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FINAL REPORT

Training Course on "Image Processing and Remote Sensing for Industrial Development in Coastal Areas"

Iglesias, Cagliari, Italy 4-8 June 2002

organized by ICS-UNIDO

in collaboration with FORGEA INTERNATIONAL (Training and Cooperation Centre in the Field of Geo-Mining and Environment)

> and the Italian Ministry of Foreign Affairs



BACKGROUND

Coastal zone management refers to the management of coastal resources. Resources are common to other areas (urban, industrial, agricultural, forestry, military, tourism and recreation, mining, dumping and waste disposal, etc.) but also they are specific to coastal areas (fishing, navigation, port and harbor, desalination and salt production, etc.).

The complex structure of coastal zone is distinguished by a particularly "attractive" area, with a high population density and economically dynamic, which includes large cities and big ports. In this particular area, made up of inland and marine ecosystems, also large-scale industries (agro-industry, iron and steel, metallurgy, chemical, mechanical, shipyards and tourism), are settled frequently compensating natural landscapes and resources.

To achieve balanced environmental planning and sustainable exploitation of natural resources in coastal areas, industrial development has to be supported by a knowledge-based management system. This system should integrate advanced technologies for monitoring the development in all its aspects.

In this context, ICS-UNIDO has organized in collaboration with FORGEA International, a Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas" from 4 to 8 June 2002. This Training Course was developed within the activities of the ICS Area of Earth, Environmental and Marine Sciences and Technologies, under the Coastal Zone Management (CZM) Sub-programme.

JUSTIFICATION

Integration of Remote Sensing with other related technologies/disciplines such as GIS (geographic information systems), DSS (decision-support systems) and computer science is essential to acquire complete benefits of this technique. However, in most cases the separation between those technologies is apparent because of the cost and the availability. Remote Sensing experts are not well exposed to the technology of GIS and GIS specialists do not receive enough knowledge and sufficient training on image acquisition and processing. Therefore, a complete knowledge fusion between GIS and Remote Sensing needed to be conducted and experts should be exposed to all technologies.

For example, spatial data processing and multi-criteria decision-support systems are necessary tools for industrial siting, urban planning and for establishing a system of monitoring stations, suitable for registering atmospheric and aquatic pollution. The integration of coastal environment engineering, GIS technology, GPS applications, numerical modelling with remotely sensed data management can be done through image processing and GIS technologies. This is useful for presenting a common "interface" and "approach" to tackle environmental problems and to control industrial impact.

In this context, the training course has provided participants with practical expertise used to solve problems in real case studies based on some lessons on remote

sensing and image processing relied on a mixture of theory and practical applications followed by exercises to complement theoretical lecture topics in order to create a good training environment.

The course has given to the participants the opportunity to access advanced software packages using remotely sensed data for elaborating some exercises developed on real case studies for supporting decision-making process.

OBJECTIVES

In this Training Course, ICS intended to train junior experts on image processing and remote sensing techniques, as supporting tools for the management of coastal areas, using advanced software packages.

The specific objective was the development of skills in the use of remotely sensed data for the purpose of knowledge of image acquisition methods, data merging, image rectification, image enhancement, image classification and image algebra.

STRUCTURE OF THE COURSE

The course started with some introductory lectures presenting an overview of the state of the environment linked to socio-economical development, in particular industrial development in coastal areas, followed by some lessons on the basic concepts of remote sensing, GIS, DSS, process simulation, multivariate analysis and multi-criteria evaluation. During the training course a practical session in which participants profited from a direct access to specific software was of great interest.

RESULTS

- Trained experts;
- Strengthening ICS collaboration with international institutions and organizations.

CONCLUSIONS AND RECOMMENDATIONS

Coastal zone is of extreme significance to human affairs. In this specific area, the real conflict is not between sea and shore but between man and nature. The management of coastal resources should be based on:

- An integrated approach, which requires an integration of advanced technologies including Remote Sensing, GIS and DSS for understanding, analyzing and managing physical, chemical and biological structure of coast in relation to anthropogenic exploitation. This approach involves various forms of analysis including simulation and modeling;
- A concrete collaboration between international institutions interested in developing strategies for sustainable industrial development by transferring technologies that are useful to monitor landscape changes in industrial areas

and their surrounding environment and to support decision-makers for coastal zone management plans.

IMMEDIATE FOLLOW-UP

- A final report of the training course, which can be subject to scientific editing and publication and should include all presentations, exercises and lecture notes;
- Build-up a mailing list for the participants to exchange information and knowledge;
- Preparation of a new presentation for GIS-GRASS users' conference according to some suggestions from participating experts.

LIST OF ANNEXES

PROGRAMME See attached Annex I.

LIST OF PARTICIPANTS See attached Annex II.

PRESENTATIONS See attached Annex III.

STATISTICS Se attached Annex IV. Annex I PROGRAMME

Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas"

Iglesias, Cagliari, Italy 4-8 June 2002

Programme

Tuesday, 4 June 2002

- 09:00-09:30 Registration
- 09:30-10:30 Welcome speech by host Introduction to ICS, Mr. G. Longo Overview on the objectives of the Training Course and presentation of the participants, Mr. E. Feoli
- 10:30-11:00 Coffee break
- 11:00-12:00 Overview of the importance of industry, trade and tourism in coastal areas: mining activity in Sardinia, in Iglesias and its territory, Mr. L. Ottelli
- 12:00-13:00 Definition of land use/land cover for planning industrial development, Mr. E. Feoli
- 13:00-14.30 Lunch break
- 14:30-15:30 Process Simulation for a sustainable industrial development, Mr. G. Longo
- 15:30-15.45 Coffee break
- 15:45-17:00 Technological tools for coastal zone management, Mr. G. Tore
- 17:00-17:15 Open discussion

Wednesday, 5 June 2002

- 09:15-11:00 GIS and Spatial Data integration, Mr. E. Feoli
- 11:00-11:15 *Coffee break*
- 11:15-13:00 Remote Sensing techniques, Mr. A. Altobelli
- 13:00-14:30 Lunch break
- 14:30-15:30 Practical exercises on "satellite images acquisition, data analysis, information extraction, correction techniques and image processing", Mr. A. Altobelli and Mr. M. Ghribi
- 15:30-15:45 Coffee break
- 15:45-16:45 GIS applications using MultiSpec for unsupervised and supervised classification of LANDSAT images, Mr. A. Altobelli
- 16:45-17:00 Open discussion

Thursday, 6 June 2002

- 09:00-11:00 Introduction to multivariate data analysis, Mr. E. Feoli
- 11:00-11:15 *Coffee break*
- 11:15-13:00 GIS, remote sensing and spatial decision support systems (SDSS) for costal zone management, Mr. M. Ghribi
- 13:00-14:30 Lunch break
- 14:30-15:30 Practical exercises using MultiSpec and IDRISI, Mr. A. Altobelli
- 15:30-15:45 *Coffee break*
- 15:45-16:45 Practical exercises on Image Processing and Multi-Criteria Evaluation (MCE) using IDRISI software package, Mr. M. Ghribi (cont.)
- 16:45-17:00 Open discussion

Friday, 7 June 2002

- 09:00-11:00 Presentation of ICS case studies, Mr. M. Ghribi
- 11:00-11:15 Coffee break
- 11:15-13:00 Application of classification and ordination methods in landscape assessment, Mr. E. Feoli
- 13:00-14:30 Lunch break
- 14:30-15:30 Practical exercises on Image Processing and Multi-Criteria Evaluation (MCE) using IDRISI software package, Mr. M. Ghribi (cont.)
- 15:30-15:45 Coffee break
- 15:45-16:45 Practical exercises on Image Processing and Multi-Criteria Evaluation (MCE) using IDRISI software package, Mr. M. Ghribi (cont.)
- 16:45-17:00 Open discussion

Saturday, 8 June 2002

- 09:00-10:45 Focus on the Impacts of industrial development on marine environment and landscape, Mr. N. Candeloro
- 10:45-11:00 *Coffee break*
- 11:00-13:00 Conclusions and Closure of the Training Course

Annex II LIST OF PARTICIPANTS

Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas" Iglesias-Cagliari, Italy 4-8 June 2002

List of Participants

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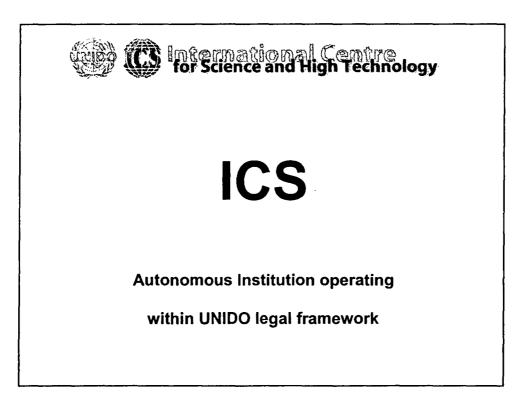
Mr. Gennaro Longo Area Director Earth, Environmental and Marine Sciences and Technologies ICS-UNIDO AREA Science Park Padriciano 99, Bldg. L2 34012 Trieste, Italy Tel.: +39-040-9228104 Fax: +39-040-9228136 E-mail: gennaro.longo@ics.trieste.it Annex III PRESENTATIONS

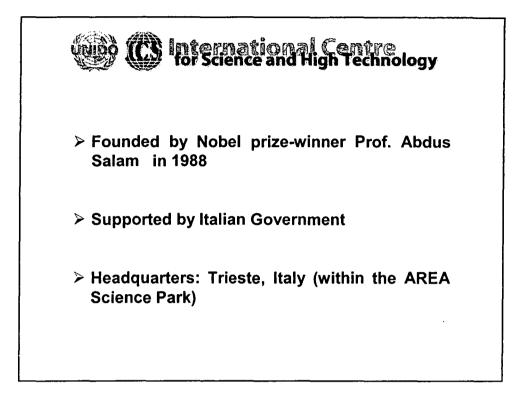
List of Presentations

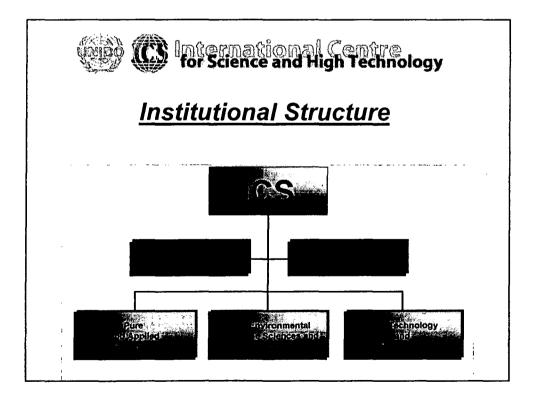
- G. Longo Introduction to ICS
- L. Ottelli Overview of the Importance of Industry, Trade and Tourism in Coastal Areas: Mining Activity in Sardinia, in Iglesias and its Territory
- E. Feoli Definition of Land Use/Land Cover for Planning Industrial Development
- G. Longo Process Simulation for Sustainable Industrial Development
- G. Tore Technological Tools for Coastal Zone Management
- E. Feoli GIS and Spatial Data Integration
- A. Altobelli Remote Sensing Techniques
- A. Altobelli and M. Ghribi Practical Exercises on "Satellite Images Acquisition, Data Analysis, Information Extraction, Correction Techniques and Image Processing"
- A. Altobelli GIS applications using MultiSpec for unsupervised and supervised classification of LANDSAT images
- E. Feoli Introduction to Multivariate Data Analysis
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- M. Ghribi Practical Exercises on Image Processing and Multi-Criteria Evaluation (MCE) using IDRISI Software Package
- M. Ghribi ICS Case Studies Presentation
- E. Feoli Application of Classification and Ordination Methods in Landscape Assessment
- N. Candeloro Focus on the Impacts of Industrial Development on Marine Environment and Landscape

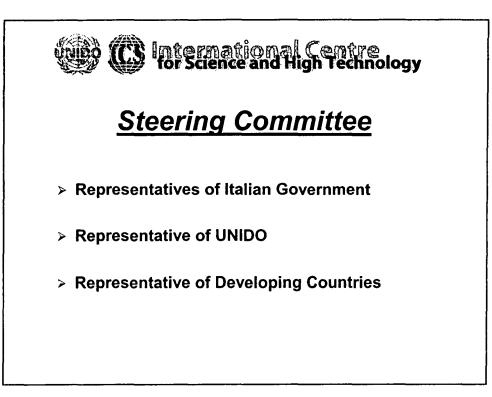
Introduction to ICS by G. Longo





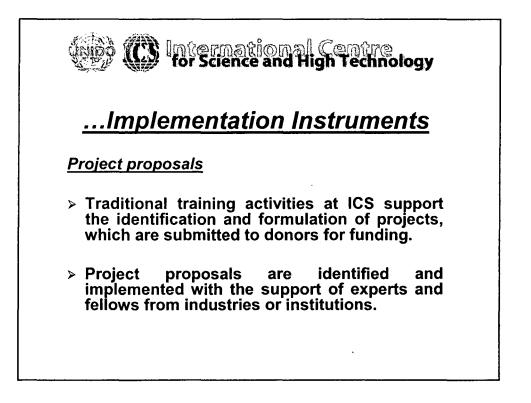














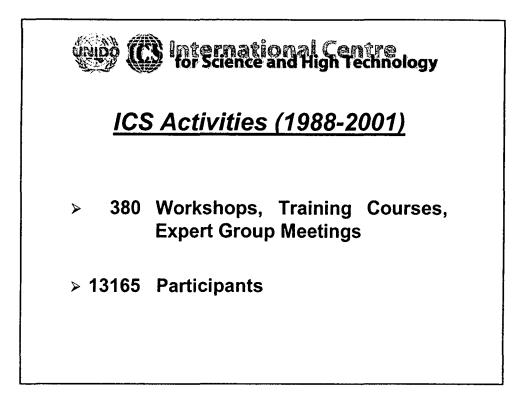
for Science and High Technology

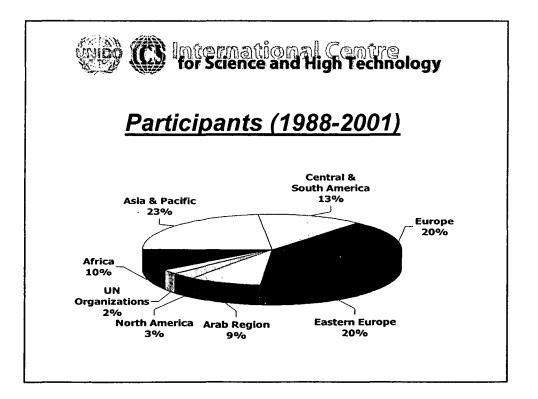
<u>Networking</u>

- > Support to the establishment of networks to improve the exchange of experience.
- Identification in various regions of the world, selection and evaluation of partner institutions willing to offer:

Cooperation and support









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Technical Areas



Pure and Applied Chemistry Area Director: Mr. Stanislav Miertus



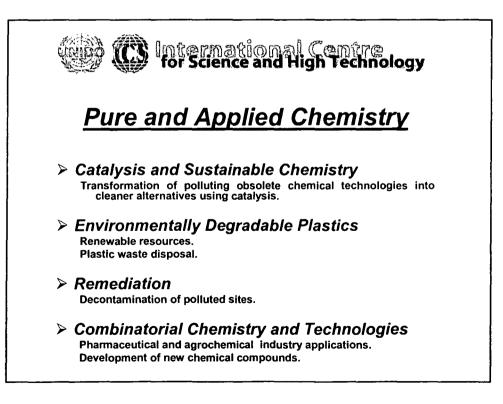
Earth, Environmental and Marine Sciences and Technologies

