#### Soil Moisture Measurements and Water Availability Index Derivation Using Remote Sensing Images

presented by

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# Why do we need to care about soil moisture?



Hydrological Cycle

#### **Outline of Presentation**

- Study Objectives
- Rationale of Remote Sensing for Soil Moisture Measurements
  - Study Area: Watershed Environment
- Field Efforts: Satellite Image Acquisition and Ground Truthing
- Modeling Process and Soil Moisture Mapping
- Future Work

#### **Study Objectives**

- Understand the rationale of space borne remote sensing for soil moisture measurement
- Validate microwave soil moisture retrieval algorithms for an existing microwave sensor systems: RADARSAT-1.
- Integrate satellite remote sensing with genetic programming model to predict the soil moisture distribution in a semi-arid watershed.

#### THE ELECTROMAGNETIC SPECTRUM



### **Electromagnetic Spectrum**



# Satellite Hydrology Geostationary Operational Environmental Satellites (GOES) Polar Orbiting Environmental Satellites (POES).



#### **Remote Sensing for Soil Moisture Measurement**

- Active measurement: A microwave pulse (radar) is sent and the power of received signal is compared to that which was sent to determine the backscattering coefficient.
  - Passive measurement: Natural thermal emission of land surface (or brightness temperature) is measured at microwave frequencies.







# **Synthetic Aperture Radar**

SAR systems take advantage of
 the long-range propagation characteristics of radar signals and
 the complex information processing

capability of modern digital electronics to provide high resolution imagery.

### **RADASAT-1 SAR Satellite**

- When using a space-borne SAR satellite with active microwave sensor, the radar backscatter is sensitive to:
  - Water content in the surface soil
  - Surface roughness and vegetation cover
  - Angle of incidence
  - Surface slope

This exhibits a potential to measure surface soil moisture

### Study Area: Choke Canyon Reservoir Watershed, South Texas



### **Differences of Elevation from 740 m to 40 m**





# Annual Rainfall (inch per year)



### **Nueces River Basin Aquifers**











#### **The Corner Reflector**

- An aluminum trihedral with the open side facing toward the SAR sensor.
- The CR is shown as a white pixel in SAR image because of the well return of the backscatter signal.





#### **Location of the CRs**

- Five corner reflectors were installed in the CCRW prior to SAR data acquisitions in April 2004.
   Four of them falls into one scene of SAR image.
  - Real-world coordinates of each CR were acquired using a sub-meter accuracy GPS unit.





#### **Ground-Truth : Sensor Technology**





#### Adapted from Time domain reflectometry (TDR) web



Adapted from HOBO web



#### **SAR Imagery Basin Wide**

![](_page_27_Figure_1.jpeg)

#### **SAR Data Calibrations**

According to Alaska Satellite Facility (ASF)\*, the ERS-1 and -2 had their absolute location accuracy of 230 m and 252 m, respectively.

• This study achieves 5 m horizontal accuracy.

\*Alaska Satellite Facility, "ASF Interferometric SAR Processor (AISP) Calibration Report, version 4.0"

### **Soil Moisture Prediction Techniques**

Simple Linear Regression
Multiple Linear Regression
Nonlinear Regression
Neural Networks
Genetic Programming

### **Genetic Algorithm (GA)**

- It is a probabilistic search algorithm that iteratively transforms a set (population) of mathematical objects, each with an associated fitness value, into a new population offspring objects using
  - the Darwinian principle of nature selection
  - The operations that naturally occurring in genetic operations such as crossover, mutation, and reproduction.

