Towards improved TDR soil water sensing for optimizing irrigation water management

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TDR Instrumentation – Historical Perspective

- 1990's and early 2000's: Tektronix 1502C cable tester (1.75 GHz bandwidth), CPU, 50 Ω coaxial cables, multiplexers...
 - Calibration influenced by coaxial cable length and interconnects
 - Extremely difficult to deploy in field
 - Only for scientific research





- 2016 to present: Acclima TDR-315 sensors (~1 GHz bandwidth)
 - Miniaturized TDR circuit
 - SDI-12 interface; waveforms acquired via another communication protocol
 - Solar gateway and nodes allow cloud access

Why TDR?

- Reduced sensitivity to apparent (bulk) EC of the soil
- Measured travel time is highly linear with soil water content





 Captured waveforms contain patterns and features that reflect soil properties

TDR measurements – sources of imprecision/inaccuracy



- Small sampling volume of probe (~300 cm³)
- Sensitivity to bulk EC increases with increasing 2:1 clay content
- Temperature influences dielectric and dc losses and the amount of bound water

- Errors 0.02 to 0.05 m³ m⁻³ (20 50% management range)
- Calibrations fail to consider interaction between mineral surfaces and water (bound water relaxation):
 - Permittivity of bound water 10 60% of free water
 - Attenuation of high frequency components biases graphical determination of travel time



Calibrations (loosely) based on theory

 Interplay between temperature and bound water (Or and Wraith, 1999) and bulk electrical conductivity (ECa) (Schwartz et al., 2009ab) – *Complex dielectric power law mixing model*

$$\varepsilon^{*} = \varepsilon' - j\varepsilon'' - j\frac{\sigma_{dc}}{\omega\varepsilon_{0}}$$

$$\varepsilon^{*} = \left[V_{air} \cdot \varepsilon^{\alpha}_{air} + V_{min} \cdot \varepsilon^{\alpha}_{min} + V_{bw}(T) \cdot \varepsilon^{*}_{bw}(T)^{\alpha} + V_{fw}(T) \cdot \varepsilon^{*}_{fw}(T)^{\alpha}\right]^{\frac{1}{\alpha}} - j\frac{\sigma_{dc}}{\omega\varepsilon_{0}}$$

- Air & Mineral permittivities are real valued
- Free water & bound water permittivities are complex and frequency and temperature dependent

Fitted Parameters: As – surface area of soil α – power law exponent

Advantages

- Improved RMSE
- Fitted surface area closely correlated with measured surface area (r = 0.989)

Disadvantages

- Not a closed form calculation
- Travel time approximated from signal attenuation ≠ increases in graphically determined travel time
- Soil specific calibration is impractical for most field applications for irrigation management (large fields, spatially variable soil properties)

Can we use waveform features to apply soil specific calibrations?

- Big Dirta Project / Acclima, Inc.
- Generation of waveform dataset with the TDR-315N
 - Mixtures of sand, kaolinite, high surface area soil (Pullman Bt) – range in specific surface areas
 - $^\circ~$ Saturation with < 1 dS m $^{-1}$ CaCl $_2$
 - Pressure extraction in chamber (to 100 kPa)
 - Resaturation with progressively increasing EC solutions
- Work in progress



Pressure extraction in packed container

Can we use waveform features to apply soil specific calibrations?

• Frequency response of soils reflect bound water associated with surfaces (amount of 2:1 clay)



- Identify waveform features useful for soil specific calibration (machine learning approach)
- Sensors would "self-calibrate" in situ
- Collaborative work with Acclima, Inc.

Can we use waveform features to apply soil specific calibrations?

Waveforms with similar travel times and EC_a (~1 dS m⁻¹)

- Kaolinite 70% clay (EPK, Edgar Minerals, FL); As = 52 m² g⁻¹
- Pullman Bt 50% clay (25% smectite, 25% mica); As = 190 m² g⁻¹



Using TDR to quantify net irrigation and infiltration

- Large uncertainties in infiltration/runoff and irrigation application efficiency
 - Generates errors in estimated crop water use 0
- Use of event based method to estimate infiltration / net irrigation



Maize (Bushland, TX 2022)

Array of sensors near surface (detection of wetting front)



- Integration permits evaluation of the net irrigation (24 mm for 3 arrays; 93% AE)
 - Wetting front does not extend to lowest sensor 0
 - Negligible drainage for short time period 0

IRT canopy temperatures as a proxy to scale measured crop ET

 Major limitation: Impractical & costly to deploy sufficient sensors to account for water content variability in large agricultural fields





- ETc calculated using water balance and change in stored soil water with TDR
- Use of IRT crop canopy temperature to infer crop ET using the a scaled crop water stress index (CWSI)

Challenges & Opportunities – soil water sensors for irrigation management

• Water content measurement accuracy is still a problem for all EM sensors in soils with significant 2:1 clays

TDR down-hole prototype Acclima, Inc.



Improvement of **downhole probes for deep soil water content sensing** (problems: air gaps in installation; reduced sensitivity)

Subsurface drip lines



- Inferring water balance/crop water requirements with EM sensors under the spatial variability imposed by surface and subsurface drip irrigation and resulting root densities
- Field variability and upscaling measured water contents to management zones
- Scarce water resources -> irrigating to avoid crop water stress may not result in the greatest water productivity
 - Fusion of crop models & soil water content measurements to improve forecasting of yield and profitability in response to irrigation



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