Area Director: Mr. Gennaro Longo

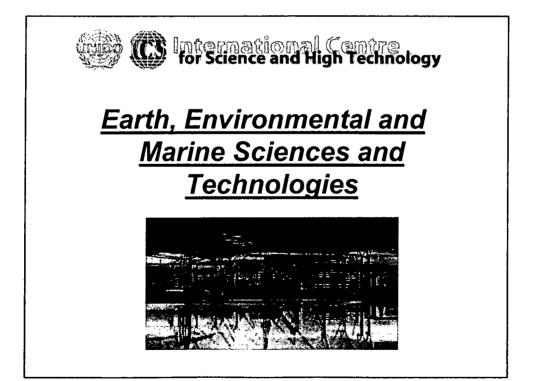


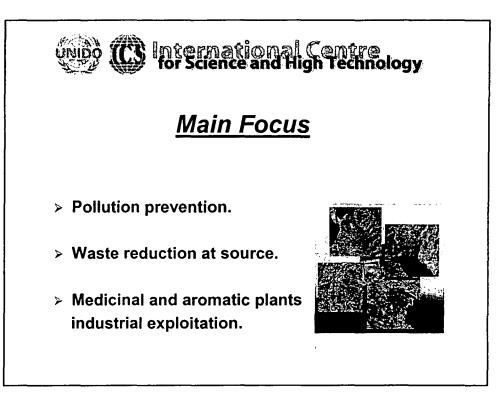
High Technology and New Materials

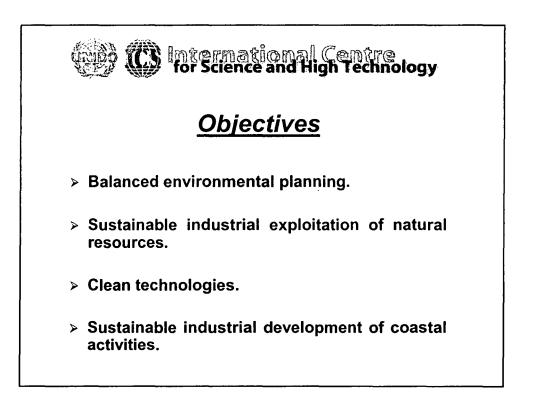
Area Director: Mr. Emilio Vento

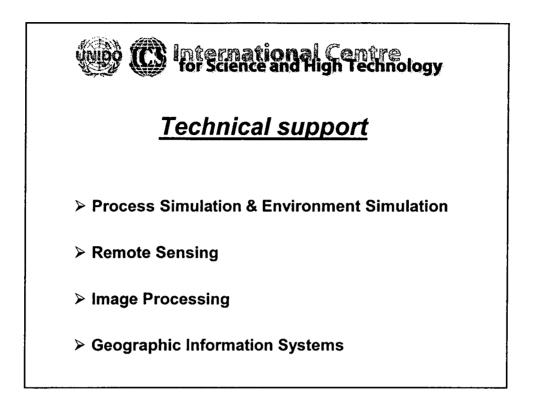


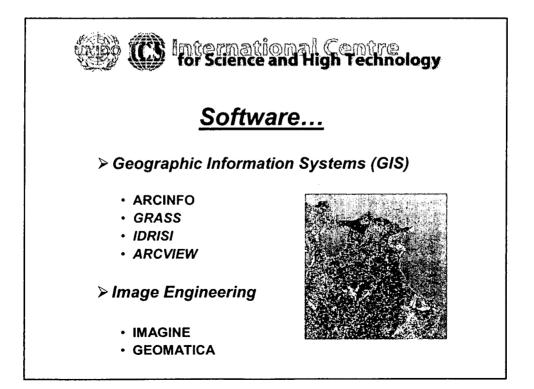


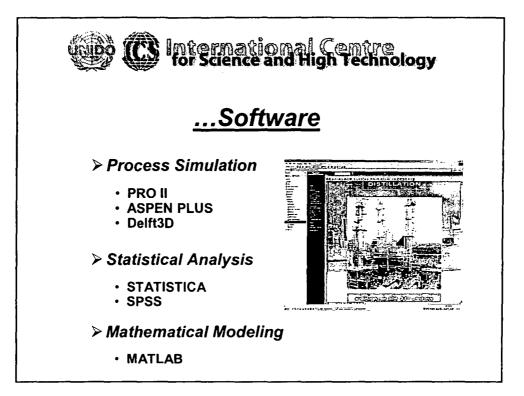


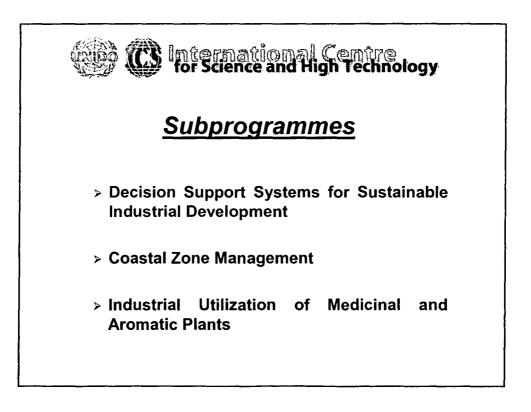




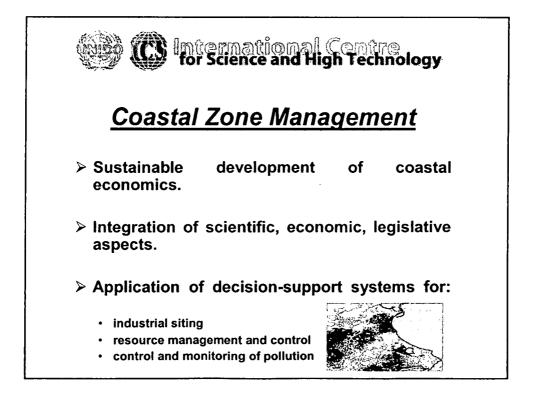




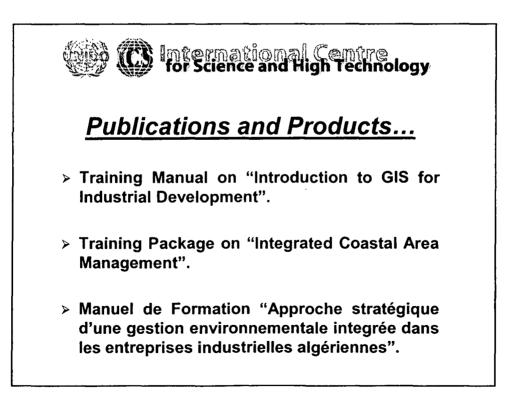


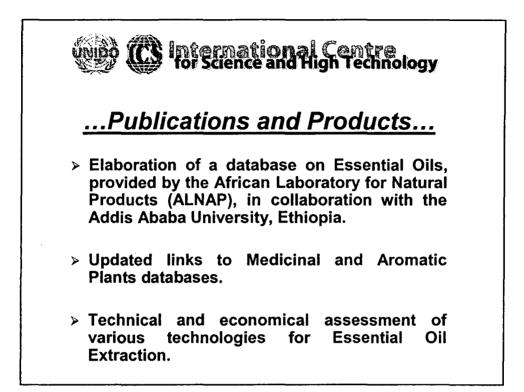


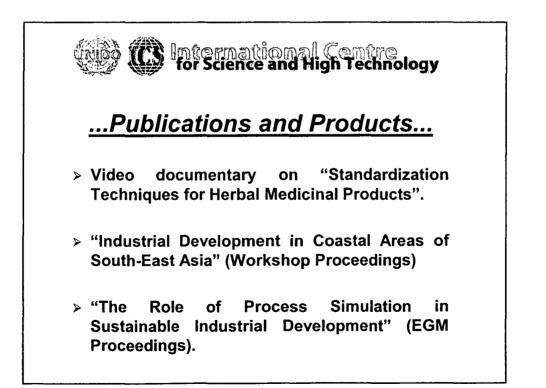


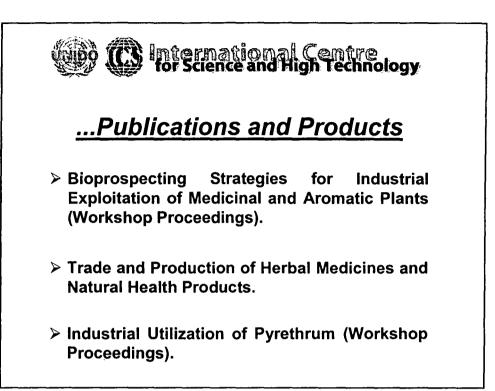


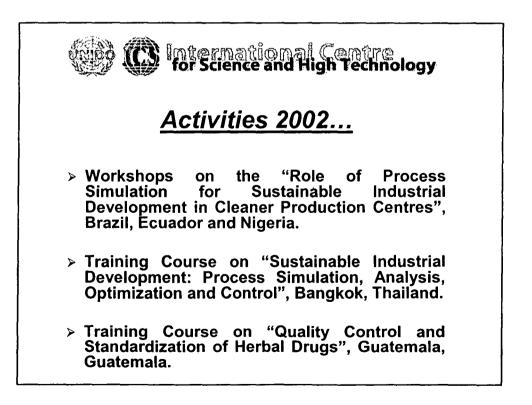


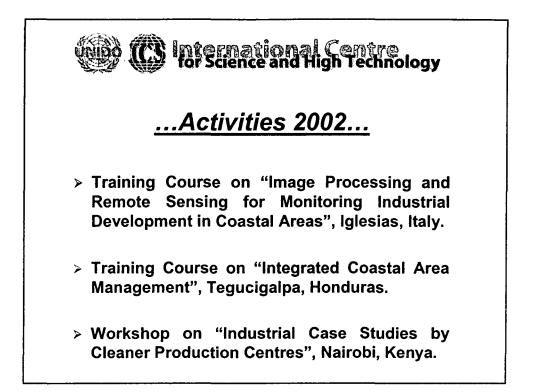


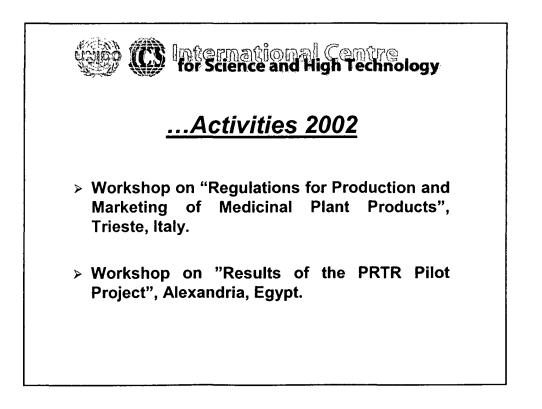


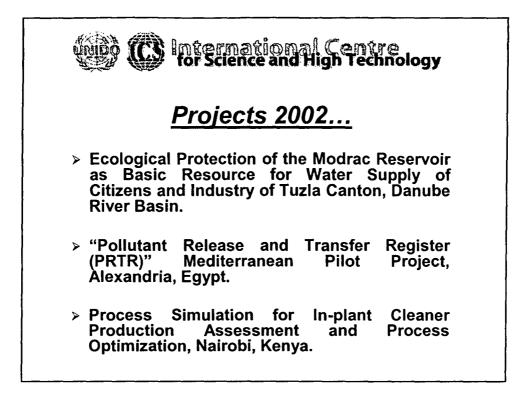


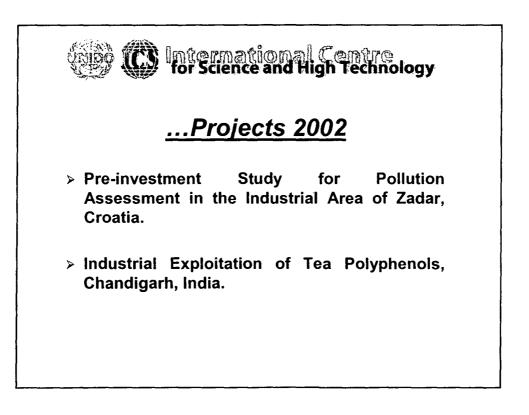




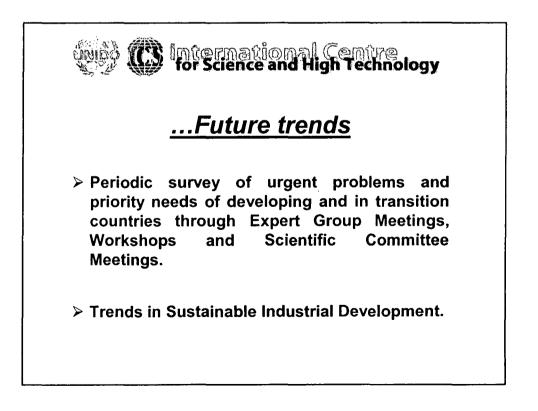














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Overview of the Importance on Industry, Trade and Tourism in Coastal Areas: Mining Activity in Sardinia, in Iglesias and its Territory by L. Ottelli

Overview of the Importance of Industry, Trade and Tourism in Coastal Areas: MINING ACTIVITY IN SARDINIA, IN IGLESIAS AND ITS TERRITORY

When the dossier on the value of the millenial mining activity on our island was presented, the attention of 170 Unesco members was impressed. They carefully listened to the long and complex geological events that led to the exploitation of the considerable deposits of the most differet minerals, that saw man struggle and suffer to mine the most secret treaures of the island subsoil.

The mining activity tells us of 8000 years of history in which our land has been object of reserch and mine working of various minerals, apart from being a meeting point for different cultures and polulations coming from all over the Mediterranean sea.

During those 8000 years, the searched and used materials have naturally changed, going from obsidian to gold, according to the change in the interested populations and, still today, this search and exploitation of new minerals go on.

The decline of the mining activity in Sardinia, that brought to the almost complete closure of the mining plants, has not been followed by the deserved attention and sensibility. Every mine, in fact, once the useful mineral has been depleted, can be reused for all its cultural, scientific and human contents that naturally possesses but that have been incremented during its development.

It is also true, however, that the mining activity leaves on the territory, heavy marks of itself, deeply modifying the landscape with superficial minings, dumps and decantation basins. Such marks represent a kind of visual pollution but, above all, a presence of heavy metals on soils and waters.

Different projects of depollution are being fulfilled in the Iglesiente area. Concerning heavy metal pollution, for instance, the main problem is represented by the settling tank of washery muds and heaps of metallurgic waste, that have proved to be sources of environmental pollution in a few cases and are, in general, a potential risk of environmental damage, and require urgent safety engineering intervention, impact minimisation and monitoring. Finally, mining activity has affected the hydrogeological system, the landscape and land morphology, and has caused significant impoverishment of woods that have been cut and used to support underground cavities or the production of energy.

The problem of soil and water pollution caused by mining activity became apparent in the mid-ninteenth century as a result of large amounts of extracted materials and the absence of environmental protection. With the abandonment of mining activity, more or less pronounced, diffused and canalised drilling processes have caused collapses and erosion near the tailings heaps. This led to the unconditioned transport of fines and their discharge into watercourses and soils, causing both surface and deep chamical contamination, a less apparent risk though, in the long term, the most dangerous.

Surveys seem to indicate that the heaped mine tailings amount to over fifteen million cubic meters, not counting the waste that a few mines have discharged directly into the sea.

