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Case Studies on the Accuracy of Soil pH and Lime Requirement Maps

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Problem Statement

- The assessment of soil variability is one of the most important steps in site-specific management
- Conventional means to attain soil variability data are incapable of accurately identifying spatial inconsistency within a production field at an economically feasible cost
- There is a need to develop equipment for mapping soil attributes on-the-go

Automated Soil Testing

Shank
Soil cutters
Coring tube

Purdue University
(West Lafayette, Indiana)
1996

Mixing
ISFET Electrode
Water jet
Sample collection for calibration
Cleaning
Add 20 ml DI H₂O
~50 g soil core

Automated Soil pH Mapping Systems

Purdue University
(West Lafayette, Indiana)
1998

US Patent No. 6,356,830

Field Mapping

RTK-level dual-system GNSS receiver

Soil pH mapping unit

Galvanic contact apparent electrical conductivity mapping unit

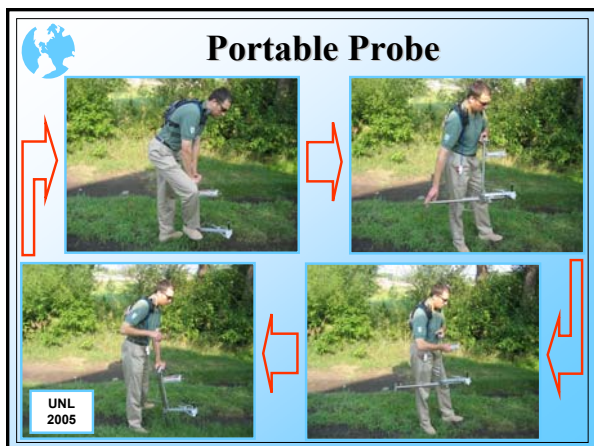
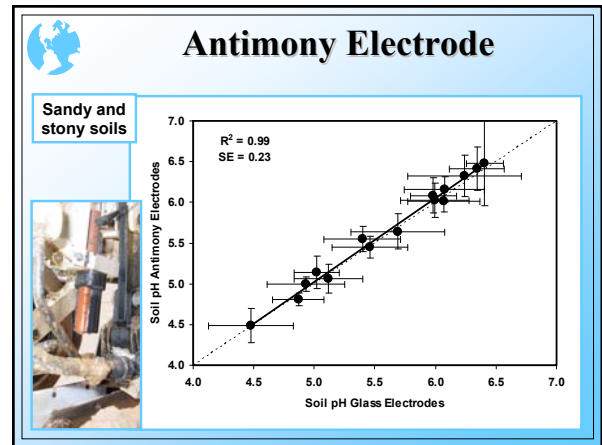
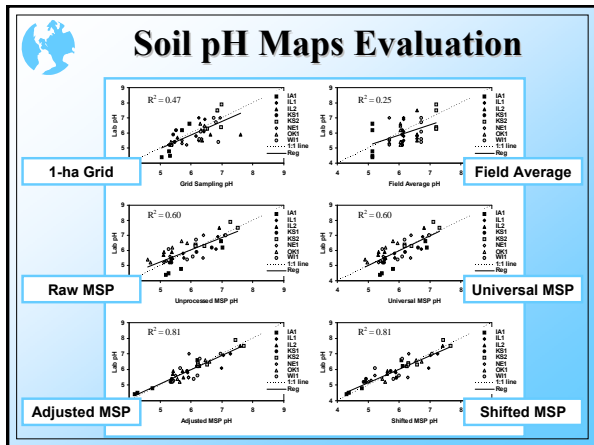
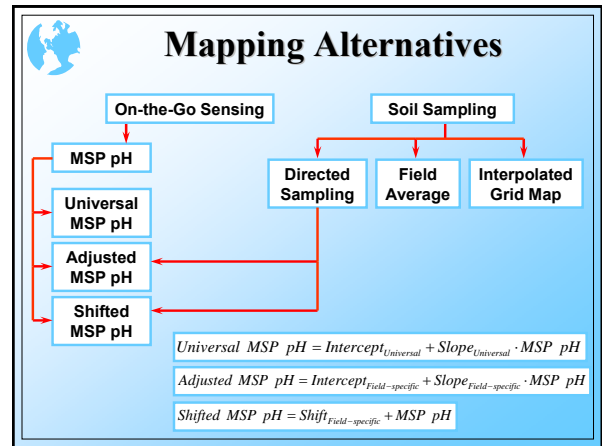
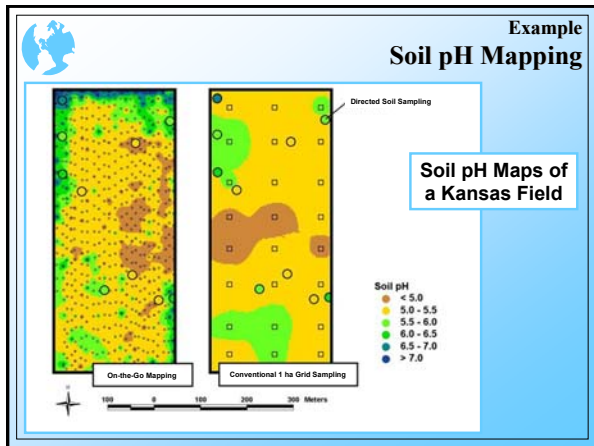
Direct Soil Measurement

