



Remote land survey

Image classification

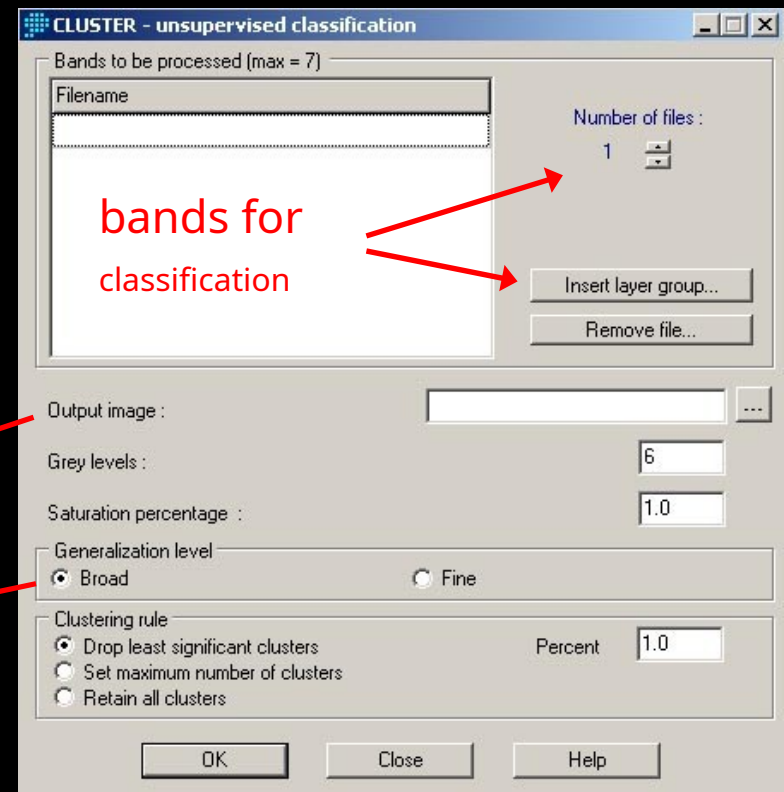


Unguided classification in IDRISI

Module CLUSTER

(Image Processing / Hard Classifiers)

- uses the technique of histogram peaks



resulting image

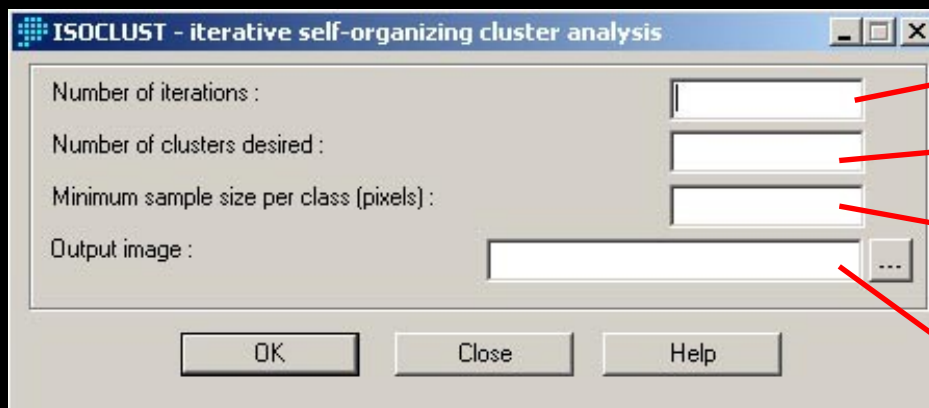
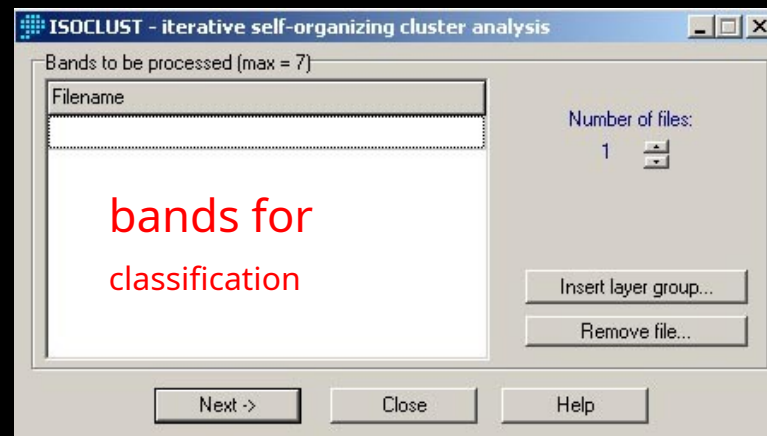
choice of algorithm:
broad (prominent peaks
only) fine (all peaks)



Module **ISOCLUST**

(Image Processing / Hard Classifiers)

- ISODATA algorithm concept
- uses minimal to classify pixels into classes spectral distance



- number of calculation iterations number
- of clusters (classes) to create
- minimum number of training pixels
- resulting image