According to the Mine Office, the quantity of material discharged in rivers between 1949 and 1958 amounts to five millions cubic meters, but specific studies on heavy pollution have only been carried out in recent years.

As a result of a few geochemical and hydrogeochemical survey carried out on vast portions of land affected by mining activity, sampling of the most representative parts of the spoil bancks confirmed the presence of high contents of Pb (>5000 ppm), As (>100 ppm) and Cd (>100 ppm), above legal toxicity limits provided for industrial wastes.

There are not specific studies on the effects on air contamination, though several environmental anlyses have shown that a few mining

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areas particularly exposed to the dominat winds present accumulations of sediments from the spoil banks and a significant amount of dust.

However, it has to be pointed out that, the mining activity through history, gives Sardinia a peculiar aspect. Its geological, mineralogical, paleontological, hydrological elements, its biologic rarities and endemics, its humid zones, its spectacular landscapes in the morphology of its coasts, internal mountains and underground cavities and all the archeological remains deeply related to the mining areas, constitute, all together, a perhaps unique heritage in the world.

Nevertheless, among these characteristics, it was the geological and mining exigencies that most intensely spelled out the social and cultural history of the island, an island with strategic natural resources, deposits whose variety and size have attracted Mediterranean and other populations from the most ancient times up to recent years.

The mineral richness of Sardinia appears through the innumerable mines scattered in its territory. All those mines have different productive, scientific and cultural values, but are all indispensable to understand the evolution of events in over 8000 years of uninterrupted exploitation. Mining history reflects the history and culture of the Mediterranean man and this is the reason why the Unesco interest has arisen together with the following recognition of the "PARCO GEOMINERARIO STORICO ED AMBIENTALE DELLA SARDEGNA" as a proposal of valorization of a cluster of cultural and environmental values that few other countries posses. And, in this contest, the Iglesiente territory is the one the most represents the evolution of mining through the centuries.

The mining activity has, in fact, left on the territory witnesses of Phoenician-Carthaginian, Roman and Pisan processing, the last one of which for the extraction of silver from the smelting of galena. (XIII and XIV century).

During the Pisan period, the Gherardesca family was the most distinguished and from 1256 onwards, ruled the South-Western part of the island, the so called Sulcis-Iglesiente. As already pointed out, this part of Sardinia was particularly rich in minerals and mines and the Pisans left many first-hand accounts of their activity as well as the first mining codex of the island entitled "BREVE DI VILLA DI CHIESA DI SIGERRO".

The Breve is a set of operational and behavioural norms and provisions that all citizens were to respect. "... The exploitation of mines in Sardinia was mainly for silver. Workings for silver-bearing ore were made through thin shafts from the surface along the metalliferous vein. At times, an incredible number of shafts were made expecially in the calcareous mountains neighbouring Iglesias... sometimes the shafts or pits were dug down to depths of more than 80-100 meters and, in some places, even to 200 meters". This fervid activity reached its apex in the second half of the XIII century with Ugolino della Gherardesca dei Conti di Donoratico and his sons Guelfo and Lotto.

Silver had at that time a higher value compared to today and the amount of material extracted in the Iglesiente area was so big to justify the building of a mint in which two kinds of coins were struck: the Aquilino d'argento and the Aquilino minuto.

The breve was unfortunately the only document escaped from the big fire of the city short after 1350.

The other witnesses of the mining activity in the area are from XIX and XX century, a period in which great engineering works were made that gathered the attention of the entire industrialized world. Some of those works are being recovered and they will soon be given back to the community for didactic, technical, scientific and turistic aims.

Different examples of the technical skills and of the sacrifices made by the workers can be found all over the territory as well as thousands of historical and technical documents testifying the different ways of working and the most futuristic techniques used. In this context, one of the most curious engineering work realized in Iglesiene is "Porto Flavia" built on the cliffs on Masua in 1924 by Eng. Cesare Vecelli who named the work after his little daughter Flavia. Porto Flavia was built in order to

4

load the ore directly onto the ships and thus avoid the transportation by ox-driven carts or by the workers who carried the ore in baskets on their backs. This system had a loading capacity of 400 tons/hour and in a few hours it could load ships that would have taken days to load with the old system.

The peculiar mining habitat in which such important architectures are present is being preserved and enhanced. It is a incomparable patrimony that tells us the great history of our island and the man sacrifice in this hard and dangerous activity.

Luciano Ottelli

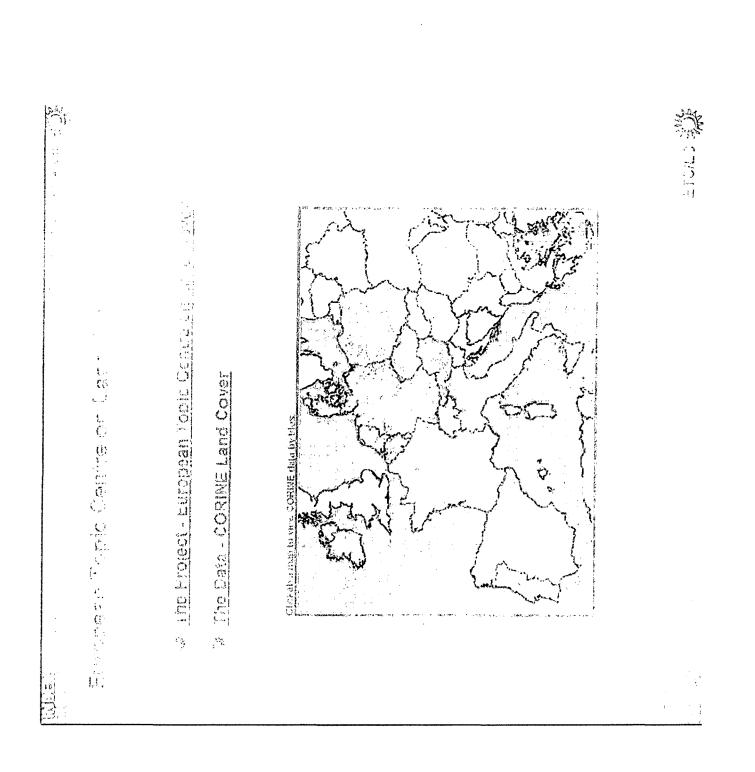
June 2002

Definition of Land Use/Land Cover for Planning Industrial Development by E. Feoli Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas"

Cagliari 4-8 June 2002

Enrico Feoli

Definition of Land Use/ Land Cover for Planning Industrial Development



1.1. The CORINE programme

overview of existing knowledge, and information which is as complete and up-to-date as possible on changes in certain If our environment and natural heritage are to be properly managed, decision-makers need to be provided with both an features of the biosphere.

To this end, the three aims of the CORINE (Coordination of information on the environment) programme of the Commission European are: · to compile information on the state of the environment with regard to certain topics which have priority for all the Member States of the Community;

to coordinate the compilation of data and the organization of information within the Member States or at international level;

to ensure that information is consistent and that data are compatible.

On 27 June 1985 the Council, on a proposal from the Commission, adopted a decision on the CORINE programme. This Commission work programme concerns 'an experimental project for gathering, coordinating and ensuring the consistency of information on the state of the environment and natural resources in the Community' (Official Journal L 176, 6.7.1985).

the tte A further objective of the CORINE programme is to bring together all the many attempts which have been made over the years at various levels (international, Community, national and regional) to obtain more information on the environment and * creating a geographical information system to provide the information on the environment which is essential when devising procedures for collicting, standardising and exchanging data on the environment in the EC Member States; objectives: In order to determine the Community's environment policy, assess the effects of this policy correctly and incorporate environmental dimension into other policies, we must have a proper understanding of the different features of programme's the to meet taken heen the geographical distribution and abundance of wild fauna and flora. the quantities of toxic substances discharged into environments, of complementary action have the geographical distribution and state of natural areas. preparing and implementing Community policies. the quality and abundance of water resources. Iand cover structure and the state of the soil, the state of individual anvironments. lists of natural hazards, etc. the way it is changing. Seudyl Two main environment:

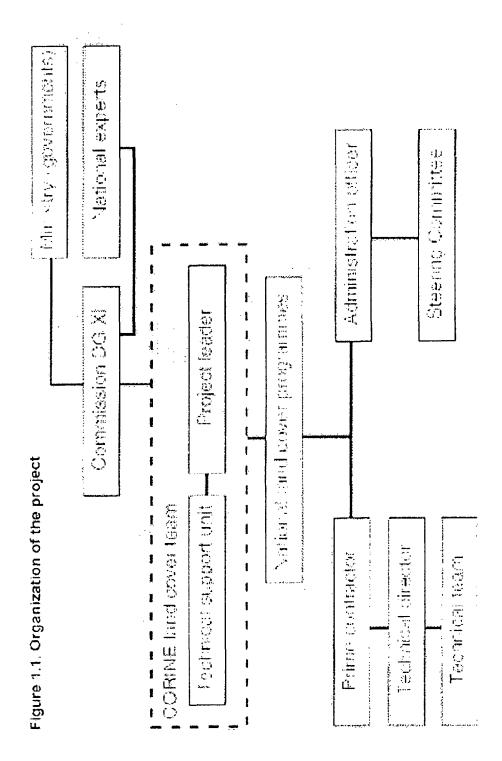


Table 1.1. The CORINE land cover project in figures

• • • • •

- Area covered: 2.3 millions km²
- * 12 countries
- * from 62° N (The Faeroes) to 28° S (Canary Islands)
- * from 14' W (Canary Islands) to 29° E (Kastellorizon)
- Working scale: 1/100000
- S * i.e. 1 500 standard map sheets produced using 10 different projection
- Area of the smallest mapping unit: 25 hectares
- * i.e. more than 700000 basic unit (polygons) or a vector data-base around of 1 gig

';

- Land cover nomenclature with three levels
- * First level: five headings
- * Second level: 15 headings

* Third level: 44 headings

National teams

DP specialists, geographers, photointerpreters and cartographers working on the producti satellite images, computer-aided interpretation and the digitization and integration na meetle into the CODME Ind answer detailorden

A central technical unit consisting of four specialists:

Michel Bossard: Photo-interpretation

Yves Heymann: Project leader

Michel Lenco: Statistics

Chris Steenmans: Data processing

1.2. The CORINE land cover project

The land cover project is part of the CORINE programme and is intended to provide consistent localized geographical information on the land cover of the 12 Member States of the European Community.

The project is necessary for the following reasons:

preliminary work on the CORINE information system showed that information on land cover, together with information on relief, drainage systems etc., was essential for the management of the environment and natural resources; information on land cover therefore provides a reference source for various CORINE database projects;

* in all the countries of the Community, the information on land cover available at national level is heterogeneous, fragmented and difficult to obtain.

determining and implementing environment policy and can be used with other data (on climate, inclines, soil, etc.) to make At Community level, in the CORINE system, information on land cover and changing land cover is directly useful for complex assessments (e.g. mapping erosion risks).

The benefits of using a single joint project to meet both Community and national (or even regional) needs considerably influenced the general features of the land cover project: scale, area of the smallest mapping unit and nomenclature.

1.2.1. Information on land cover

Until recently, it was generally assumed that in the tong term human activity had little lasting effect on the land thanks to nature's ability to restore itself. This view remained prevalent for a long time despite the fact that farming practices have been ceusing irreversible damage in certain areas for centuries. Over the last few decades, the effects of certain phenomene have shown that we do need to look after land cover and all its various components. These include:

the gradual desertification of certain regions.

the rapid disappearance of vast areas of forest.

* the wholesale of poor farmland,

* the gradual drying-up of wetlands.

continuous urban development along coastlines, etc.

rationally, information on land cover is essential. Yet, despite the urgency and the scale of the problem, confirmed by all the If the aim is to do more than resort to basic emergency action in the face of disaster and instead manage vast areas of land studies, progress in this area is limited and often disappointing.

making and updating land cover inventories. This may be because the serious nature of the effects of some of man's actions on the biosphere has only recently been fully understood, or because data compilation and management techniques did not leritories and keeping up-to-date inventories and maps of land ownership have never seriously considered the problem of Those industrialised countries which have devoted considerable resources to producing large-scale maps of national previously lend themselves to this type of operation.

As a result, information on land cover has been available only for small areas affected by urban development, agricultural development, major infrastructure projects, etc. Against this background, the CORINE land cover project launched by the Commission of the European Communities sets out to meet a new need and provides support for the Commission in its efforts to use and develop advanced datacompilation Over the last decade, awareness of serious environmental problems has lent new urgency to the question of land cover and management techniques in carrying out its policies. inventories.

We can no longer rely on thematic information from topographic maps which are updated every 10 or 15 years, and even less use can be made of statistical assessments which are the by-products of agricultural and forestry inventories and surveys, when attempting to cope with the following:

- * disappearance of areas of wetlands,
- destruction of Mediterranean woodland by fire.
- intensification of agriculture in vulnerable areas,
- ' development of tourism along coastlines.
- emergence of scrub on farmland,
- disappearance of species deprived of their biotopes, etc.

For environmental management purposes, land cover information will have to meet special requirements: it must be cortographic as well as statistical and it must be possible to reproduce the information at different scales to be useful at various decision-making levels.

Scale	Needs met	Tyne of devision
1:1000 000		Guiding national and Community programmes and preparing major development programmes
1:1 00 000	National management of the environment: identifying and locating major problem areas	Monitoring the implementation of Comn and national policies
1.25 000	National land use policy: deciding which areas to protect Monitoring regional land use Managing sensitive areas	Local management

1.2.2. Origins of the project

in small areas. This is particularly evident when attempting to compare the land cover statistics of several countries or to Previous projects in cortain countries have shown how difficult it to delineate, categorise and map types of land cover even collect all the information available on land cover in one single country.

success in compiling land cover inventories of vast areas. The marketing of geographical information software has also made it easier to use the information from these inventories and to update them more rapidly. In fact the Council's recommendation to use satellite data is well justified in the particular case of the CORINE land cover project. However, over the tast 15 years or so, the Jaunching of Earth observation satellites seems to have improved the chances of

Earth observation satellite data have the following characteristics:

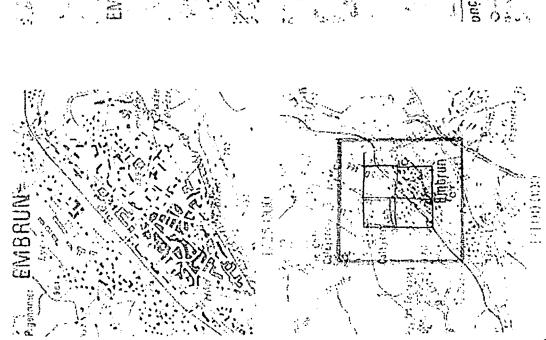
and taking account of cloud patterns over certain countries, it is estimated that all the countries of the can be covered twice a 1 they are available on a regular basis for all points on the globe (subject to there being no cloud cover); data may be acquired every 16 days in the case of Landsat and every 26 days in the case of SPOT; with the satellites currently in service, year.

they are inexpensive.

2.1. The mapping scale

The scale chosen for the project is 1:100 000.

Figure 2.1. Cartographic scales





There are several reasons for this choice:

Commission. At such scales, the size of the smallest unit mapped is very large and the corresponding nomenclature includes "Land cover data provided on smaller scales (1:250 000, 1:500 000) are not detailed enough to be useful to the relatively few headings (e.g. level-2 nomenciature, with about 15 headings).