Purdue University
(West Lafayette, Indiana)
Veris Technologies, Inc.
(Salina, Kansas)
UNL (Lincoln, Nebraska)

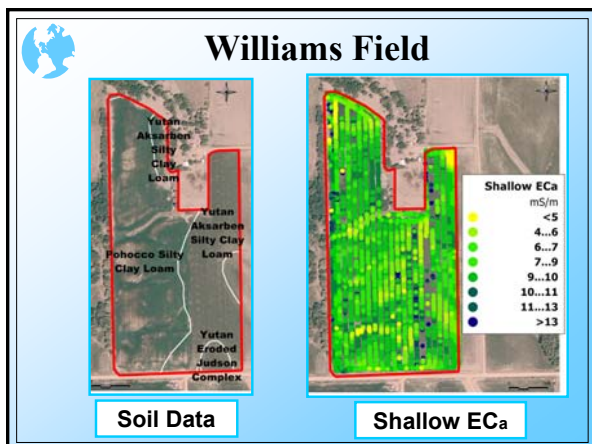
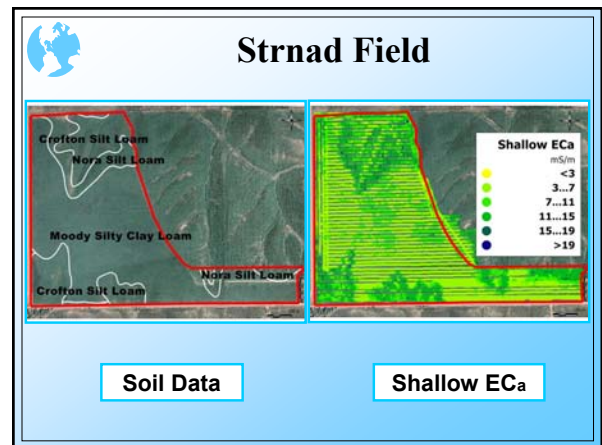
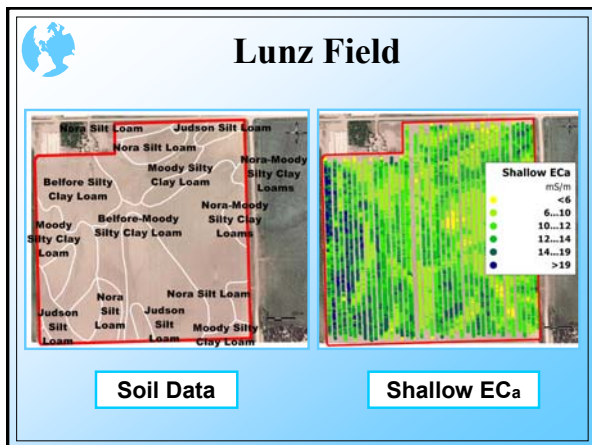
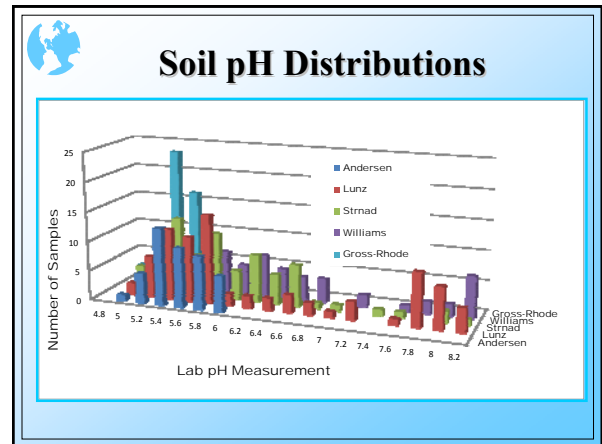
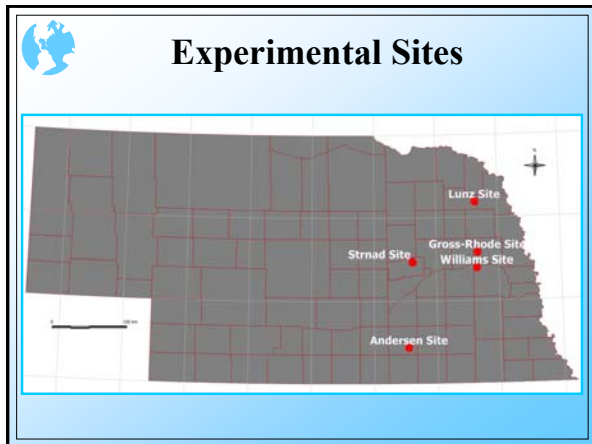
Ion-selective Electrodes

