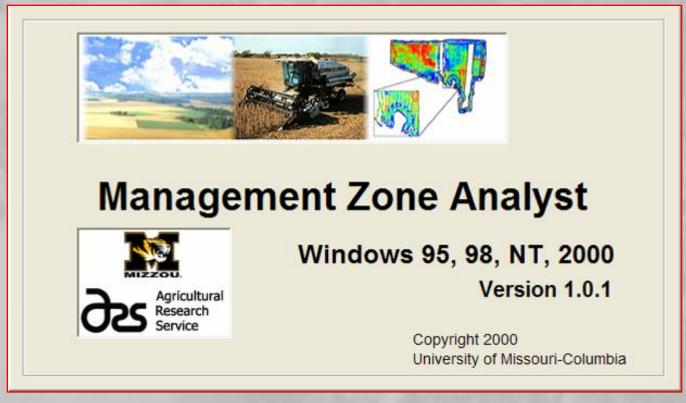
# Management zone analyst (MZA): software for subfield management zone delineation

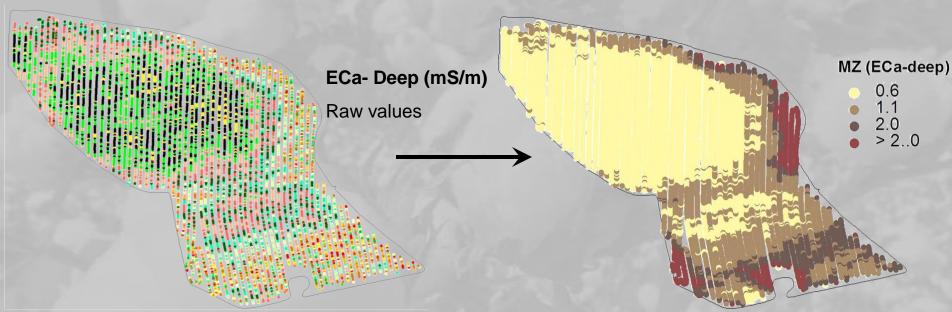


Fridgen, J.J., N.R. Kitchen, K.A. Sudduth, S.T. Drummond, W.J. Wiebold, and C.W. Fraisse. 2004. Management zone analyst (MZA): software for subfield management zone delineation. *Agronomy Journal*. 96: 100-108.

http://www.ars.usda.gov/services/software/download.htm?softwareid=24&modecode=36-20-15-00



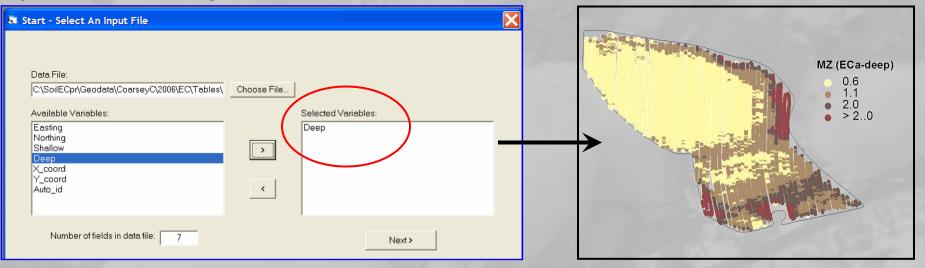
# How does MZA classify a data set into zones?

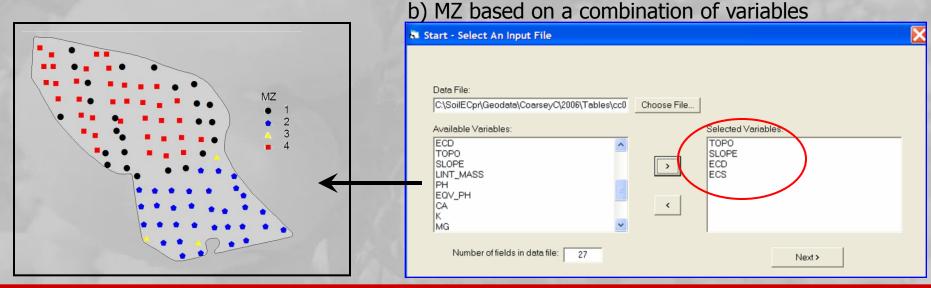


- MZA using an unsupervised fuzzy classification:
  - Find the "most alike" areas in the field.
  - © Compare all the observations to each other and cluster the similar ones together.
  - @ Generate clusters or "zones".