" It is well suited to serve as a basis for specific studies at larger scale within a country, such as preliminary investigations for civil development projects or environmental protection.

* It (e.g. coastal erosion), but it is also compatible with projects which use a smaller scale (e.g. 1:1000000), since space-bome remote sensing allows generalisation at that basic cartographic scale. "It is consistent with budgetary constraints and time limits for carrying out such a programme in the 12 Community countries.

. The maps can be fairly easily updated on a regular basis.

This is a common mapping scale for most Community countries, albeit with a wide range of projection systems (see Tuble) 2.1.).

(in and the second se		
Country		Scale	Projection	h.	Number of sheets = 2/2	 جر س
Belgium	(B)	1:100 000 Lambert 1:50 000 Lambert	Lambert Lambert		া না দা শেষ লো	· · · · · · · · · · · · · · · · · · ·
Denmark	(DK)	(DK) 1:100 000 UTM	UTM			
Germany -former Federal Republic of Germany -former German Democratic Republic	<u></u>	1:100 000	1:100 000 Gauss Kruger		1151	··· ·
Greece	(GR)	(GR) 1:200 000 UTM 1:100 000 UTM 1:50 000 Azim	:200 000 UTM 54 :100 000 UTM 7100 000 134 :50 000 Azimuthal equidistance 333	stance	54 - 34 333 4	; 、
Spain	(F)	1:100 000 UTM	UTM		297	
France	(E)	1:100 000 Lambert	Lambert		203	
Ireland	(IRL)		1:126 720 Transverse Mercator	ltor	20	
lialy	(1)	1:100.000	1:100 000 Gauss Boaga		N.C	
Luxembourg	(11)	1:100 000 Gauss	Gauss		_	
Netherlands	(NL)	1:100.000	1:100 000 Stereographic		×	
Portugal	(d)	1:100 000 Bonne	Bonne		55	
United Kingdom Northern Ireland	(UK)	1:50 000 1:250 000 1:126 720	1:50.000 Transverse Mercator 1:250.000 Transverse Mercator 1:126.720 Transverse Mercator	ator ator	204 17 2	
Total number of sheets (1:100000 and 1:126 720)	000 and	1.126 720)			1413	

Definition of the unit area and size of the smallest unit mapped 2.2.

Thematic mapping of the biophysical cover of the Earth's surface may be approached from two 5/Herent angles:

Land cover essentially concerns the nature of features (forests, grops, water bodies, bare rock, etc.).

Land use is concerned with the socio-accionnic function (agriculture, habitat, environmental protection) of basic surfaces.

Since the aim of the CORINE programme is to provide users with information on the state of the environment, a cartographic inventory of land cover is required; this is the CORINE land cover project.

The feasibility study showed the land cover project's technical unit that the use of satellite data for the land cover mapping of extensive surfaces would not only raise a nomenclature problem, but would also necessifiate detailed consideration of the unit area to be mapped.

Remarks on land cover mapping

In many cases thematic mapping is carried out for economic management purposes and is therefore concerned with land 0.56. Land cover maps often implicitly adopt the cadastral land unit as the unit area. However, the cadastre, a unit area recognized in law is essentially limited to large-scale burd use maps. The cadastral unit is very often heterogeneous in its composition and its area can vary from a few square metres to several thousand hectares.

With vegetation maps, which are a form of land cover map, the problem of the unit area (to be) mapped has not arisen

The results of centain statistical surveys (arealar surveys of land use/cover) might suggest that the problem of the unit area is not really problematical. However, it must be remembered that these surveys sample points, not reveaces. Nonetheless, in the field it is still possible to classify a point as an item in a land cover nomenclature.

2.2.1. The unit size area

To assist the understanding and use of this unit, we have decided to define a by loting its main character stos.

- cumespands when to an area whose ower may be considered, homogeneous (press, when, in lows), etc.) or to a combination of elementary areas (homogeneous as defined above) which, in its voriations, represents characteristic land rower structures (covering large surfaces which can be considered to consitute a single type of land cover in the Member States of the Community);
- given the scale, the unit must represent a significant area of land.
- it is clearly distinguishable from surrounding units.
- its structure in terms of land cover is stable enough to serve as a unit for the collection of more precise information.

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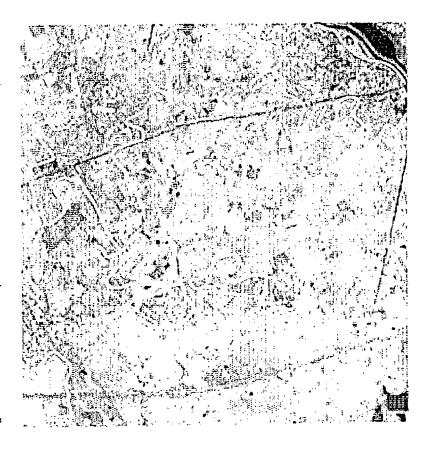
1.4.1.1.1

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Figure 2.2 - Unit area example: 1.1.2 discontinuous urban fabric, Landsat TM intage



In the CORINE land cover project the unit area to be mapped has two functions:

conceptual tool for land cover analysis;

format - a particular representation of the heterogeneity and diversity of the objects covering the Earth's surface. * a tool for reading and organizing spaceborne remote-sensing data which constitutes - in the images analogue

combination of surfaces which are to a greater or lesser degree homogeneous/heterogeneous, whatever the scale When defining this unit, it must always be borne in mInd that, in reality (in the field) land cover always occurs as a used. Furthermore, irrespective of how they have been processed, data acquired by spaceborne remote-sensing systems do not provide a representation of the actual land cover situation; nor can land cover be mapped in all its complexity/diversity.

meets the following two requirements: (a) its content must provide the thematic data required by the users, and (b) it Given these circumstances a spatial/statistical unit must be conceived, for all land cover mapping activities, that must provide an acceptable representation of reality.

2.2.2. Size of the smallest unit mapped

The surface area of the smallest until mapped in the project is 25 heaters.

Establishment of the minimum surface area to be mapped must comply with three basic requirements:

" legibility of the printed map, or in the case of the land cover project, easy digitization from the interpretation manuscripts; · it must provide a representation of the essential features of the terrain in terms that serve the thematic objectives of the project; it must represent a trade-off between project operating costs and provision of land cover information requirements within overall project budgetary constraints.

Taking these requirements into consideration, project managers set the minimum unit mapping size at 25 hectares.

On a scale of 1: 100 000, 25 hectares is represented by a 5 x 5 mm square or a circle with a 2.8 mm radius.

2.3. The nomenclature

In any land cover cartographic inventory four elements are inextricably linked:

* the scale.

* the surface area of the smallest unit to be mapped,

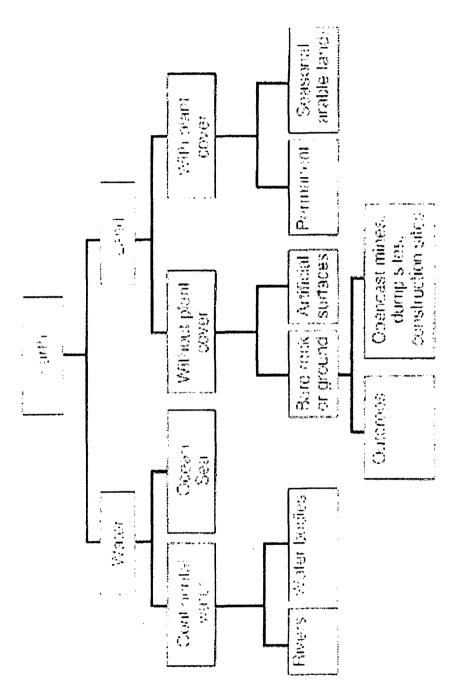
* the nature of the basic information used, in this case. Earth observation satellite data,

* the structure of the nomenclature and the number of items it contains.

On the basis of the first three elements listed above and the provisional nomenclature used for the feasibility study, the land cover team has formulated the definitive nomenclature for the project.

This nomenclature and accompanying definitions have been the object of numerous discussions both with the final users of the CORINE database and with various experts in the Member States.

The following diagram (Figure 2.3) sets out the 'logical framework' which was used as a hasis for establishing the nomenclature. Figure 2.3 - Theoretical schematic construction of a land cover nomenclature



Based on this logical framework, the selected nomenclature must meet a certain number of requirements:

* it must be possible to map all Community territory; in other words there can be no heading for 'unclassified land';

* the headings must correspond to the needs of future users of the geographic database, i.e. for the CORINE base, the state of the environment: * heading terminology must be unambiguous and avoid the vague terms often resorted to by many photointerpreters when encountering uncertain areas.

Moreover, it must be remembered that the nomenclature applies to relatively large units, i.e. 25 hectares or more, and the total number of headings in a nomenclature is always a compromise between the user's needs and the financial constraints affecting the project.

Explanatory notes

(i) Terminology

Most nomenclatures used for mapping or statistics relating to space are land use nomenclatures produced for the purpose of compiling an inventory of human activities.

The terminology available (to designate the items in a land cover nomenclature) therefore relates to land use, so specific terminology for land cover inventories has yet to be developed. It is fairly difficult, in one project, to establish a nomenclature which encompasses every single aspect of the area unit to be mapped.

(ii) Level of geographical application

The CORINE land cover nomenclature, as set out below, comprises three levels:

• the first level (five items) indicates the major categories (abstract to a greater or lesser degree) of land cover on the planet;

the second level (15 items) is for use on scales of 1:500 000 and 1: 1 000 000;

* the third level (44 items) will be used for the project on a scale of 1: 100 000.

(iii) National and/or regional nomenclature

The CORINE land cover nomenclature is organized on three levels; a fourth level could be added for some or all of the items, subject to the following requirements: - additional items must include all the land covered by the corresponding level-3 item (four-figure codes are used for those items only);

· newly created items must not relate to more than one three-figure item;

• the CORINE land cover mapping (three-figure nomenclature) must be completed prior to initiation of level-4

mapping.

Finally, care must always be taken to see that newly created items are compatible with the scale, the size of the smallest area to be mapped and the basic information, i.e. satellite data.

Table 2.2. CORINE land cover nomenclature	over nomënclature	
Level 1	u Level 2	Level 3
1 Artificiet surfaces	1. Urban fabric	1.1.1. Contrinuous urban fabric
		1.1.2. Discontinuous urban labric
	 1.2 industrial, commercial and trensport units	1.2.1. Industrial or commercial units
		1.2.2. Road and rail networks and associated land
		1.2.3. Port areas
		1.2.4, Airports
	1.3. Mine, dump and construction sites	1.3.1. Mineral extraction sites
		1.3.2. Dump sites 👘 🖉
		1.3.3. Construction sites
	 Artificial non-agricultural vegetated areas 	1.4.1. Green urben areas
		1.4.2. Sport and leisure facilities $1,2,5$ for $h_{1}h_{2}h_{2}h_{2}h_{3}h_{3}h_{3}h_{3}h_{3}h_{3}h_{3}h_{3$
2. Agricultural areas	2.1.Arable tand	2.1.1. Nort-imigated arabie land 2.1.2. Permanently irrigated land
		2.1.3. Rice fields
	2.2. Penhanent crops	2.2.1. Vineyards
		 2.2.3. Fruit trees and benry plantations 2.3. Olive groves
	2.3. Pastures	2.3.1. Pastures

	2.3. Pastures	2.3.1. Pastures
	 2.4. Heterogeneous agricultural areas 	2.4.1. Annual crops associated with permanent crops
		2.4.2. Complex cultivation patlems
		2.4.3. Land principally occupied by agriculture, with significant areas of natural vegetation
· · · · · · · · · · · · · · · · · · ·		SEA16 .
 Forests and semi-natural areas 	3.1. Forests	3.1.1. Groad-leaved forest
		3.1.2. Confisious lorest
	f .	3.1.3. Mixed lorest
	3.2 Shrub and/or herbaceous vegetation association	3.2.1. Natural grassland
		3.2.2. Noors and heatmand
		3.2.3. Scierophylous vegetation
		3.2.4. Transitional woodiging shrub
	$3.3, {\rm Open}$ spaces with little or no vegetation	3.3.1. Beaches, dunes, and sand plains
		3.3.2. Bare rock
		3.3.3. Sparsoly vegetated areas
		3.3.4. Burnt areas

----3.3.5. Gladers and perpetual show 4.2.3. Interlice! Ilbus 5.1.1 Weter courses 4.1.1. Inland marshes 5.2.1 Coastaf lagoons 5.2.3 Sea and ocean 4.2.1. Saltmarsher 5.1.2 Waterbodies a.1.2. Peatrogs 5.2.2 Estuanes 4.2.2. Spaintes . . . 4.2. Constal wellands 4.1. Inland wetlands 5.2 Marine waters 5.1 Inland waters 5. Water bodies ä 4. Wetlands

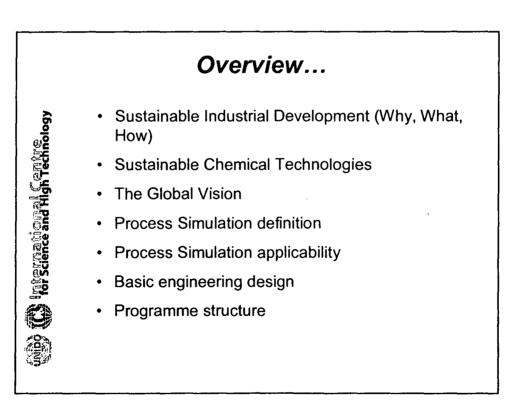
Process Simulation for Sustainable Industrial Development by G. Longo

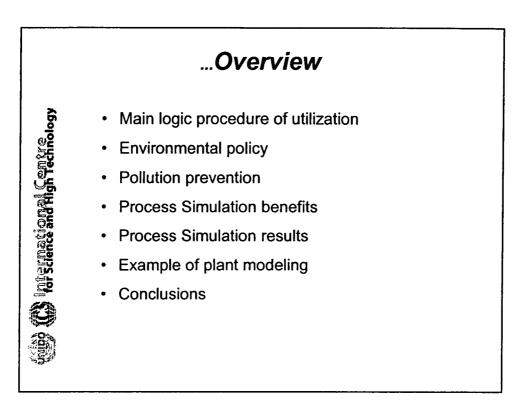
Process Simulation in Cleaner Production and Sustainable Industrial Development

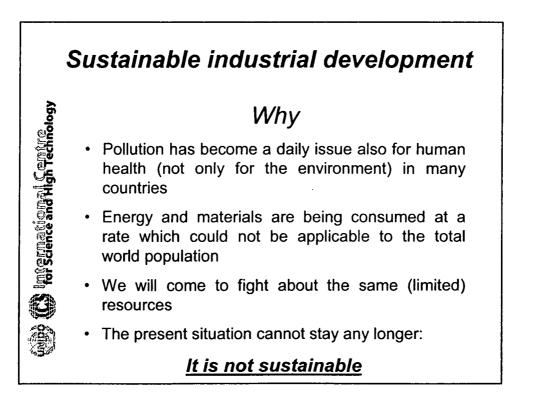
Gennaro Longo

Der Science and Algo Technology

Earth, Environmental and Marine Sciences and Technologies Area ICS-UNIDO AREA Science Park, Padriciano 99, Building L2, 34012 Trieste, Italy Tel: +39-040-9228104, Fax: +39-040-9228136, E-mail: gennaro.longo@ics.trieste.it







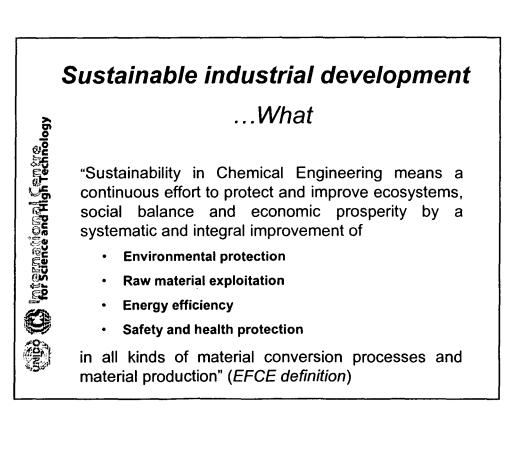
Sustainable industrial development

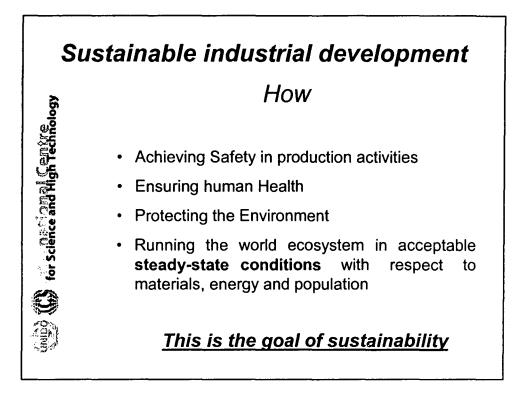
What...