Water Nozzle

Soil Sampler

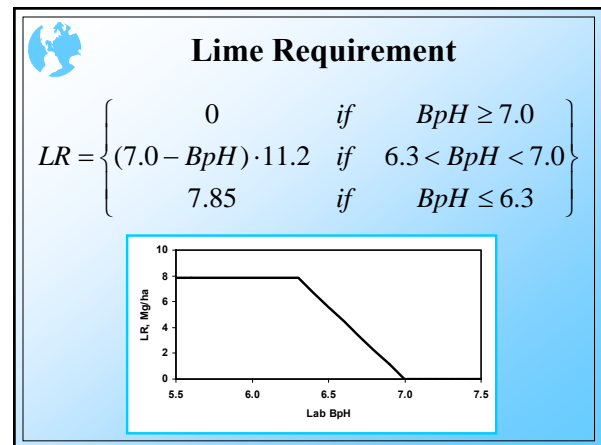
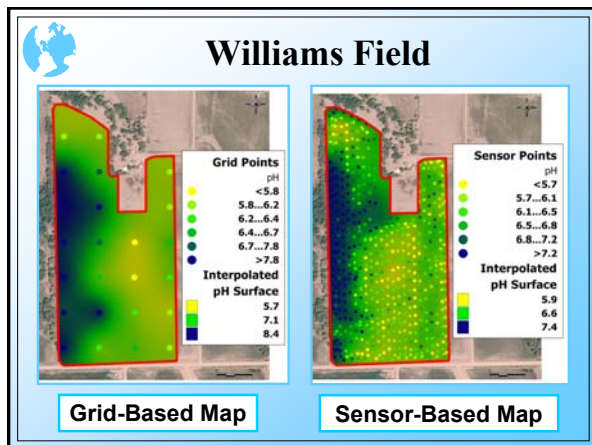
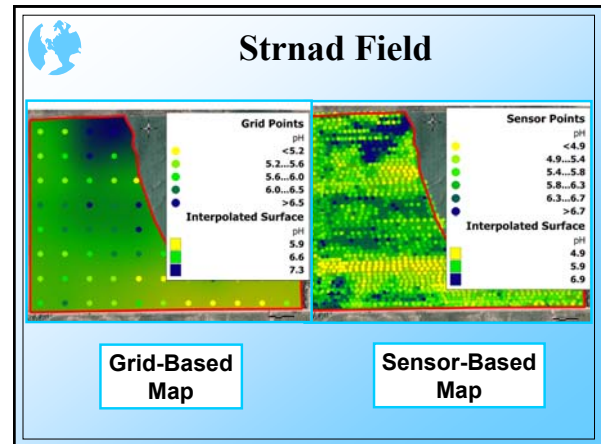
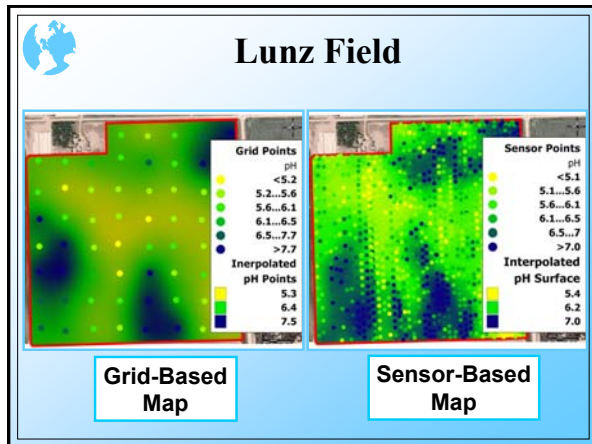