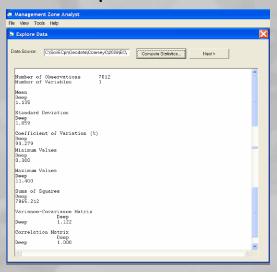
#### MZA: software for subfield management zone delineation

a) MZ based on a single variable

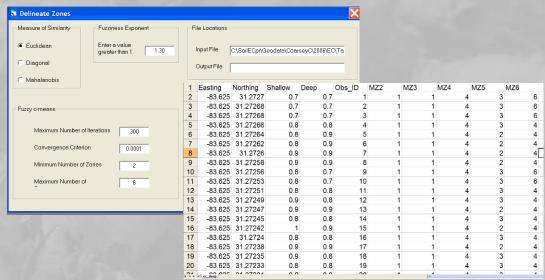




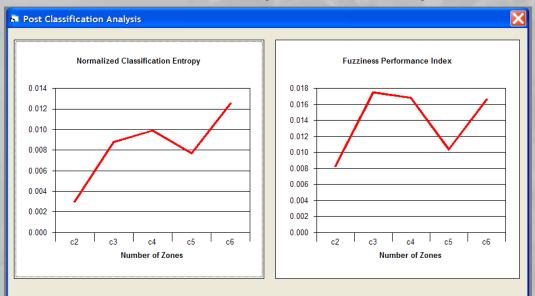
#### 1. Descriptive statistics



2. Delineation of the zones



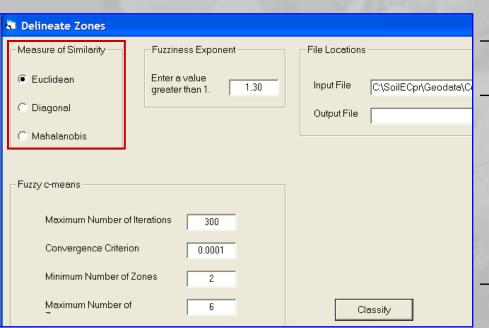
3. Evaluation of classification performance by the number of zones



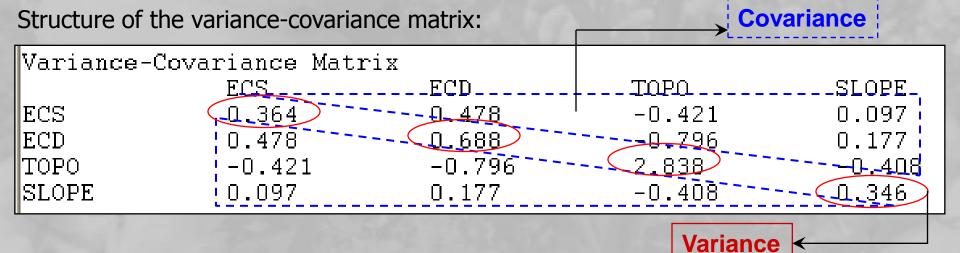
1. Descriptive statistics (Univariate and Multivariate statistics)

Explore Da	ita				X
Data Source:	C\SoilECpr\Geodata\(	CoarseyC\2006\Tabl	Compute Stat	istics Next >	
	Observations Variables	99 4			
Mean ECS 0.818	ECD 1.104	TOPO 78.224	SLOPE 1.225		≡
Standard ECS 0.604	Deviation ECD 0.829	TOPO 1.685	SLOPE 0.588		
Coefficie ECS 73.833	nt of Variation ECD 75.125	(%) TOPO 2.154	SLOPE 48.035		
Minimum V ECS 0.000	alues ECD 0.000	TOPO 74.282	SLOPE 0.222		^
Maximum V ECS 5.123	alues ECD 6.026	TOPO 80.658	SLOPE 3.520		
Sums of S ECS 35.709	quares ECD 67.387	TOPO 278.127	SLOPE 33.932		
Variance- ECS ECD TOPO SLOPE	Covariance Matri ECS 0.364 0.478 -0.421 0.097	ECD 0.478 0.688 -0.796 0.177	TOPO -0.421 -0.796 2.838 -0.408	SLOPE 0.097 0.177 -0.408 0.346	
Correlati ECS ECD TOPO SLOPE	on Matrix ECS 1.000 0.955 -0.414 0.272	ECD 0.955 1.000 -0.570 0.362	TOPO -0.414 -0.570 1.000 -0.411	SLOPE 0.272 0.362 -0.411 1.000	<b>=</b>

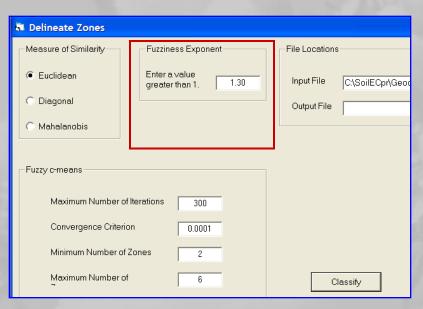
#### 2. Delineation of the zones



Scenario	Measure of Similarity
One classification variable (i.e. Yield or ECa-deep)	Euclidean
Equal variances; covariances ≈ 0	Euclidean
Unequal variances; covariances ≈ 0	Diagonal
Unequal variances; covariances ≠ 0 (i.e. ECS, ECD, Topo, SLOPE)	Mahalonobis