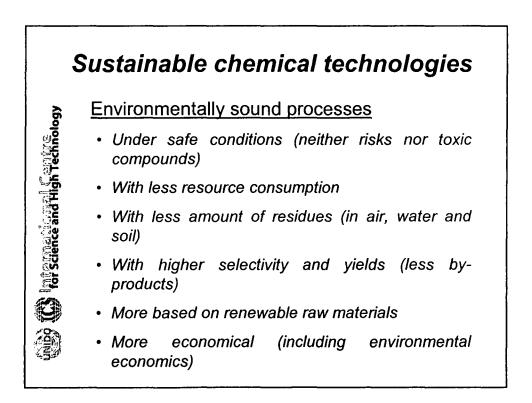
"Sustainable development is the means of improving the quality of human life with the carrying capacity of the supporting ecosystems" (UNEP)

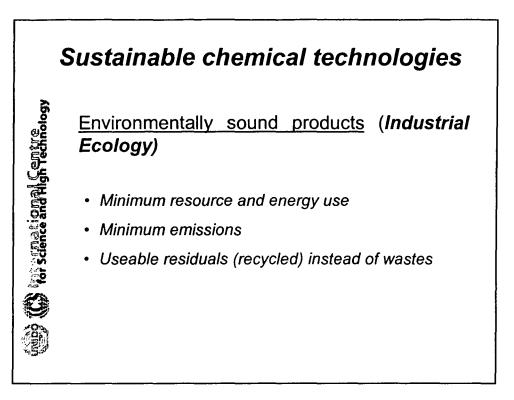
C Brscence and Hor Technology

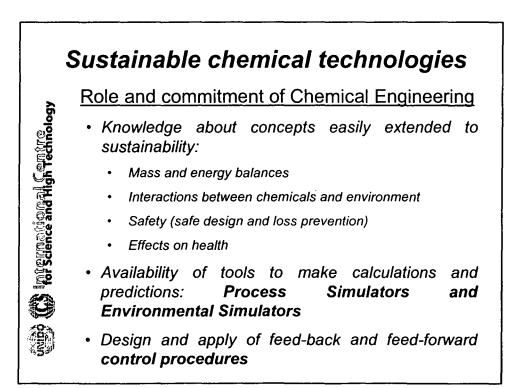
"Humanity has the ability to make development sustainable, to ensure that it meets the needs of the present without compromising the ability of future generations to meet their own needs" (*The Bruntland Report, 1987*)

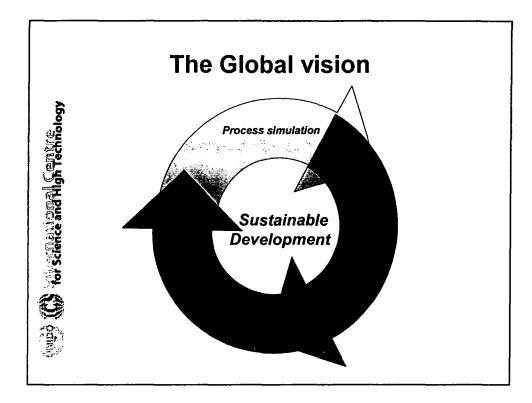


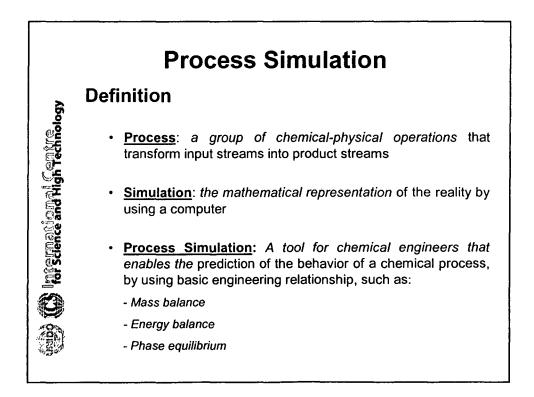


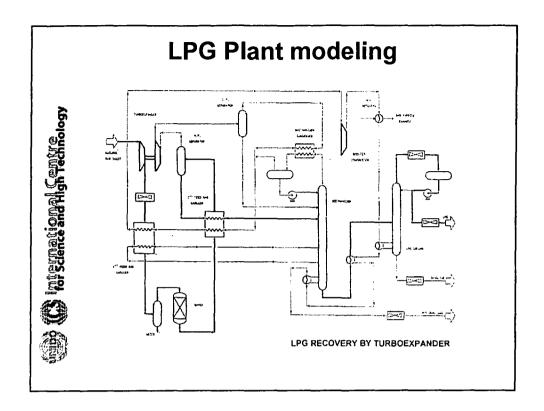


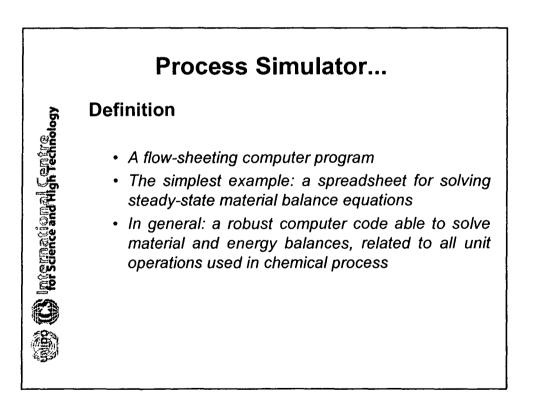


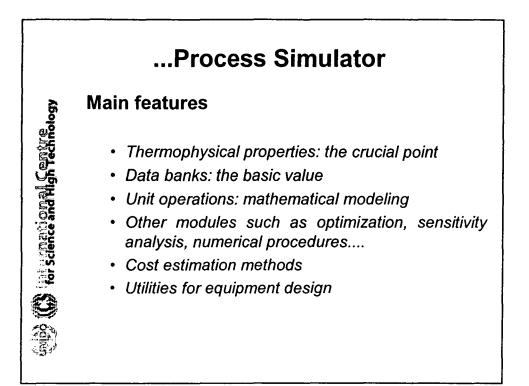


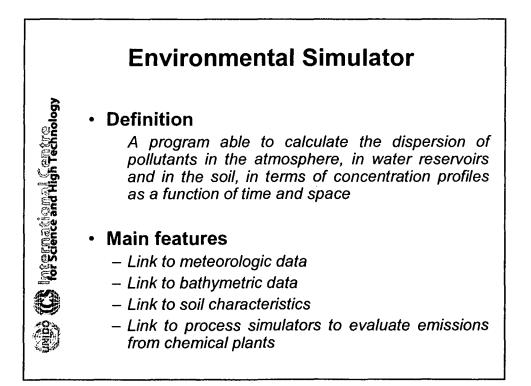


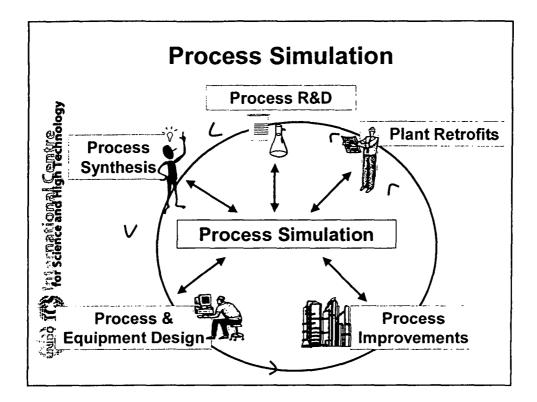


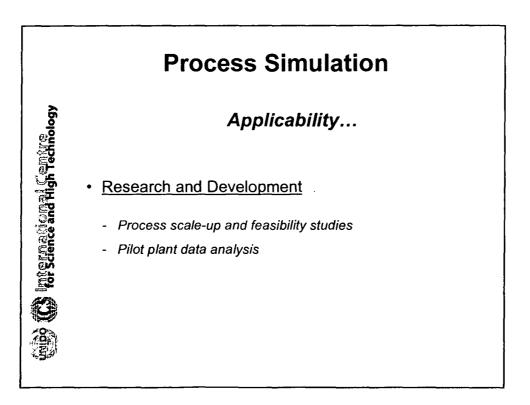


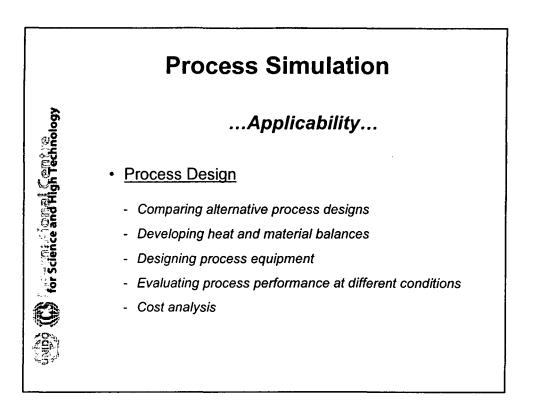


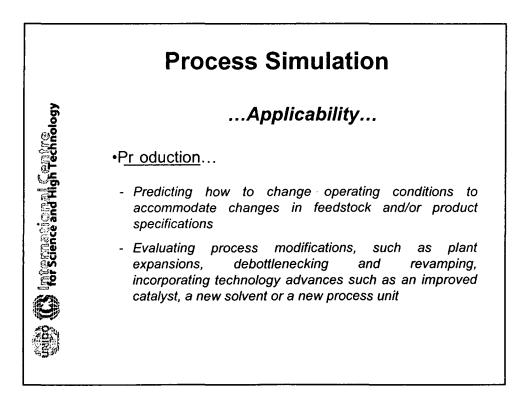


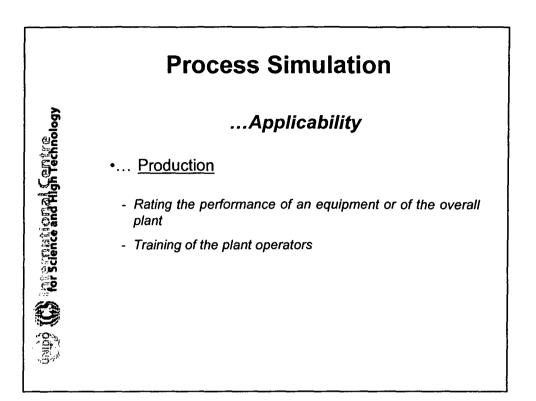


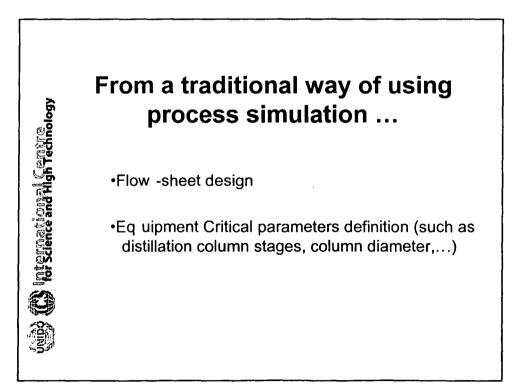










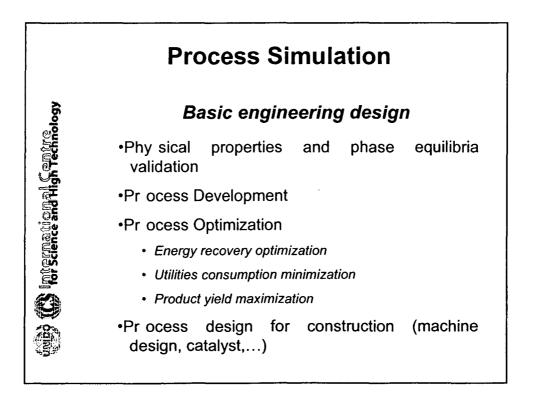


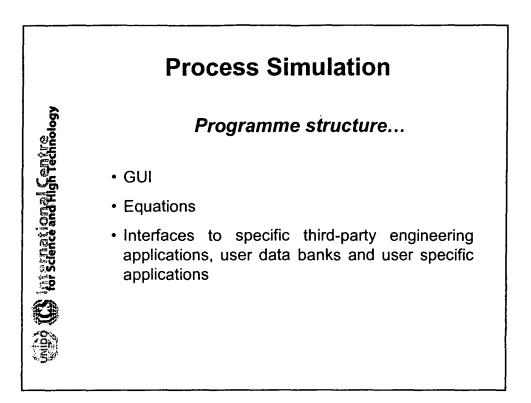
... to the comprehensive use of Process simulation in the entire 'life' of the plant

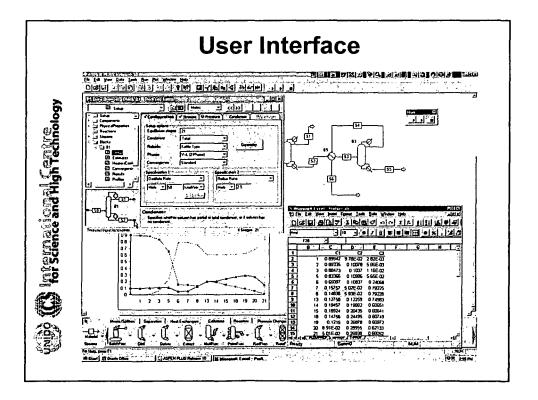
- Control strategies design
- Process parameters optimization (==> 'better' processes)
- Time evolution of the process (start up and shut down) (==> risk analysis)
- Operator training

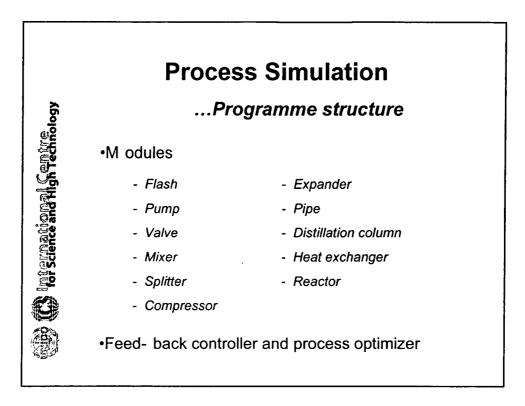
S International Centre For Science and High Technology