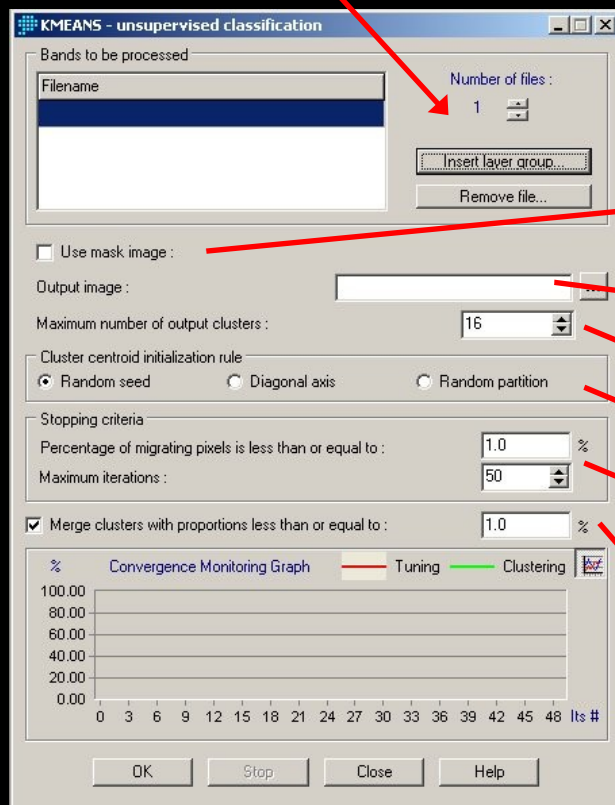


Module KMEANS

(Image Processing / Hard Classifiers)

bands for
classification

- calculation of the average vectors of the clusters and their comparison with the classified pixels



mask

resulting image

the number of clusters (clusters, classes) to create

the method of distribution of the initial clusters

process termination criteria

- max % of migrating pixels

- max. number of iterations

choosing to merge clusters below the threshold



Managed classification in IDRISI

- Creation of training sets
digitizing from the screen



name created
vector layers

display palette

data type (integer/real)

element identification value

Digitize

Name of layer to be created :

Symbol file for display :

Data Type
 Integer Real

ID or Value :

Layer Type
 Point
 Line
 Polygon
 Text

Digitize as :
Flood Polygon

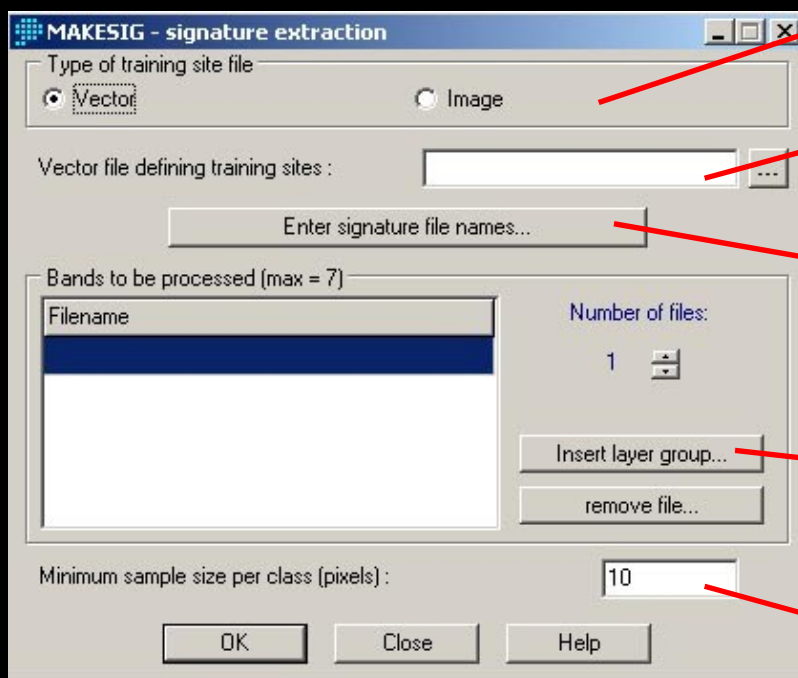
OK Close Help

vector type

with multiple polygons for one
class, all these polygons need to be assigned the same ID



- creation of signatures– module **MAKESIG**
(Image Processing / Signature Development)



the type of file defining the training set and its name

naming of individual signatures (classes)

insertion of bands from which signatures are to be calculated

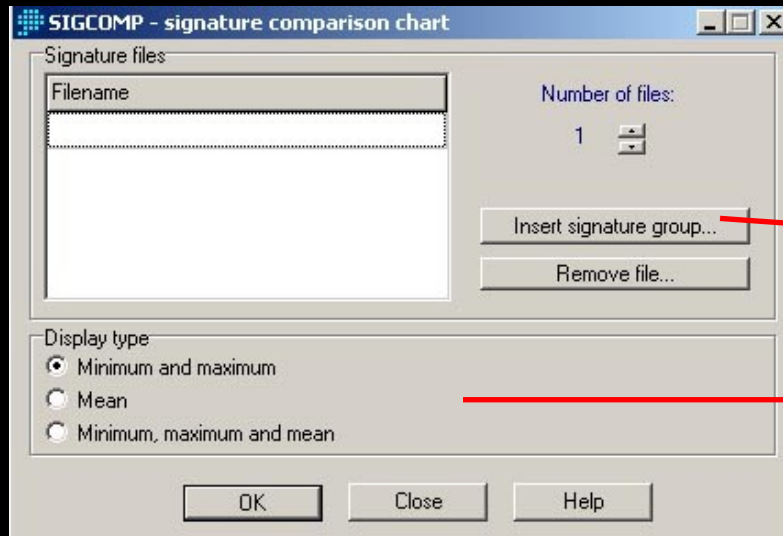
the minimum number of training pixels for each class



