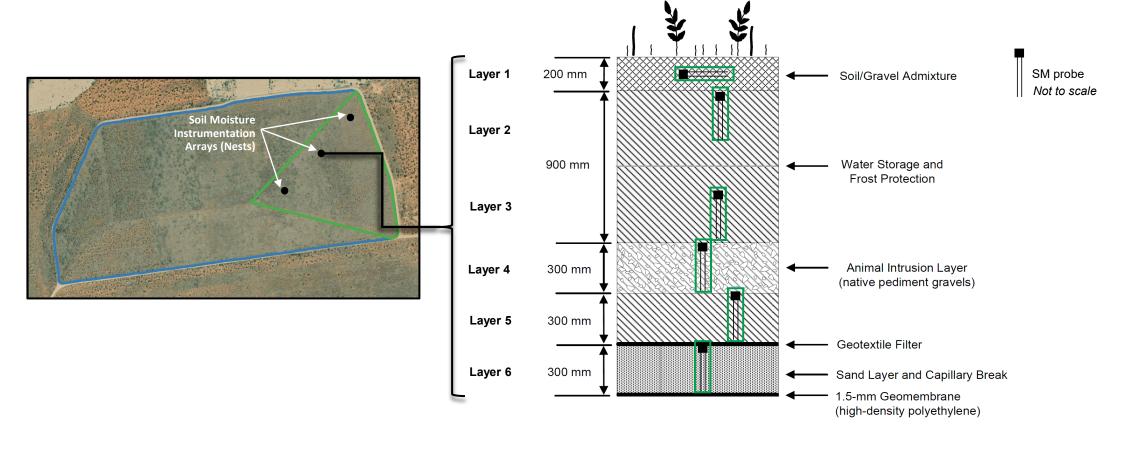






Soil Moisture Probes





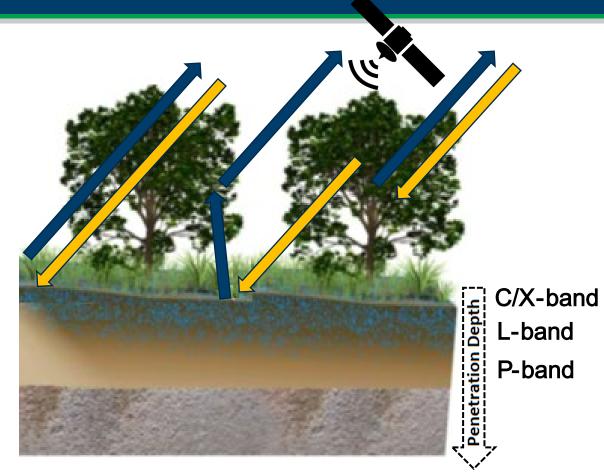


Methods – Radar Backscatter Modeling



SAR backscatter modeling:

- Radiative transfer model used for simulating multifrequency (P-, L-, C-, X-band) microwave scattering and soil moisture sensitivity (Du et al. 2015)
- Vegetation modeled as a collection of dielectric scatterers (e.g., discs for leaves, cylinders for stems)
- Soil scattering modelled using Advanced Integral Equation Method (AIEM)
- Multiple soil layers used to represent soil profile
- Model parameterized using site vegetation and soil data



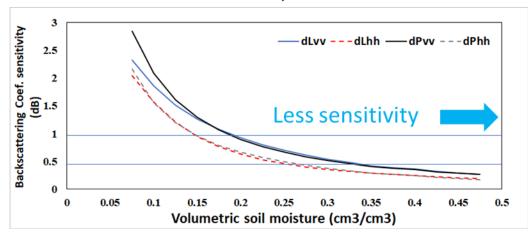
Radar backscatter consists of contributions from soil and vegetation and interactions between soil and canopy.

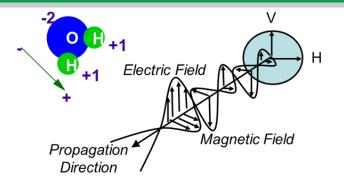
Results – Radar Backscatter Modeling



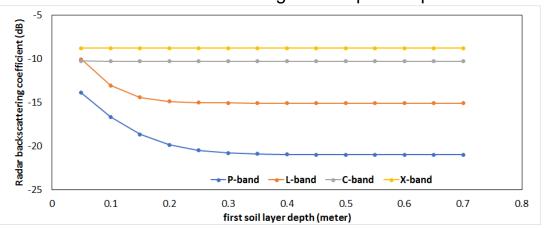
- Estimated multifrequency SAR sensitivity: P-band (0.43 GHz), L-band (1.27 GHz), C-band (5.404 GHz), X-band (9.6 GHz); incidence angle: 35 degrees; moderate shrubland vegetation cover
- Results confirm strong microwave sensitivity to soil moisture variability

