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Spatially Constrained Geospatial Data Clustering for Multilayer Sensor-Based Measurements

N.M. Dhawale
V.I. Adamchuk
S. O. Prasher
 Department of Bioresource Engineering
 McGill University

P. R. L. Dutilleul
 Department of Plant Science
 McGill University

R.B. Ferguson
 Department of Agronomy and Horticulture
 University of Nebraska-Lincoln


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Precision Agriculture and Sensor Systems Team

- Development of Proximal Soil and Plant Sensing Systems
- Geospatial Data Processing and Management
- Practical Implementation of Precision Agriculture

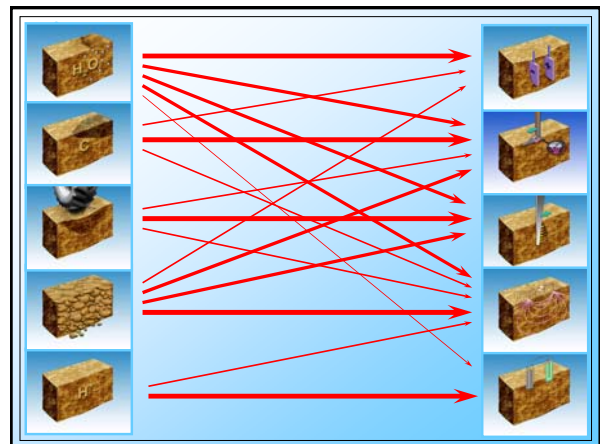



Proximal Soil Sensing

Proximal Soil Sensing (PSS) is a set of technologies developed to measure the physical, chemical and biological properties of soil when placing the sensor in contact with, or at a proximal distance (less than 2 m) to, the soil being characterized

operation	measurement	energy	inference
mobile	non invasive	passive	indirect
static	invasive	active	direct
PROXIMAL SOIL SENSING			
	in situ / ex situ		

Advances in Agronomy 113: 237-283



Sensor Fusion

Veris® MSP3

Soil EC and Field Elevation Mapping

Data Integration

Aerial Imagery System

Site Specific Data Calibration

$$OF = \sqrt[5]{S_{opt} \cdot D_{opt-pH} \cdot D_{opt-EC} \cdot H_{cr-pH} \cdot H_{cr-EC}}$$

- S-optimality
- D-optimality (soil pH)
- D-optimality (soil EC)
- H-criteria (soil pH)
- H-criteria (soil EC)

Geoderma 163:63-73

Semi-Automatic Soil Sampling

The image shows a red pickup truck equipped with a semi-automatic soil sampling rig. A person is shown in the driver's seat operating a handheld device that displays '39' and 'F8'. The rig is mounted on the back of the truck and is used for soil sampling in a field.

Soil Profiling Sensors

The image shows a vertical soil profiling sensor rig. A graph displays soil profile data, showing a curve that starts high and then levels off. The rig is used for soil profiling in a field.

On-the-Spot Analyzer

The image shows a green tractor-mounted soil analyzer. A detailed view of the sensor is shown, labeled 'SENSOR SCALE 1:8'. A scatter plot shows the relationship between Predicted NO_3^- (mg L⁻¹) and Observed NO_3^- (mg L⁻¹), with $R^2 = 0.87$.

Wireless Sensor Networks

The image shows a map of a field with sensor locations. A network diagram shows the connections between the sensors. A final map shows the real-time water stress impact assessment, with a color-coded scale from blue (low stress) to red (high stress).

Present Tools

- K-means clustering
- Distance matrix is based on data
- No account for spatial distances
- The results depends on the selection of initial centroids
- No repeatability
- Need for cross-validation
- Complexity
- Frequently occurring discontinuities

Neighbourhood Search Analysis

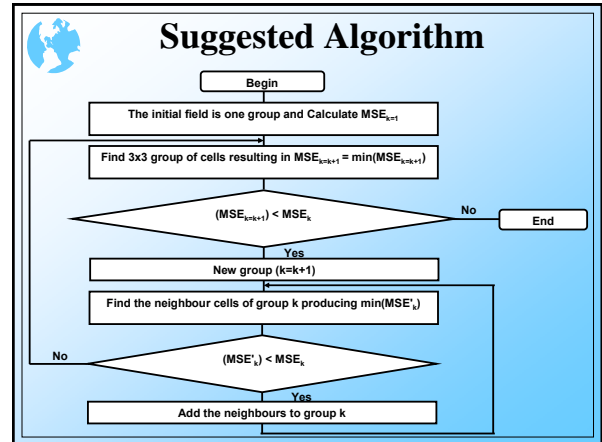
- Every delineated area must be contiguous
- Not every part of the field needs to be delineated
- The rest of the field is the rest of the field
- Small areas should not be considered
- Minimum variability within delineated areas
- Multiple layers with no artificial weighting

Performance Indicators

$$MSE = \frac{\sum_{j=1}^k \sum_{i=1}^{n_j} (X_{ij} - \bar{X}_j)^2}{N}$$

$$R^2 = 1 - \frac{MSE}{MSE_{k=1}}$$

X_{ij} = sensor-value for i^{th} grid cell within j^{th} group
 \bar{X}_j = sensor-value average for j^{th} group
 k = the number of grid cell groups
 n_j = the number of grid cells within j^{th} group
 N = the total number of non-zero grid cells

