### Chapter







Methods





## Study I: Published



# Clustering Tools for Integration of Satellite Remote Sensing Imagery and Proximal Soil Sensing Data

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# **Data Clustering**

#### Chapter



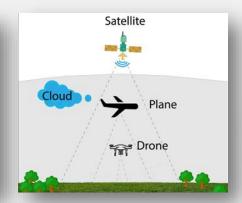


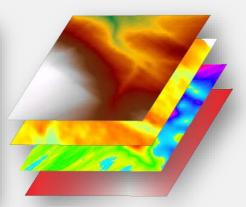






- A large amount of geospatial data provide essential information for developing thematic maps.
- Hierarchical data clustering techniques are crucial for data mining, and high-density data analysis is important for field management.
- This study presents the process used to develop a new and enhanced clustering technique to better understand field variability.





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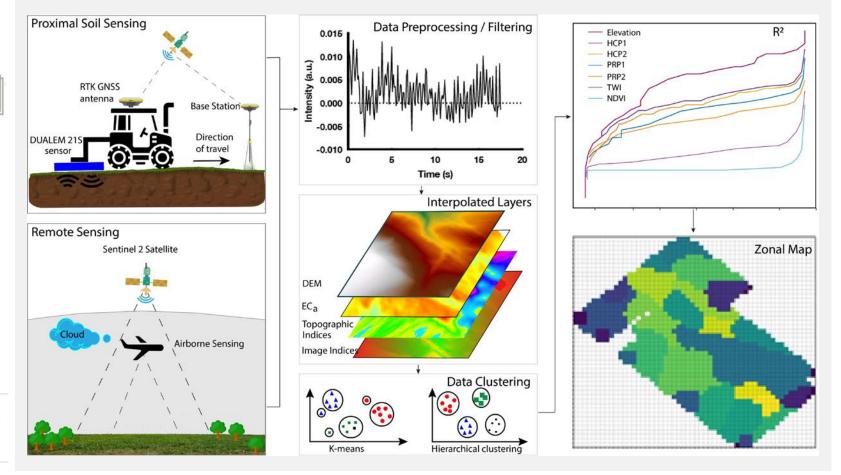




Results



## Sensor Data > Data processing > Clustering > Comparison



# **Neighborhood Search Analyst Tool**

### Chapter



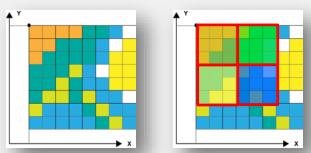








- NSA developed on open-source programming platform to enable hierarchical spatial clustering of multilayer data.
- The algorithm evaluates data structure on location-based information of the multiple layers to be used for delineating spatially constrained groups of high-density sensor-based measurements.
- The algorithm seeks to minimize the mean squared error (MSE) during the interactive grouping of spatially adjacent measurements similar to each other and different from other parts of the field.



# **Data and Study Sites**

## Chapter



# ackground

# Methods





## **PSS and RS data**

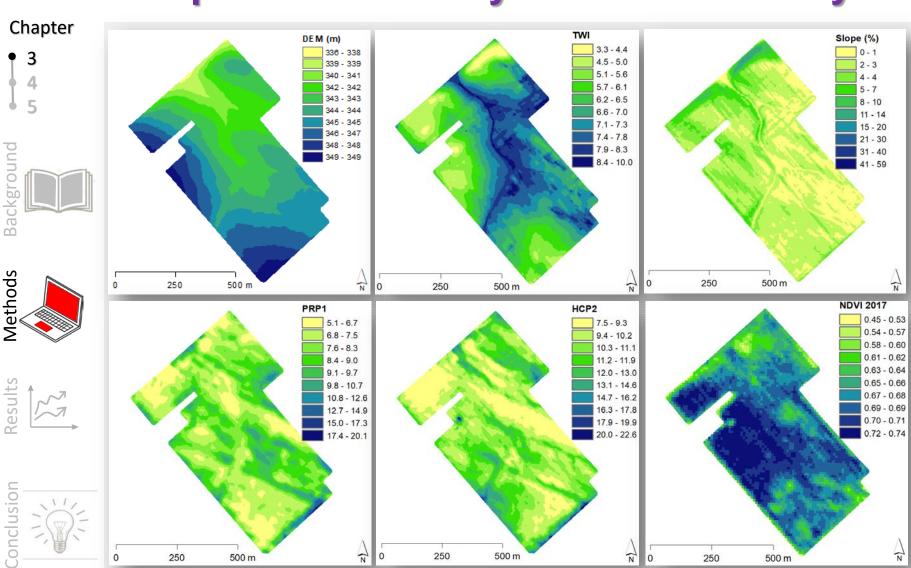
- Apparent Electrical Conductivity (EC<sub>a</sub>)
- Elevation, slope, aspect ratio & topographic wetness index (TWI)
- Normalized difference vegetation index (NDVI),
  Normalized difference red-edge index(NDRE)