Radar backscatter sensitivity to Soil Moisture increase





Radar backscatter change with topsoil depth*



*Two-layer soil: first layer depth from 5 cm to 70 cm, SM = 0.15 cm³/cm³;

second layer: SM = 0.5 cm³/cm³



Results – Radar Backscatter Modeling



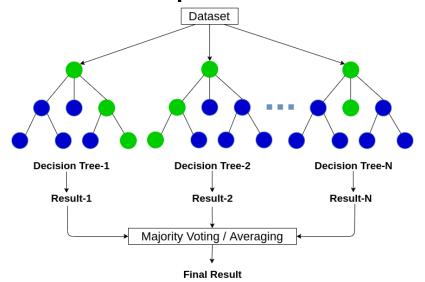
- Radar backscatter physical model established for simulating multifrequency radar sensitivity to soil moisture profile over semiarid shrubland
- Confirmed favorable SAR sensitivity over dry to moderate soil moisture levels
- All frequencies (C-, X-, L-, P-band) sensitive to similar SM dynamic range, but lower frequencies (L-, P-band) sensitive to deeper soil conditions
- Higher frequencies (C-, X-band) most useful for direct sensing of surface (0–1 cm) soil moisture and roughness
- Demonstrated capability of utilizing multiple SAR frequencies for mapping soil moisture profile



Methods – ML Models



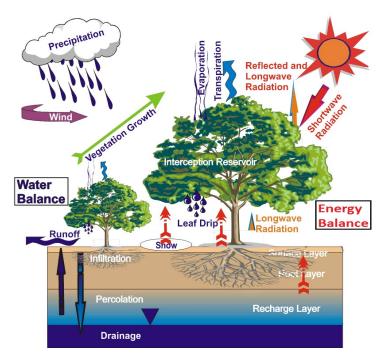
- Random forest (RF)
- A forest of randomly created decision trees. Each node in the decision tree works on a random subset of features to calculate the output.
 The random forest then combines the output of individual decision trees to generate the final output.





Methods – ML Models





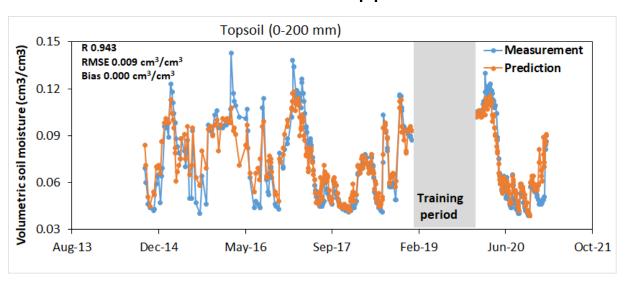
Soil moisture dynamics at different depths are governed by water and energy balance and are related to precipitation, ET, runoff, and soil properties and reflected by vegetation health.

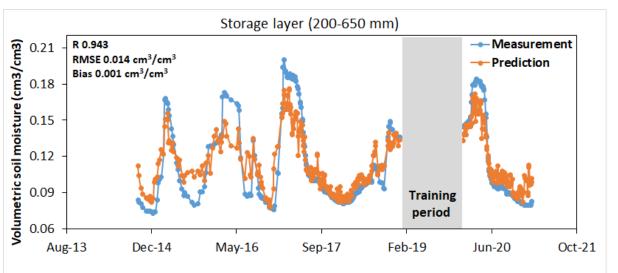
- ML models developed and tested for mapping soil moisture profile dynamics
- Predictors:
- Sentinel-1SAR C-band backscatter (30 m; 6- to 12-day repeat)
- Landsat normalized difference vegetation index (NDVI)
 (30 m; 16-day repeat)
- MODIS day/night (1:30 a.m./p.m.) land surface temperature (LST) difference (8-day)
- Accumulated precipitation (24-hour, 1-day; 1-, 2-, 3-, 4-, 5-week) measured by Daymet (Daily Surface Weather Data on a 1km Grid for North America)
- Target variable used for model training and validation: measured soil moisture
- Platform: GEE





- In situ measurements from 2019 were used for training an RF model
- The RF model was applied to the other years for a more comprehensive accuracy assessment
- The RF model was applied to the whole study area and all the years of data production