- ### Objectives
- To generate lime requirement maps using different mapping approaches
 - Average application rate
 - 1-ha grid sampling
 - On-the-go sensor mapping
 - To compare these maps in terms of the accuracy of soil pH, buffer pH, and lime requirement predictions



Data Collection

Sampling/measuring	Lunz field	Strnad field	Williams field
Grid-based samples (1-ha)	61	50	24
Calibration samples	10	10	10
Validation samples	12	14	15
On-the-go soil pH measurements	1125	1698	610
On-the-go EC _a measurements	13069	19717	7024



BpH Prediction for Alkaline Soils

- $BpH = a_0 + a_1 \cdot pH$
 - pH is soil pH in a 1:1 soil-water solution
 - a_0 and a_1 are regression parameters
- $BpH = 7.0 + a(pH - 6.5)$
 - BpH of 7.0 corresponds to pH of 6.5
- $BpH = 2.7 + 0.67pH$
 - BpH of 7.0 corresponds to pH of 6.5
 - BpH of 8.0 corresponds to pH of 8.0

Lab pH Predictions

$$Lab \ pH = b_0 + b_1 \cdot Sensor \ pH$$

- 1) $b_0 = 0$ and $b_1 = 1$ – original data
- 2) $b_0 \neq 0$ and $b_1 = 1$ – data shift
- 3) $b_0 = 0$ and $b_1 \neq 1$ – data scale
- 4) $b_0 \neq 0$ and $b_1 \neq 1$ – linear regression



Lab BpH Prediction

$$Lab\ BpH = c_0 + c_1 \cdot Sensor\ pH + c_2 \cdot EC_a + c_3 \cdot Sensor\ pH \cdot EC_a$$

- 1) $c_1 = 1, c_0 = c_2 = c_3 = 0$ – raw data
- 2) $c_0 \neq 0, c_1 = 1, c_2 = c_3 = 0$ – data shift
- 3) $c_0 \neq 0, c_1 \neq 1, c_2 = c_3 = 0$ – linear regression without ECa
- 4) $c_1 \neq 1, c_0 \neq c_2 \neq 0, c_3 = 0$ – linear regression with ECa but without the product of sensor pH and ECa
- 5) $c_1 \neq 1, c_0 \neq c_2 \neq c_3 \neq 0$ – full regression



RMSE (Lab pH Prediction)