Comparison of signatures

Idris module **SIGCOMP**

(Image Processing / Signature Development)



inserting a signature file

choice of graph type

Histogram of signatures - module **HISTO**



Idris module MAXLIKE

options for a priori setting of probabilities for each class

MAXLIKE - maximum likelihood classification

Use equal prior probabilities for each signature
 Specify a prior probability image for each signature
 Specify a prior probability value for each signature
 Specify either a value or an image for each signature

Signatures to use in classification:

Signature	Probability value/image	Probability definition
	Value	0.5
	Value	0.5

Number of files: 2

Insert signature group...
remove file...

Proportion to exclude:
 0% (classify all pixels)
 1%
 5%
 Chi squared value:

Output image: [] Title: []

OK Close Help

insertion groups signatures

determination of % pixels, which can remain not classified

the name of the resulting file



Other classification modules in IDRISI

- **PIPED**– rectangle classifier
- **MINDIST**– minimum distance classifier
- **Soft classifiers**
 - they do not create a definitive classification of pixels into classes
 - the result is several images (according to the number of signatures) showing the probabilities of each pixel belonging to each class
 - Idrisi modules: **BAYCLASS**, **BELCLASS**



Literature

- Eastman, JR 2009: Idrisi TAIGA Guide to GIS and Image Processing, Clark Labs, Worcester, MA



Remote land survey

Classification accuracy



Rating classification

- **Error matrix**– compares the relationship between reference data and classification results for all classes
- **Kappa coefficient**– compares the resulting classification with the classification created by a purely random process of assigning pixels to classes

a value of 0.9 means that we have avoided 90% of the errors that would have been caused by random assignment



Error matrix

- **errors of omission**(omission error) - "forest" pixels are classified elsewhere (columns)
- **errors from incorrect classification**(commission error) - pixels representing a different surface (lines) are included in the "forest" class
- **accuracy from the user's point of view**(user's accuracy) - the probability with which a pixel assigned to a certain class actually represents it
- **accuracy from the point of view of the processor**(producer's accuracy)
= number of correct ears. pixels / number of pixels used for testing the given class



Calculation of the error matrix

- average accuracy

classification of all classes:

$$110+90+55+30 / 301 = 95\%$$

	Reference data					
Classification	class	<u>forest</u>	field	water	houses	Σ
	<u>forest</u>	110	0	5	0	115
	field	10	90	0	0	100
	water	0	0	55	0	55
	houses	1	0	0	30	31
	Σ	121	90	60	30	301



Error of omission

- $= 11/121 = 9\%$
- sum of pixels in the column off the main diagonal / number of reference data pixels

		Reference data				
Classification	class	<u>forest</u>	field	water	houses	Σ
	<u>forest</u>	110	0	5	0	115
	field	10	90	0	0	100
	water	0	0	55	0	55
	houses	1	0	0	30	31
	Σ	121	90	60	30	301



Misclassification error

- $= \frac{5}{121} = 4\%$
- sum of pixels in the row outside the main diagonal / number of reference data pixels

		Reference data				
Classification	class	<u>forest</u>	field	water	houses	Σ
	<u>forest</u>	110	0	5	0	115
	field	10	90	0	0	100
	water	0	0	55	0	55
	houses	1	0	0	30	31
	Σ	121	90	60	30	301



Accuracy from the user's point of view

- $=110/115 = 96\%$
- number of correctly classified pixels / total number of pixels assigned to the class

		Reference data				
Classification	class	<u>forest</u>	field	water	houses	Σ
	<u>forest</u>	110	0	5	0	115
	field	10	90	0	0	100
	water	0	0	55	0	55
	houses	1	0	0	30	31
	Σ	121	90	60	30	301



Accuracy from the point of view of the processor

- $=110/121 = 91\%$
- number of correct spikes. pixels / number of pixels used for testing the given class

		Reference data				
Classification	class	<u>forest</u>	field	water	houses	Σ
	<u>forest</u>	110	0	5	0	115
	field	10	90	0	0	100
	water	0	0	55	0	55
	houses	1	0	0	30	31
	Σ	121	90	60	30	301



Literature

- Campbell, JB 1996: Introduction to Remote Sensing, Taylor and Francis, London
- Dobrovolný, P. 1998: Remote sensing of the Earth, digital image processing, Masaryk University, Brno