Ref: Koza, J.R., Genetic Programming IV. Stanford University, CA. E-mail: koza@stanford.edu

#### **Genetic Programming (GP)**

GP applies the approach of the Genetic Algorithm to the space of symbolic regression problems
Genetic Operations

Reproduction
Crossover
Mutation

## **Animation: Crossover**

![](_page_32_Picture_1.jpeg)

![](_page_32_Picture_2.jpeg)

# **Animation: Mutation**

![](_page_33_Picture_1.jpeg)

![](_page_33_Picture_2.jpeg)

# **Model Formulation**

Assumption:  $VMC = fn(\sigma_0, \phi, \alpha, C, A)$ VMC = volumetric moisture content in measuring with the TDR 300 probe (%) = SAR data in decibel (decibel)  $\sigma_{_0}$ = percent slope (%)  $\phi$  $\alpha$  = Aspect (slope direction) C = Land cover A = Soil type

#### **Results and Discussion**

Model calibration with the training data

Name	Approach	Model	R <sup>2</sup>	RMSE
Model 1	Linear Regression	$VMC = -4.712 \sigma_0 - 13.067$	0.10	20.2
Model 2	Multiple Regression	$VMC = -11.11 b_0 + 3.17 $ $\phi - 0.889 \alpha - 198867$	0.15	44.23
Model 3	GP		0.83	10.72
$VMC \ (\%) = \left\{ 3 * \left[ \frac{\sin \left( \cos \left( INC \right) \right) * Sigma}{0.9177978} + 1.531 \right] \right\} - \left( 7 * SLOPE \right) - \left( 3 * Sigma \right) + INC \right\}$				
Model verification with the unseen data				

#### **Soil Moisture Mapping in Sep., 2004**

![](_page_36_Figure_1.jpeg)

#### **Agricultural Area**

Surface Soil Moisture Prediction

SAR Imagery (August 1, 2003)

![](_page_37_Figure_3.jpeg)

#### **Forest / Grassland**

Surface Soil Moisture Prediction

SAR Imagery (August 1, 2003)

![](_page_38_Figure_3.jpeg)

## Hill / High Slope

#### Surface Soil Moisture Prediction

![](_page_39_Figure_2.jpeg)

![](_page_39_Figure_3.jpeg)

SAR Imagery (August 1, 2003)

![](_page_39_Figure_5.jpeg)

1:250,000 0 3.75 7.5 15 Kilometers

#### **Extended Work:**

- Water cycle analysis
- Carbon cycle analysis
- Modeling coupled water and couple cycles
- Meteorological model
- Ground penetration radar
- Riparian buffer zone change detection

#### Multiscale Water Infrastructure Characterization

#### Multi-disciplinary approach for practical solutions

Spatial and temporal GIS analysis of water supply availability, future supply-demand imbalance, and impacts on water quality and ecological systems

![](_page_41_Picture_3.jpeg)

Remote sensing and satellite imagery for spatial assessment of drinking water source quality and quantity, and evaluation of program effectiveness and outcomes

![](_page_41_Picture_5.jpeg)

Water utility infrastructure conditions and SDWA compliance assessment under predicted future global change scenarios (climate, demographic and economic)

Regional analysis on water and wastewater infrastructure sustainability. Examples:

- CSO/SSO (eastern US, gulf states)
- -Salt-water related pipe corrosion (FL, east and west coasts)
- Water reuse and allocation in ecological and human consumption (CA, TX, AZ, FL, PR, and other Plain states)

#### Water Budget

![](_page_42_Figure_1.jpeg)

(WATER INFLOW)-(WATER OUTFLOW)=(CHANGE IN WATER STORAGE)

#### Typical water budget components

#### WATER INFLOW

- -Precipitation
- -Surface-water flow into basin
- Imported water
- -Ground-water inflow

#### WATER OUTFLOW

- -Evaporation
- -Transpiration by vegetation (evapotranspiration)
- -Surface-water outflow
- -Exported water
- -Ground-water outflow

#### CHANGE IN WATER STORAGE, increased/decreased water in:

- -Snowpack
- -Streams, rivers, reservoirs
- -Aquifers

![](_page_43_Figure_0.jpeg)

# NEXRAD

# National Doppler Radar Network Provide estimation of rainfall region wide

![](_page_44_Picture_2.jpeg)

# Thank You

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