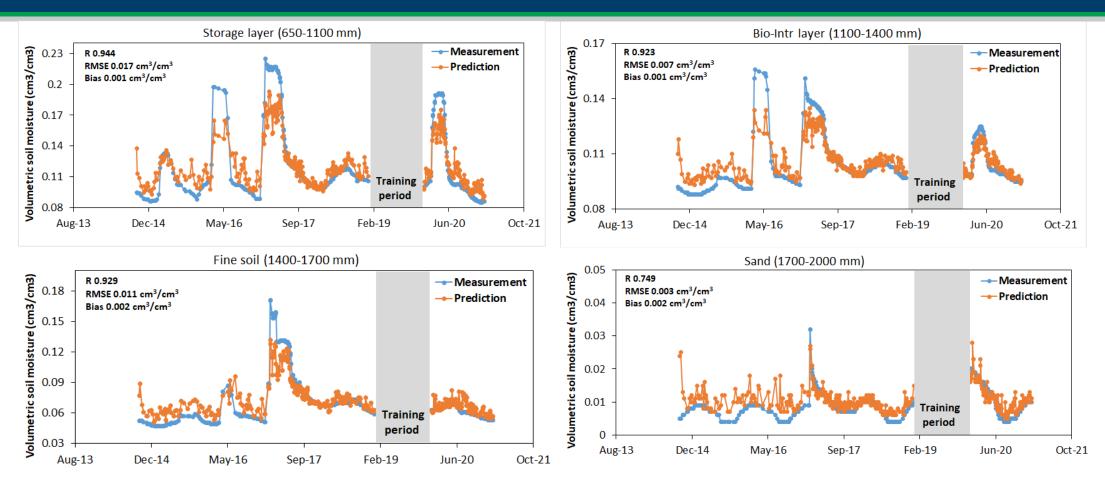




High correlations (R ~0.94) and low uncertainties (RMSE ~0.01 cm3/cm3; bias ~0.00 cm3/cm3) between measured and predicted soil moisture in the first two layers (0–650 mm).







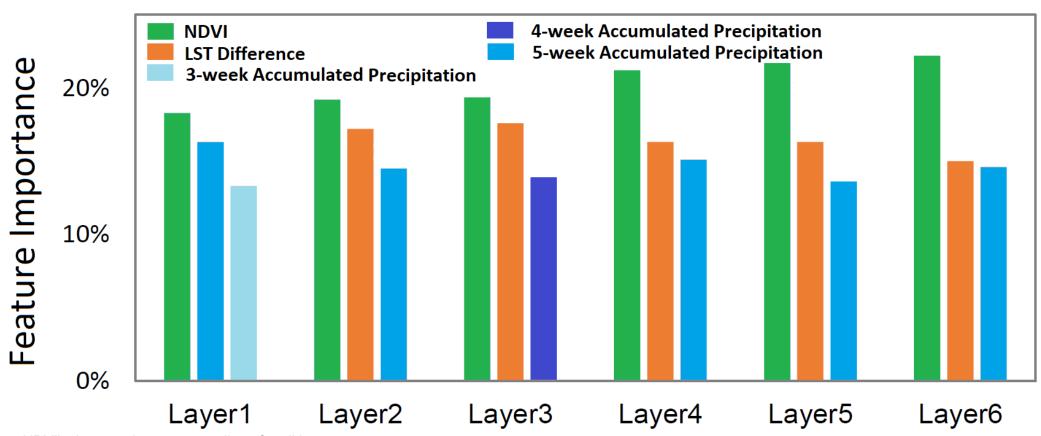
High correlations (R >0.92) and low uncertainties (RMSE <0.02 cm3/cm3; bias ~0.00 cm3/cm3) between measured and predicted soil moisture for deeper layers, except for the deepest (sand) layer.

(R 0.749) where soil moisture variation/magnitude is low.