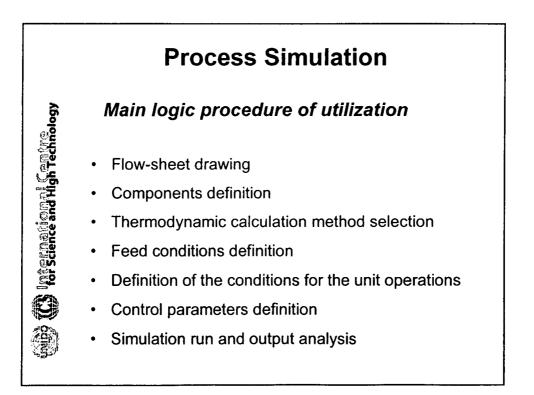
- Definition of procedure to reduce the unsteady state operations
- Process design

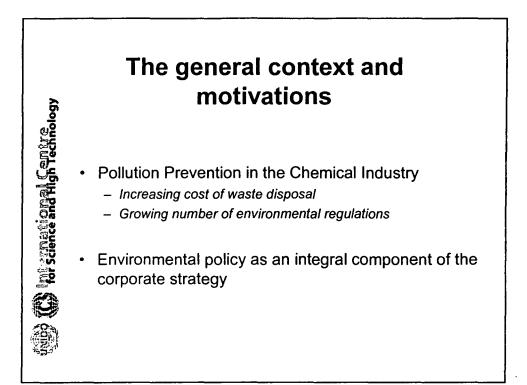


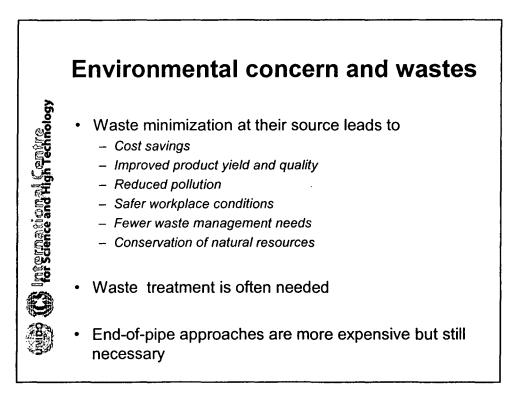


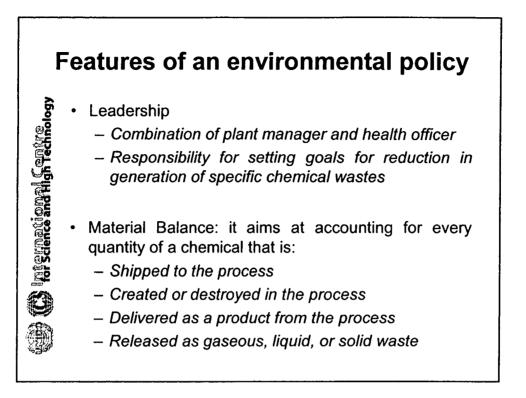


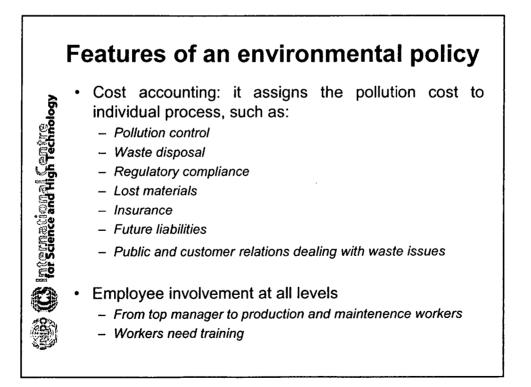


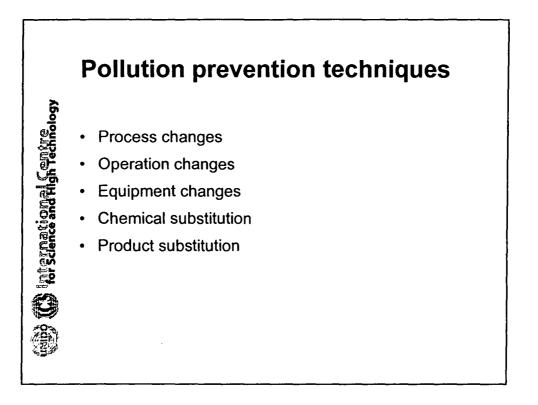


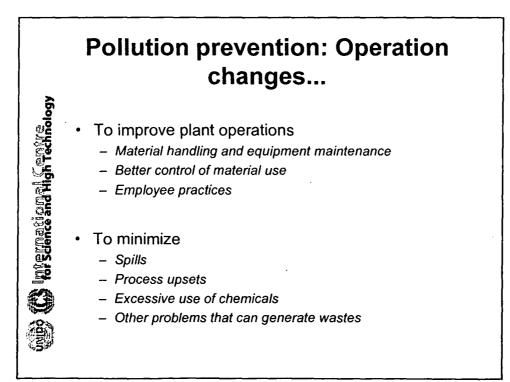


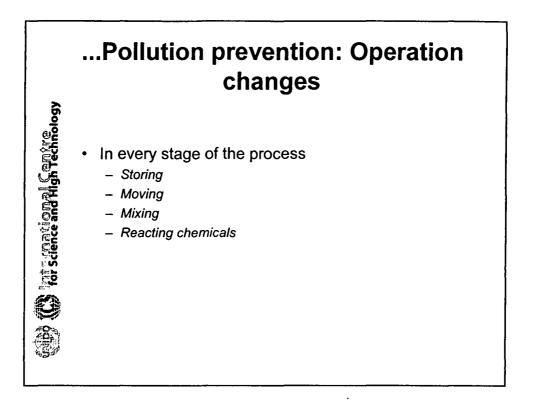


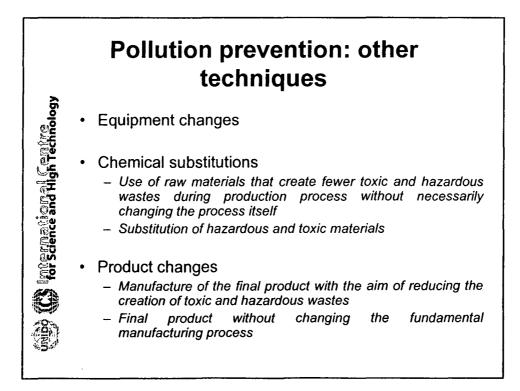


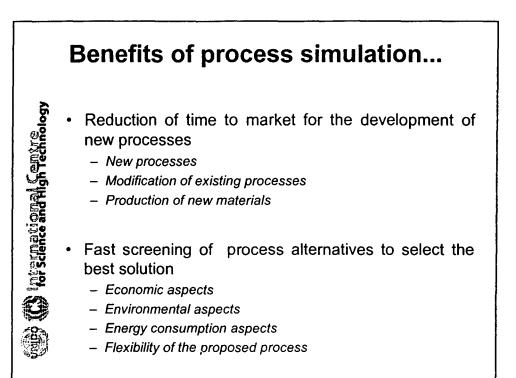


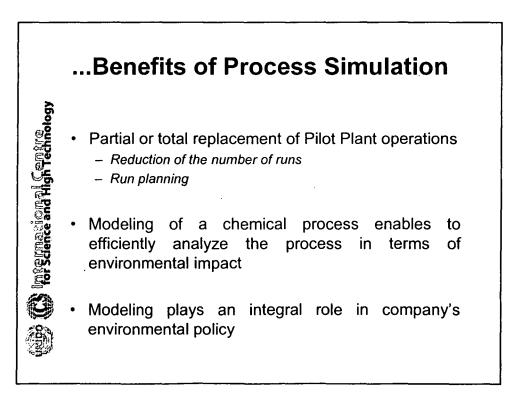


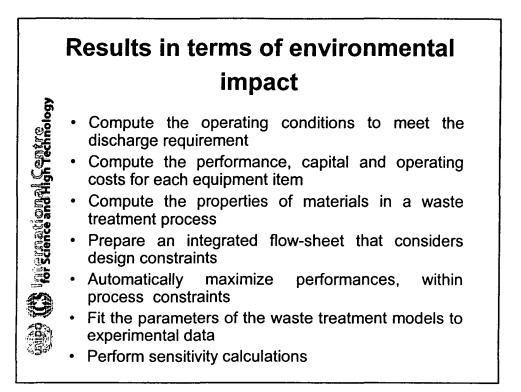


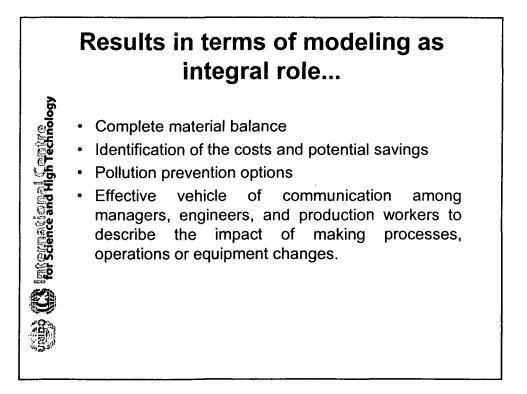










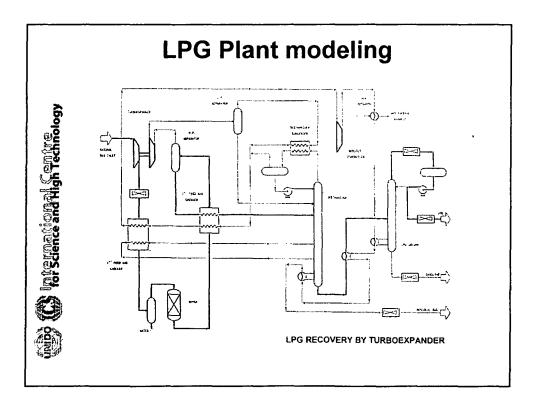


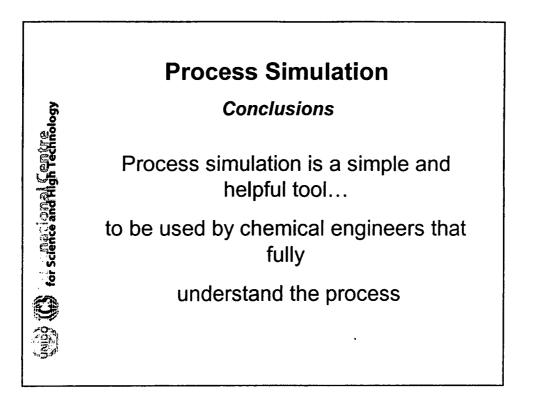
...Results in terms of modeling as integral role

· What if scenarios can be evaluated

Conternational Centre Contre

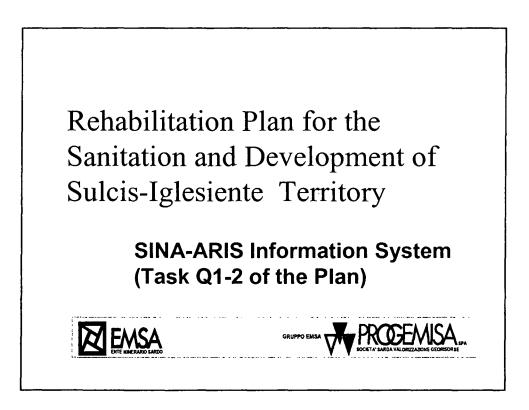
- · Model allows accurate support of pilot plant tests
- Process simulation aids the engineer in understanding the process design and in evaluating process alternatives
- The waste treatment process can be optimized to identify the operating conditions which achieve the most effective and economic treatment within regulatory constraints

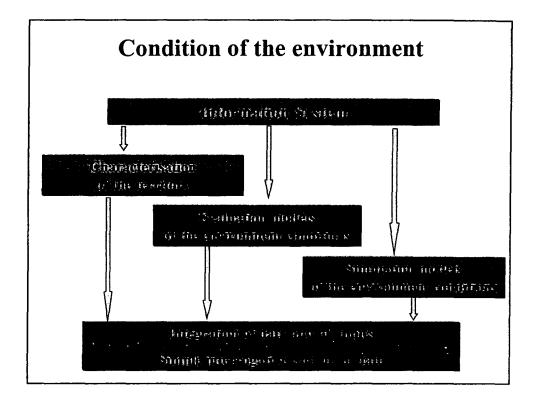


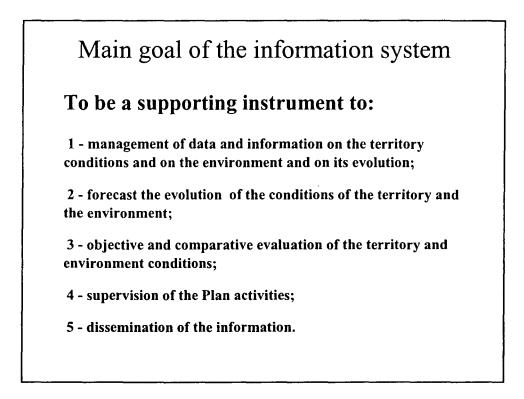


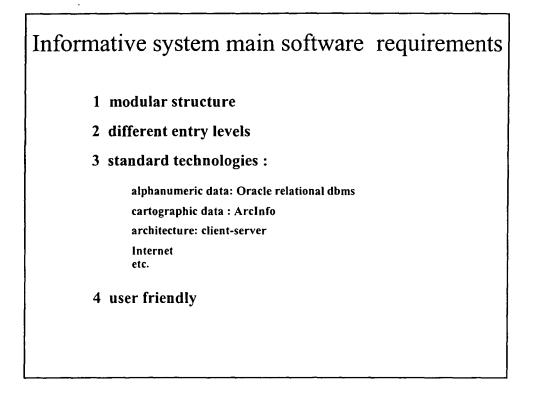
Technological Tools for Coastal Zone Management by G. Tore