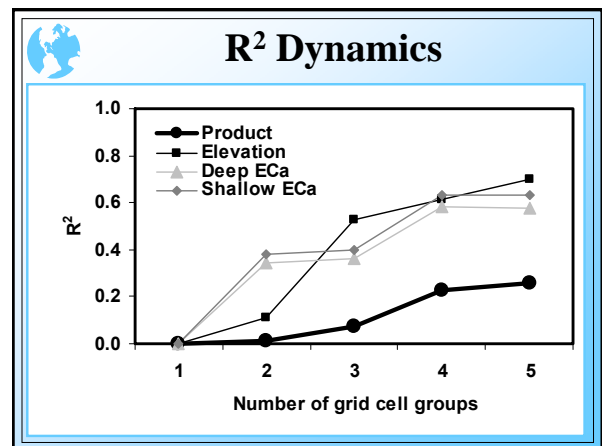
$$R^2_{Product} = R^2_{ShallowECa} \cdot R^2_{DeepECa} \cdot (R^2_{Elevation})^2$$


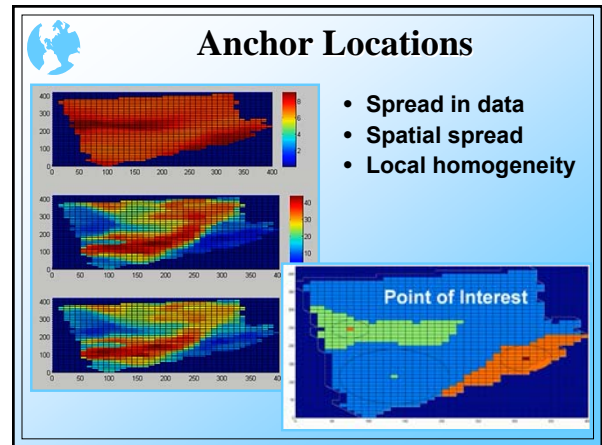
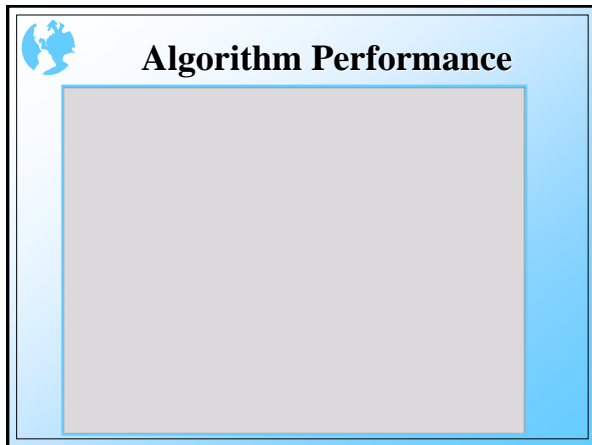
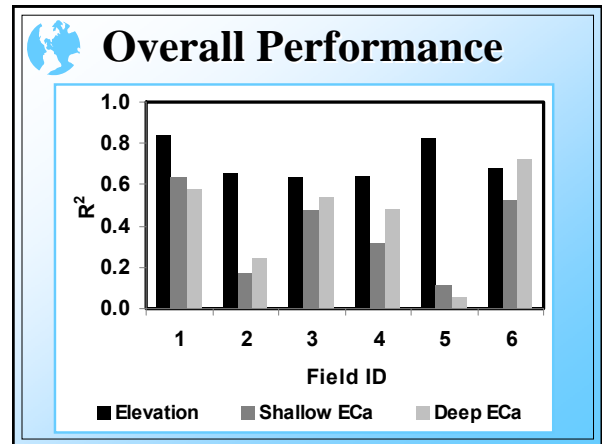
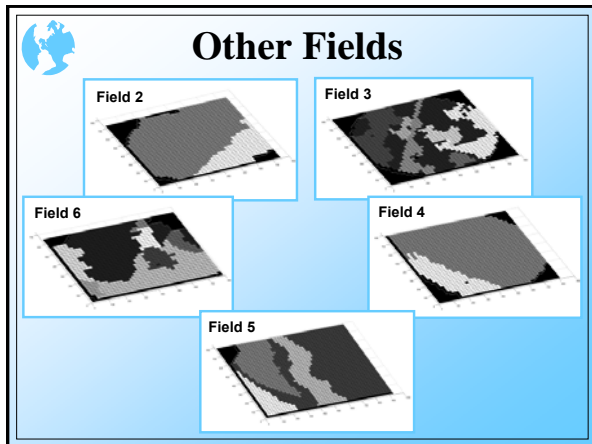
Algorithm Evaluation

Summary of Sensor Data

Field ID	Area, ha	Mean	Range	SD
		Field elevation, m		
1	25	1.50	3.20	0.53
2	46	4.95	17.82	3.99
3	50	7.10	11.54	2.09
4	55	8.07	27.44	5.68
5	67	4.22	8.09	1.60
6	44	6.15	10.59	2.15
		Shallow ECa, mSm ⁻¹		
1	25	0.73	1.58	0.28
2	46	3.99	13.14	1.67
3	50	6.21	11.64	1.84
4	55	2.44	9.04	1.72
5	67	7.25	9.32	1.88
6	44	2.29	7.42	0.82
		Deep ECa, mSm ⁻¹		
1	25	7.62	27.66	3.76
2	46	30.24	86.90	14.39
3	50	4.10	8.68	1.71
4	55	16.31	61.97	12.06
5	67	51.01	80.77	14.07
6	44	25.72	81.74	14.36

Neighbourhood Search Clustering





Summary

The spatial clustering algorithm developed in this study is based on a neighbourhood search method and seeks to minimize variance inside each group of interpolated grid pixels corresponding to an unlimited number of sensor-based data layers.

Preliminary tests of the algorithm using six production fields illustrated algorithm robustness when delineating field areas with different field elevations and soil EC_a measurements.

Each spatially constrained group of grid cells with the exception of the first group designated as “the rest of the field” emerged in response to every unique combination of data values relatively constant within each group.

<http://adamchukpa.mcgill.ca>
 E:mail: viacheslav.adamchuk@mcgill.ca