#### 2. Delineation of the zones

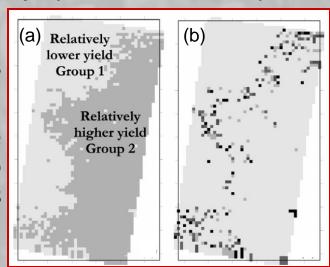


- MZA uses a Fuzzy unsupervised classification method.
- Observations can be members of more than one zone. This scenario usually occurs at the transitory areas (edges) between zones
- Fuzziness exponent ≈ 1 more distinct zones
  (i.e. no membership sharing)

## **Example:**

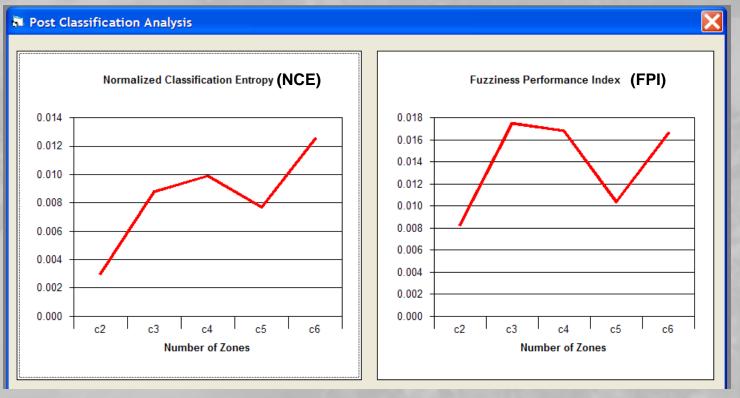
MZ based on yield data (Boydell and McBratney, 2006)

- (a) High and low yielding zones
- (b) The dark colors represent areas where membership is spread equally between the two zones



3. Evaluation of classification performance by the number of zones

HOW MANY UNIQUE ZONES A FILED SHOULD BE DIVIDED INTO?



NCE – Measures the homogeneity of the zones.

FPI – Measures the degree of separation between the zones.

"The optimum number of zones is when both indices are at the minimum values".

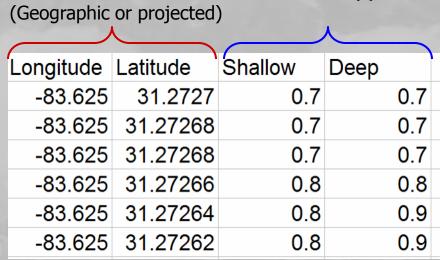
FPI ≈ 0, distinct classes → less membership sharing

# **Example: MZ based on EC<sub>a</sub>-Deep data**

Coordinates

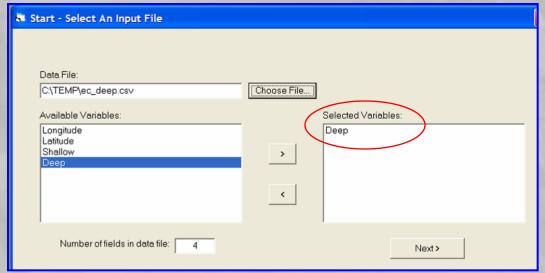
1. Data as comma-delimited ASCII text files (csv or txt)





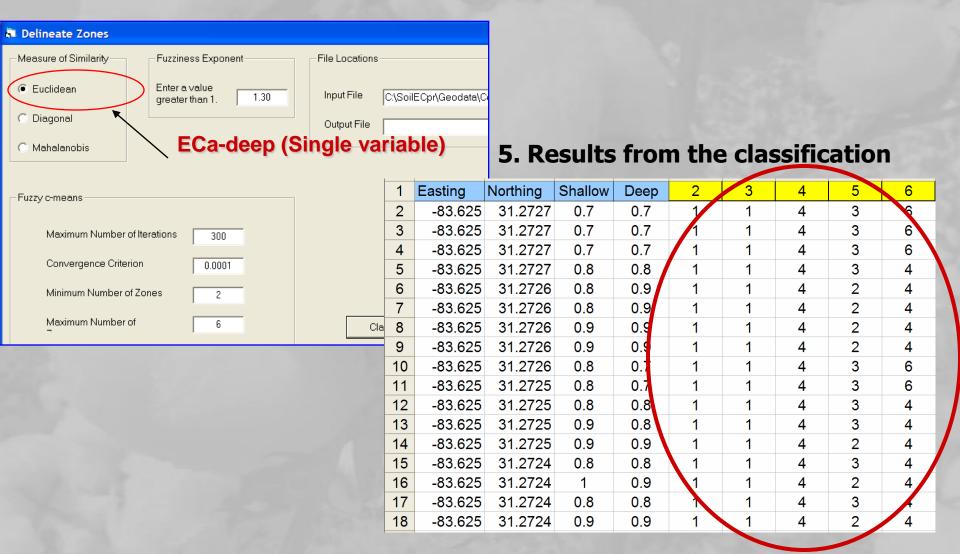
Variable(s)

2. Selecting the variable(s) for the zones delineation



# **Example: MZ based on ECa-deep data**

- 3. Calculating descriptive statistics
- 4. Classifying the ECa-deep data



#### 5. Selecting the best number of zones