In Partnership with Amentum and TFE

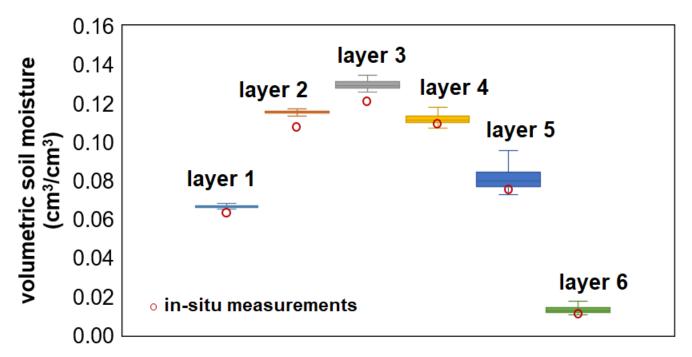
United for the Legacy Management Mission



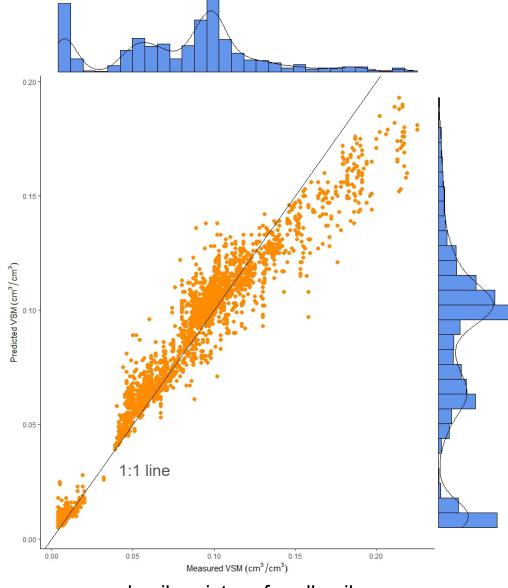


- NDVlis the most important predictor for all layers
- Accumulated precipitation and LST day-night difference are among the second and third most important factors





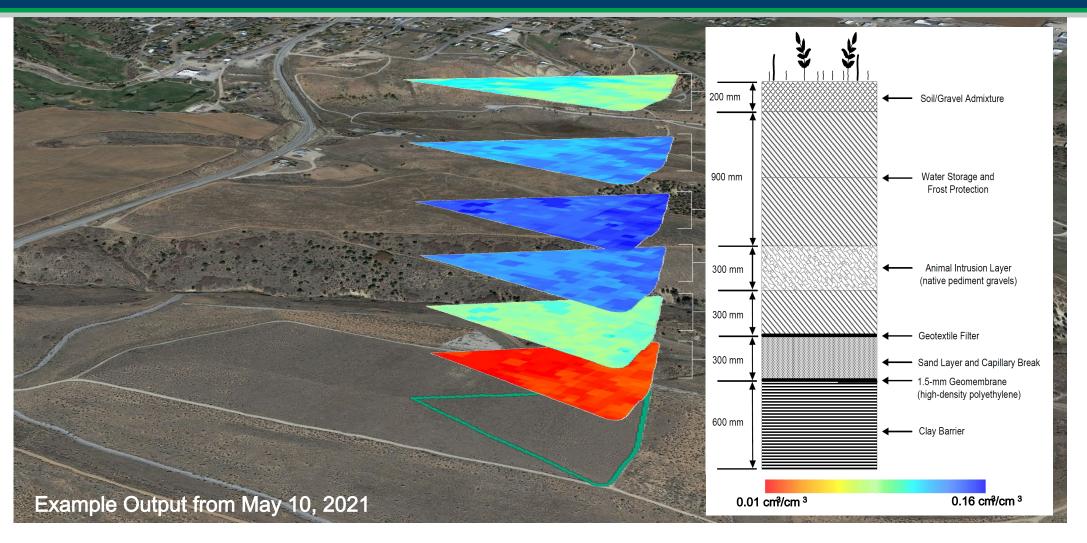
Soil moisture variability over the lysimeter for the growing season (May to September) from 2014 to 2021



United for the Legacy Management Mission

Predicted versus measured soil moisture for all soil layers and months/years (R = 0.97)







Summary



- Conventional ML (RF) models, driven by free satellite and other geospatial data, accurately predict 3D soil
 moisture profile (0–2 m) distribution and seasonal dynamics
 - Very low error/bias
 - Only 1 year of training data provided accurate multiyear (20-12-021) results
- Could be effective tool for addressing regulator recommendations on ET cover performance monitoring
- NDVIwas the most important predictor among all soil layers
 - Illustrates role of plants/rooting characteristics in moderating dryland SM dynamics
 - May expand utility of ML approach to vegetated, rock armored covers where SAR cannot penetrate
- ML results suggestconventional probe measurementsmay not adequately account for SM heterogeneity across horizontal space (layers 43)—additional SM arrayswill help verify
- Effect of direct low frequency (L-band) GNSS radar measurements TBD





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