Field	Area (ha)	Soil Class	Target crops
WH	39.60	Loam	Soybean/Wheat
LD	21.00	Sandy loam - SL	Soybean
RB	75.00	Fine SL	Soybean/Wheat



STUDY III

# **Interpolated Data Layers and Variability**



STUDY II

BACKGROUND OBJECTIVES

# k-means vs. NSA Clustering

#### Chapter









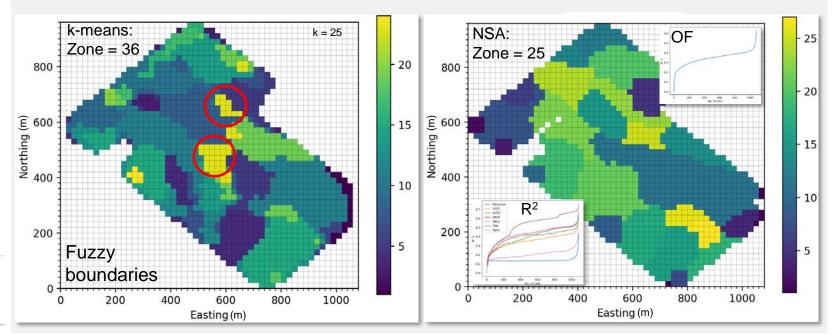


## K-means

- R<sup>2</sup> and zones varied since there was a random component.
- User-defined clusters with isolated boundaries.

## NSA

- Max. 70% field variance (R<sup>2</sup> = 0.70) achieved in NSA.
- Optimum number of zones without defining cluster centers.



# **Comparison: Two Clustering Methods**

Chapter



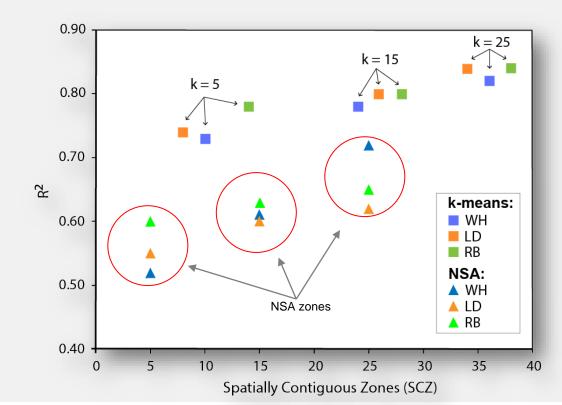








- **User defined clusters**: k = 5 / 15 / 25
- k-means: WH, LD, RB: 36, 34, 38 zones (k = 25)
- NSA: Stable and spatially contiguous zones



k-means:  $R^2 = 0.80$ 

NSA:  $R^2 = 0.60$ 

### Chapter





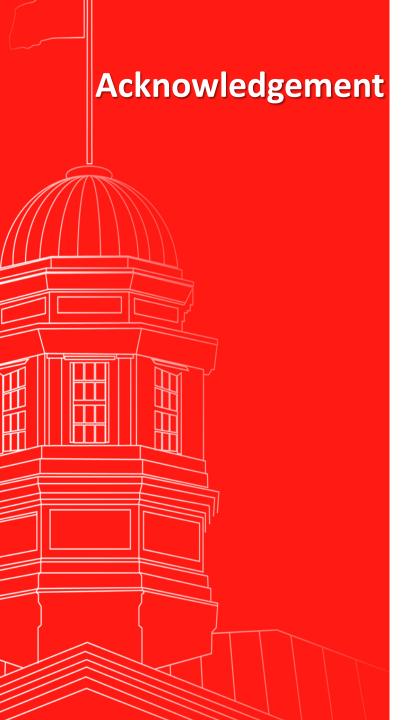






- The high-density and multivariate data clustering approach provided user-defined number of zones for the agricultural fields.
- Compared to fuzzy clustering algorithms, NSA has a unique capability for optimum zone separation.
- It proved to be capable of handling a significant number of variables and data layers for delineating the optimum number of zones in a more robust way.
- Sampling optimization and georeferenced thematic maps are useful for variable rate technologies and for other management purposes.

STUDY II





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