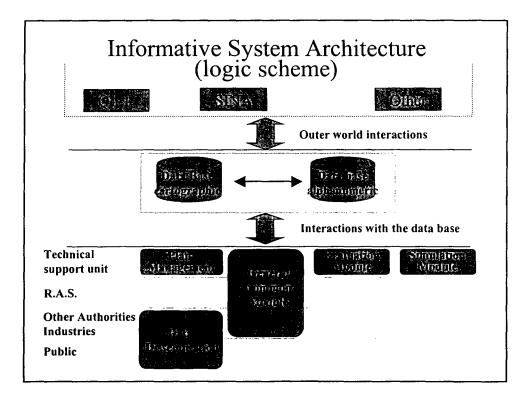


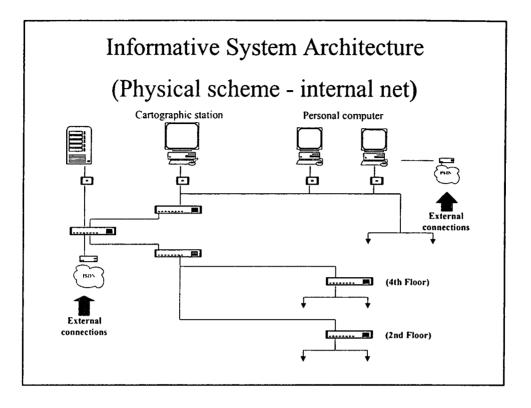


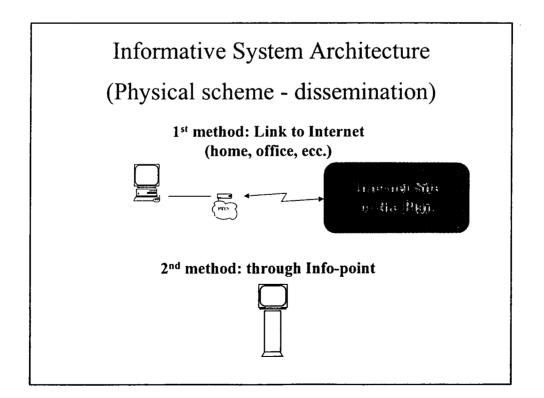


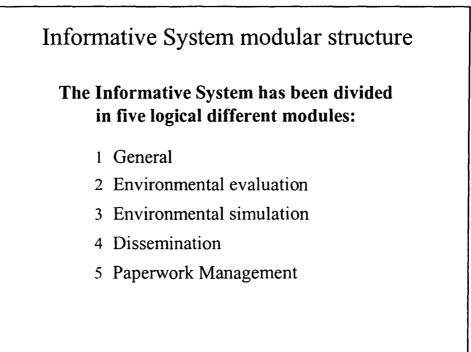


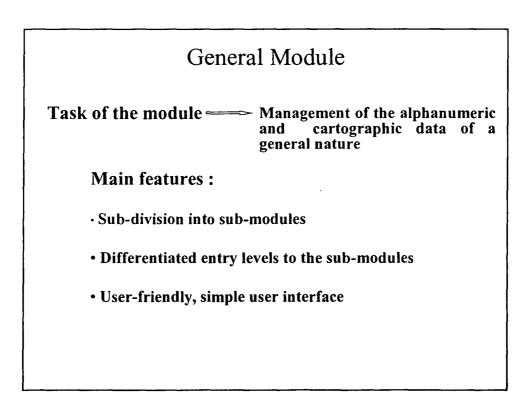


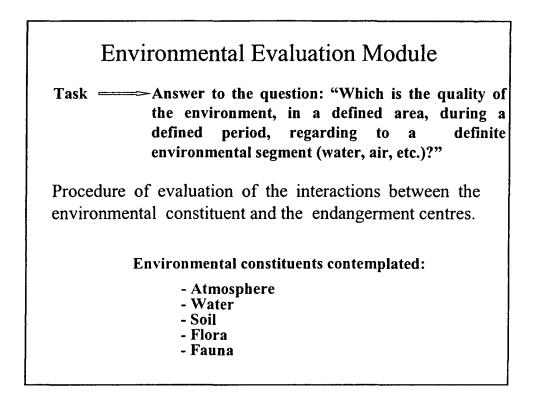


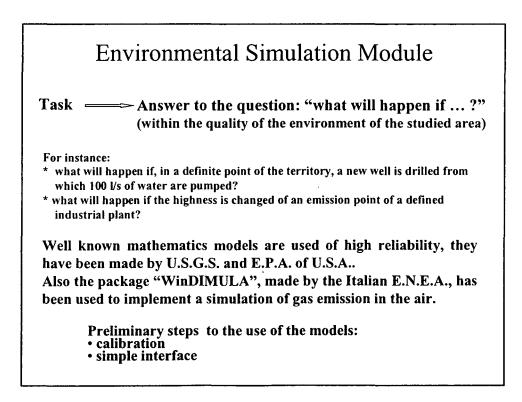


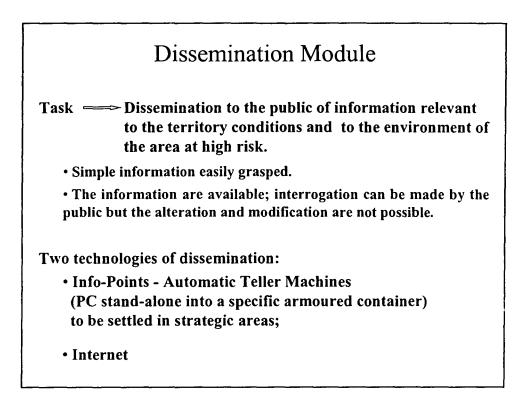


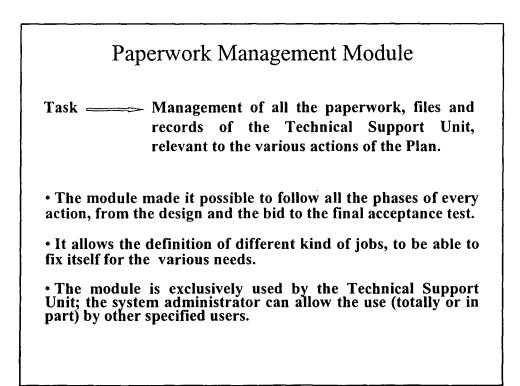


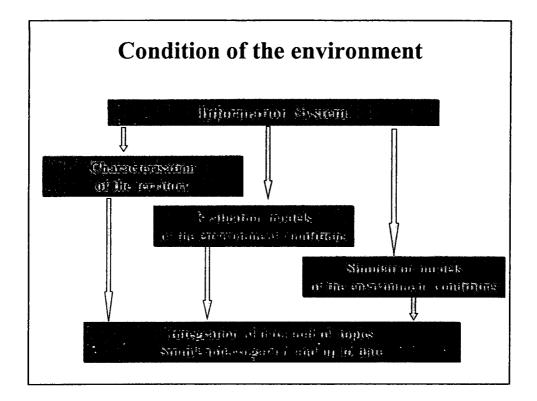


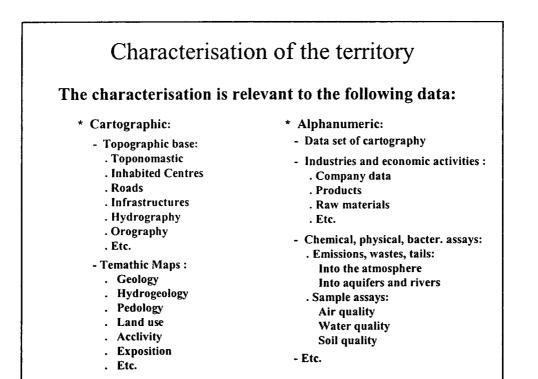


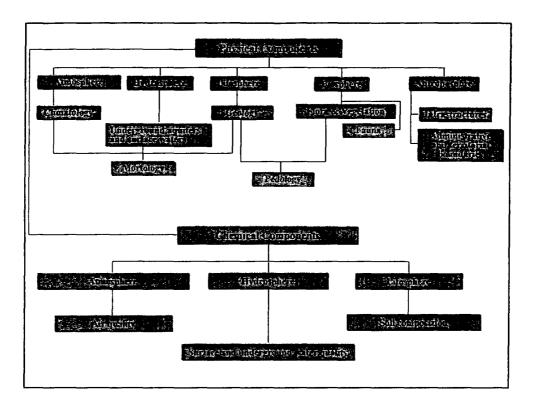


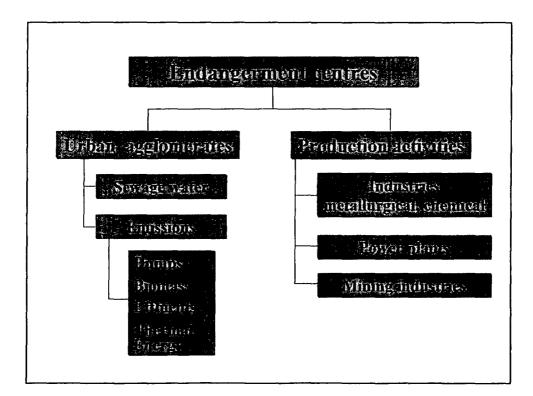


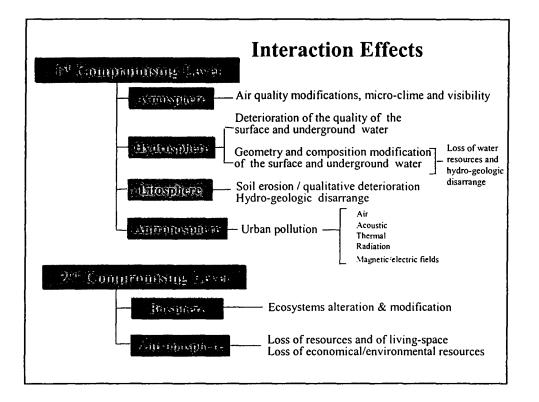


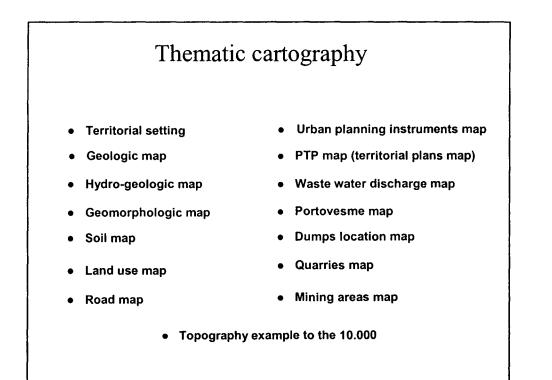








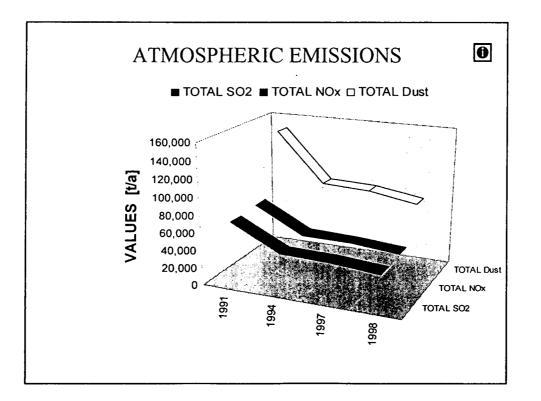


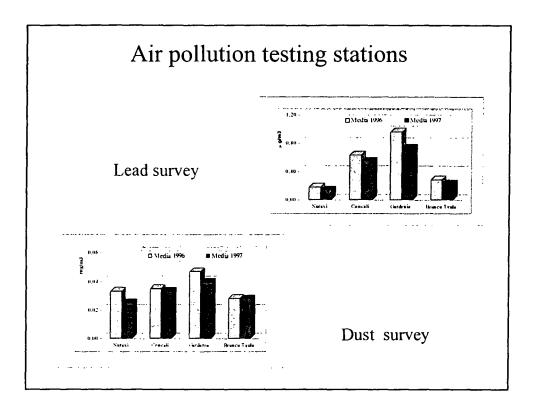


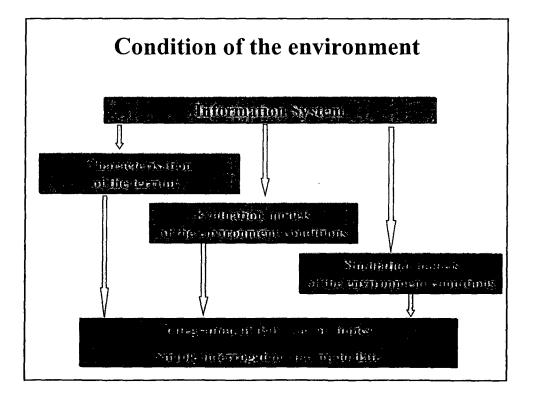
SO ₂ CONVEYE	D EMISS	IONS [t/:	a	
	1991	1994	1997	1998
ALCOA	3,000	2.800	3,000	3,000
ENEL **	50,000	21.000	19,410	15.80
ENIRISORSE	2,100	2.800	2,800	2,800
LURAL CUMINA	13,500	13,500	14.000	14.00
IGCC		-		-
41 X	200	160	120	120
SUAMAG	1,000	1.000	700	-
TOTAL	69,800	41,260	40,030	35,72

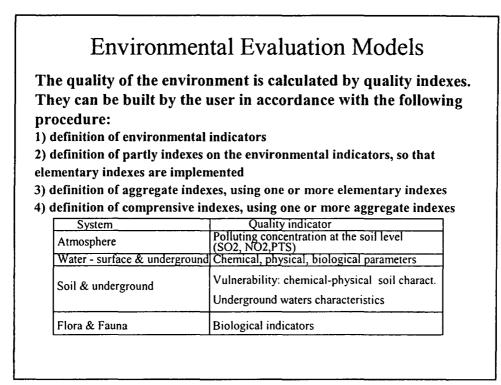
CONVEYED EMR	CONTRACT	NE N735	144.51	
	1991	1994	1997	1998
ALCOA	-			-
ENEL.**	~		8.450	11,300
ENIRISORSE				
EURALLUMINA		<u> </u>	1.700	1.700
IGCC				
Ш.А			<u>-</u>	
SEAMAG				w.
TOTALE	0	e .	19.150	13,000

CONVEYED DUST E.	MISSION	<u>5 [t/a]</u>	1	·
	1997	1994	1997	199
ALCOA	600	450	660	660
ENEL **	2,500	1,400	1.400	1,65
ENIRISORSE	80	70	7()	70
FURALLI MINA	1.100	1,100	4(11)	400
IGCC				_
<u>II.A</u>	26	26	10	10
SEAMAG	150	150	70	
	4.4511	3.13.1	2.5 (0)	





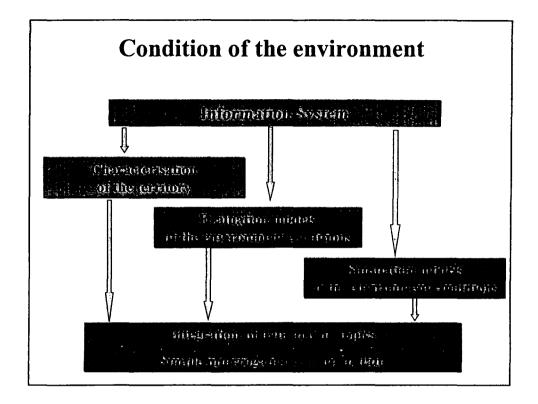


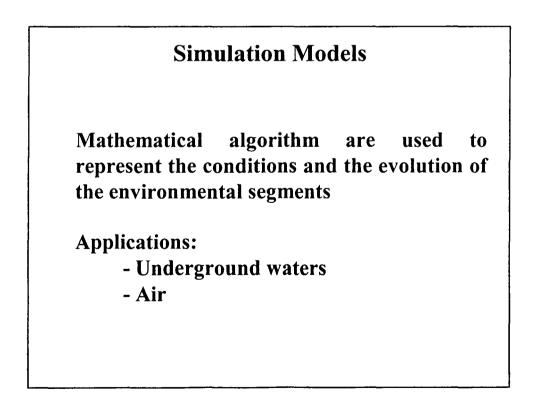


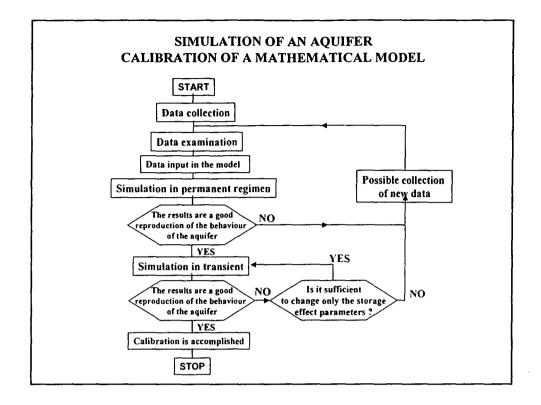
	Atmosphere		
Polluting component	Parameter	Limit value 80 μg/m ³	
Sulphur bioxide as SO2	Mean of the 24 hours average concentrations measured during 1 year		
Nitrogen Oxide NO2	98° percentile of the 1 h average of concentrations measured during 1 year	200 µg/m	
Fluor Compound as F-	Mean of the 24 hours average concentrations measured during 1 month	10 µg/m ³	
Ozone O3	Mean concentration of 1 hour not to be reached more than once per month	200 µg/m	
Carbon dioxide	Mean concentration of 8 hours Mean concentration of 1 hour	10 mg/m 40 mg/m	
Lead	Mean of the 24 hours average concentrations measured during 1 year	10 μg/m ⁻	
Suspended inert dust	Mean of the 24 hours average concentrations measured during 1 year	150 μg/m	