Field	Data Set	pH Partial Case			
		1	2	3	4
Lunz	calibration	0.81	0.59	0.54	0.32
	validation	0.93	0.82	0.81	0.78
Strnad	calibration	0.49	0.42	0.40	0.36
	validation	0.47	0.52	0.53	0.52
Williams	calibration	0.52	0.50	0.49	0.29
	validation	0.64	0.62	0.61	0.46

- 1 $b_0 = 0$ and $b_1 = 1$ (raw data)
- 2 $b_0 \neq 0$ and $b_1 = 1$ (data shift)
- 3 $b_0 = 0$ and $b_1 \neq 1$ (data scale)
- 4 $b_0 \neq 0$ and $b_1 \neq 1$ (linear regression)



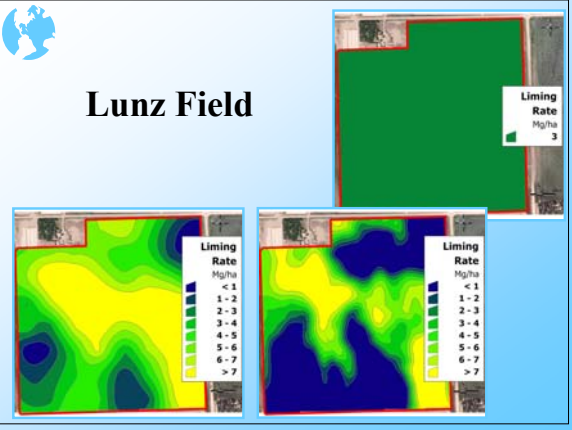
RMSE (Lab BpH Prediction)

Field	Data Set	BpH Partial Case				
		1	2	3	4	5
Lunz	calibration	0.56	0.19	0.16	0.13	0.12
	validation	0.63	0.34	0.32	0.38	0.39
Strnad	calibration	0.84	0.25	0.24	0.27	0.22
	validation	0.79	0.41	0.41	0.42	0.44
Williams	calibration	0.61	0.30	0.16	0.16	0.16
	validation	0.77	0.37	0.33	0.34	0.42

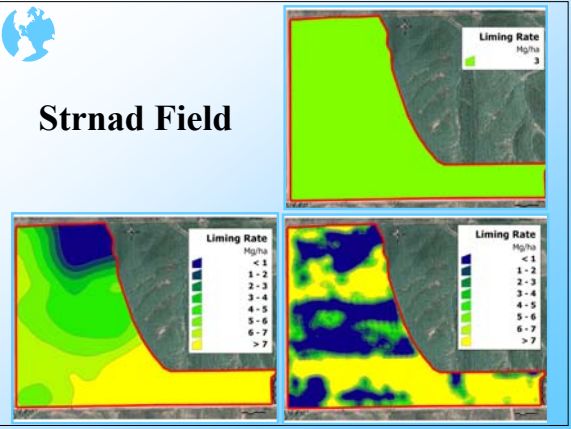
- 1 $c_0 = 0, c_1 = 1, c_2 = 0, \text{ and } c_3 = 0$ (raw data)
- 2 $c_0 \neq 0, c_1 = 1, c_2 = 0, \text{ and } c_3 = 0$ (data shift)
- 3 $c_0 \neq 0, c_1 \neq 1, c_2 = 0, \text{ and } c_3 = 0$ (sensor regression)
- 4 $c_0 \neq 0, c_1 \neq 1, c_2 \neq 0, \text{ and } c_3 = 0$ (sensor + ECa regression)
- 5 $c_0 \neq 0, c_1 \neq 1, c_2 \neq 0, \text{ and } c_3 \neq 0$ (full regression)



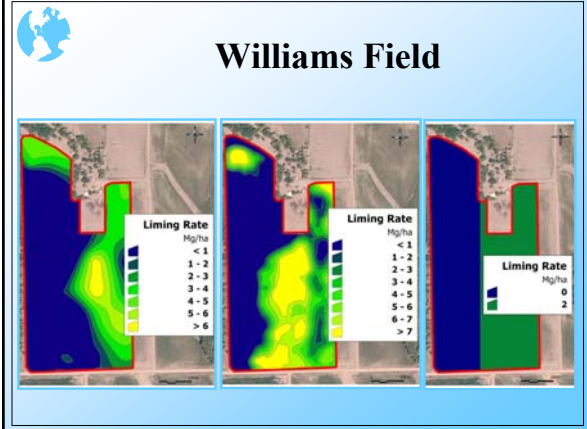
Lunz Field



Strnad Field



Williams Field



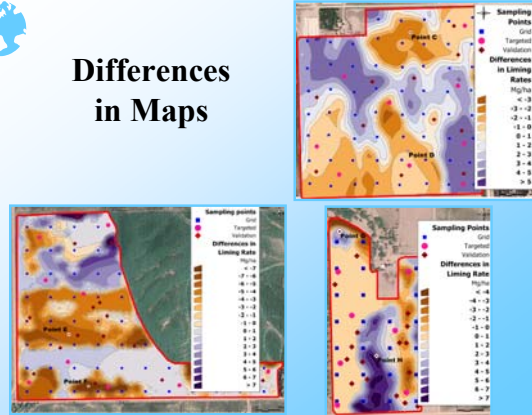


Coefficients of Determination

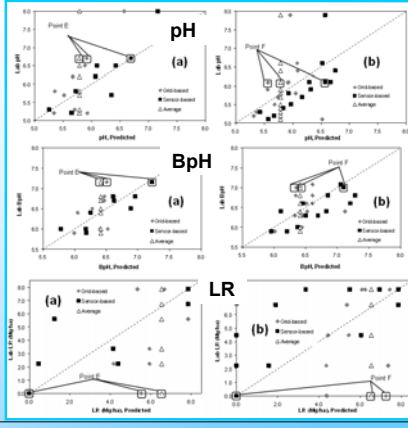
Map	Dataset	Lunz		Strnad		Williams	
		Grid	Sensor	Grid	Sensor	Grid	Sensor
Lab pH	Calibration	0.47	0.92	0.62	0.81	0.91	0.91
	Validation	0.50	0.52	0.05	0.58	0.37	0.79
Lab BpH	Calibration	0.61	0.91	0.65	0.85	0.90	0.96
	Validation	0.67	0.60	0.15	0.50	0.54	0.84
LR	Calibration	0.55	0.97	0.45	0.75	0.73	0.87
	Validation	0.75	0.61	0.03	0.64	0.36	0.46



Differences in Maps



Strnad Field



Mean Absolute Error

Fields	Calibration			Validation			
	Sensor	Grid	Average	Sensor	Grid	Average	
Lunz	pH	0.254 ^a	0.824 ^b	1.280 ^b	0.506 ^a	0.533 ^a	0.970 ^a
	BpH	0.160 ^a	0.390 ^b	0.580 ^c	0.229 ^a	0.231 ^a	0.460 ^b
	LR	0.144 ^a	0.914 ^{ab}	1.800 ^b	0.395 ^a	0.503 ^a	1.700 ^b
Strnad	pH	0.402 ^a	0.700 ^a	0.810 ^a	0.560 ^a	0.487 ^a	0.580 ^a
	BpH	0.309 ^a	0.524 ^a	0.642 ^a	0.398 ^a	0.359 ^a	0.378 ^a
	LR	0.420 ^a	1.007 ^{ab}	1.500 ^b	1.138 ^a	1.332 ^a	1.150 ^a
Williams	pH	0.194 ^a	0.244 ^b	0.650 ^b	0.422 ^a	0.398 ^a	0.391 ^a
	BpH	2.552 ^a	2.713 ^a	2.709 ^a	0.840 ^a	0.889 ^a	0.834 ^a
	LR	1.824 ^a	2.277 ^a	2.550 ^a	1.177 ^a	1.522 ^a	1.450 ^a



Conclusions

- Lime application maps based on sensor data with ten calibration points provided better delineation of acidic soil areas that needed lime than grid sampling or field average methods
- When defining a site-specific relationship between corresponding sensor pH and lab pH/BpH measurements, it is not always necessary to adjust each parameter of a corresponding regression model



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