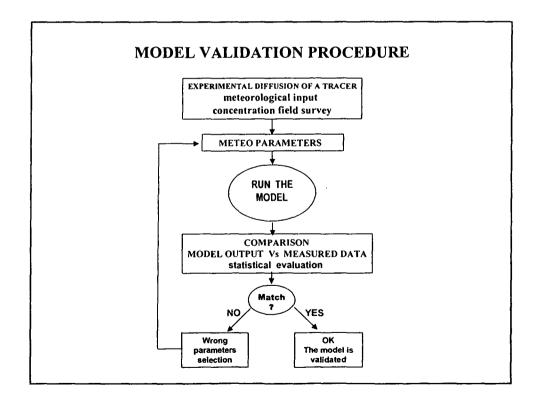
Indicators	Partly indexes					
		5	4	3	2	1
рН		6.5-8.5	5.5 - 6.5 8.5 - 9.5	4.5 - 5.5 9.5 - 10.5	2 - 4.5 10.5 - 12	0 - 2 11 - 14
Conductivity	mS	< 400	400 - 750	750 - 1000	1000-2000	> 2000
COD	nıg/l	< 10	10-15	15-20	20-30	> 30
Chloride	ng/l	< 25	25 - 200	200 - 350	350 - 1200	> 1200
Sulfate	mg/l	< 25	25 - 150	150 - 250	250 - 1000	> 1000
Sodium	mg/l	< 20	20-150	150 - 210	210 - 500	> 500
Nitrate (NQ	mg/l	< 5	5 - 30	30 - 50	50 - 90	> 90
Magnesium	mg/l	12	12÷24	24÷36	36÷53	>53
Manganese	mg/l	< 20	20 - 50	50 - 200	200 - 1000	> 1000
Alluminium	µg/1	< 50	50 - 200	200 - 1000	1000-5000	> 5000
Iron	μg/l	< 50	50 - 200	200 - 1000	1000-2000	> 2000
Copper	μg/1	< 20	20 - 200	200 - 1000	1000-2000	> 2000
Zinc	µg/1	< 100	100 - 400	400 - 1000	1000-2000	> 2000
Cromium (total Cr	l\gц (-	5-10	10-20	20-50	>50
Cadmium	μ <u>e</u> /1	•	1 - 3	3 - 5	5 - 20	> 20
Arsenic	μg/1	•	20 - 50	50 - 100	100 - 500	> 500
Mercury	μ <u>φ</u> /1	-	0.5 - 1	1 - 4	4 - 5	> 5
Lead	µg/1	•	10 - 50	50 - 200	200 - 1000	> 1000
PCB e PCT	μg/1	<0.01	0.01-0.03	0.03-0.05	0.05-0.1	>0.1
Bacterial Load tot (36°C)	N/100ml	< 5	5 - 20	20 - 100	100 - 500	> 500

	Soil				
sc	DIL CHARACTERISTIC	CS	VULNER	ULNERABILTY	
Texture		S-SF-FS-A-AI F-FA-FSA FLA-AS	_	HIGH LOW MIDDLE	
Internal vertical drainage		Quick, very sl Obstructed	DW.	нісн	
	U	Normal Slow	<50	LOW MIDDLE HIGH	
Soil thickness (cn	n)	Very shallow Middle depth Deep		MIDDLE	
		Very deep.	>150	LOW	
Depth of the und	erlying rock and	<1		HIGH	
of the pedogenetic substrate (m)		1-3		MIDDLE	
		>3 <50	LOW HIGH		
Depth of the water	proof horizon (cm)	<50 50 - 70		MIDDLE	
		>70	LOW		
Pol	lyhedral coarse			HIGH	
•••-	assive			HIGH	
	ismatic			HIGH	
	lyhedral fine or e middle			MIDDLE	
	lyhedral very fine coherent			LOW	
110	CONCI CHI	Little	< 0,5 %	HIGH	
Porosity		Common	0,5% - 5%	MIDDLE	
2		Abundant	> 5%	LOW	

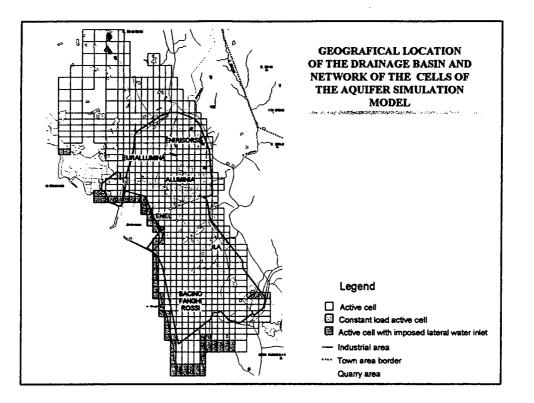


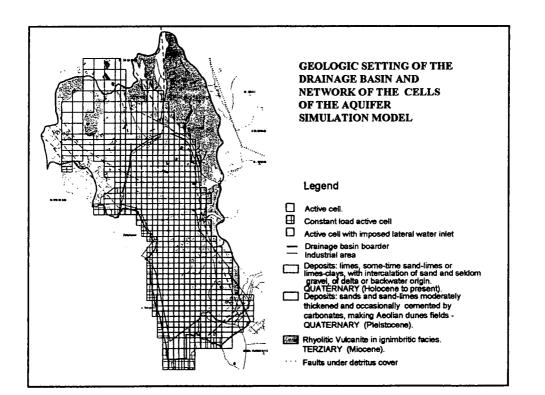




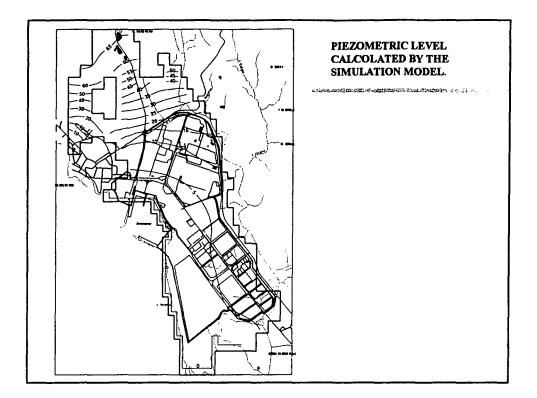


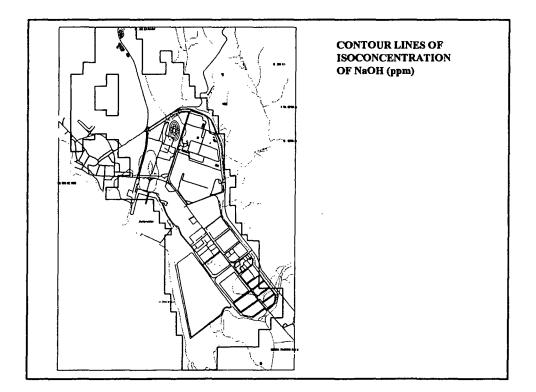
Piano di Disinquinamento per il Risanamento del Territorio del Sulcis-Iglesiente



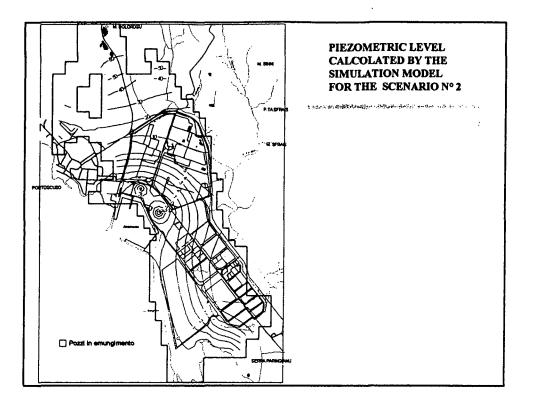


Piano di Disinquinamento per il Risanamento del Territorio del Sulcis-Iglesiente





Piano di Disinquinamento per il Risanamento del Territorio del Sulcis-Iglesiente



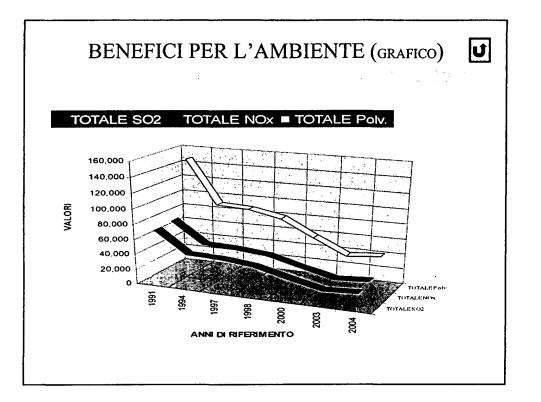
EMISSIONI COM	VOGLIATE	E DI SO2		
	1991	1994	1997	1998
ALCOA	3,000	2,800	3,000	3,00
ENEL **	50,000	21,000	19,410	15,80
ENIRISORSE	2,100	2,800	2,800	2,80
EURALLUMINA	13,500	13,500	14,000	14,00
IGCC		-		<u> </u>
ILA	200	160	120	120
SEAMAG	1,000	1,000	700	-
		·		

Piano di Disinquinamento per il Risanamento del Territorio del Sulcis-Iglesiente

STIMA DELLE EMISS	IONI CONVO	GLIATE	di NOx		r	····	- <u>1</u>
	1991	1994	1997	1998	2000	2003	2004
ALCOA		-	-	-	-		<u> </u>
ENEL **			8,450	11,300	4,000	4,000	4,000
ENIRISORSE	<u> </u>	<u> </u>	<u> </u>	-	-	<u> </u>	<u> </u>
EURALLUMINA		_ <u> </u>	1,700	1,700	1,700	1,700	1,700
IGCC			<u> </u>	· ·		<u> </u>	2,150
ILA		<u>-</u>	<u> </u>	<u>-</u>			<u> .</u>
SEAMAG	-	. <u> </u>	-	-	-	-	
		. @	1000ED	raint			

	1991	1994	1997	1998	2000	2003	2004
ALCOA	600	450	660	660	660	660	660
ENEL **	2,500	1,400	1,400	1,650	1,650	800	800
ENIRISORSE	80	70	70	70	70	70	70
EURALLUMINA	1,100	1,100	400	400	180	180	180
IGCC	-	-	-	-	<u> </u>	<u> </u>	110
ILA .	20	20	10	10	10	10	10
SEAMAG	150	150	70	-	-	-	<u> </u>
VOVA-IS	6.000	3400	2010	270	2.50	3.520	1.299

Piano di Disinquinamento per il Risanamento del Territorio del Sulcis-Iglesiente



GIS and Spatial Data Integration by E. Feoli

GIS and Spatial Data Integration

by Enrico Feoli and Munir Ghribi

Draft for ICS-UNIDO Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas" Iglesias-Cagliari, Italy, 4-8 June 2002

1. Introduction

The coastal zone management requires the integration of often very large volumes of disparate information from numerous sources. The coupling of large information with efficient tools for assessment and evaluation allows broad interactive participation in the planning, assessment, and decision making process; and effective methods of communicating results and findings to a broad audience.

The ultimate objective of a computer based spatial decision support system for coastal zone resources management is, or should be, to improve planning and decision making processes by providing useful and scientifically sound information to the actors involved in these processes, including public officials, planners and scientists, and the general public.

2. The components of SDSS

- Geographic Information System (GIS)

A Geographic Information System (GIS) is designed as a computer tool to efficiently capture, store, update, manipulate, analyse, and display all forms of geographically referenced information (e.g., ESRI 1994). A GIS typically links data from different sets, using geo-referencing, i.e., spatial co-ordinates, as a common key between the data set. The power of a GIS stems from its ability to combine many data sets and display them in a common framework as thematic maps. To obtain thematic maps the GIS may have internal software tools for database and data analysis and image processing, or may be easily interfaced with external software. There are many GIS on the market, commercial and public domain GIS and they are all different each other. They may answer more or less easily and completely to the following typical questions depending on how they have been designed:

- Spatial identification: find locations where certain conditions are satisfied, including thelogical (Boolean) or arithmetic combination of multiple layers (overlay analysis);
- Trends: find the differences within an area over time or along spatial dimensions;
- *Patterns*: find correspondences between distribution of different attributes; calculate similarity between patches, fractal index, indices of fragmentation, and so on;
- *Modelling*: answer "What if" questions.

However, to answer more complex questions one needs to combine (static) geographic data layers with dynamic and spatially distributed models. A progress toward this integration may be achieved by the object-oriented approach (Malczewski, 1999). This can be considered as an alternative or a complement of the layer data base approach common in most of the GIS.

The maps obtained by GIS may be seen not only as cartographic representations of a classification of the landscape at the end of an analytical process, but mainly as data sources for the landscape spatial pattern analysis through the many different indices of the landscape structure such as shape, fragmentation, fractal, diversity, etc. (Ebdon, 1977; Turner, 1989; Milne, 1991; Gardner and O'Neill, 1991; Fabbri, 1991; Baker and Yunming, 1992; Cullinam and Thomas, 1992; Gustafson and Parker, 1992; Olsen, Ramsey, and Winn, 1993). Many GIS have internal data analysis and image processing systems that can calculate different pattern indices. Some GIS such as IDRISI, ILWIS, GRASS (see Malczewski, 1999, for a comparison between different GIS) have the possibility to treat remote sensing data (Image processing) coming from LANDSAT, SPOT, NOAA, etc. however there aren't GIS including data analytical and statistical techniques that are able to classify specific Operational Geographic Units (OGU) according Crovello (1981). The classification may be obtained by applying the clustering algorithms (Orlóci, 1978; Legendre and Legendre, 1983; Goodall and Feoli, 1991) or other multivariate techniques. Feoli and Zuccarello (1996) treat this aspect. GIS can manage different OGUs to obtain maps. In GISs, raster or vector (advantages and disadvantages of using raster or vector-based GISs are described in Burrough (1986). Aronoff (1989), Fabbri (1991) and Johnston (1992)), the OGUs are stored in data bases generally structured according to the relational model (Wildi, 1991). In the databases different characters may describe the OGUs. The probabilistic similarity functions proposed by Goodall (1964, 1966, 1968) that have been suggested and that have been included in the information system developed in ICS/UNIDO (Feoli 1993) would improve the interpretation by introducing the probabilistic aspect in the prediction. They are included in a package described by Goodall, Ganis and Feoli (1991). The functions can deal with 6 categories of characters: 1) binary; 2) qualitative with more than two alternative values; 3) ordinal; 4) only small integer values; 5) quantitative grouped into classes; 6) quantitative not grouped into classes.

- Modelling and Expert Systems

In GIS, the basic concept is one of location, of spatial distribution and relationship; basic elements are spatial objects. In environmental modelling, by contrast, the basic concept is one of state, expressed in terms of numbers, mass, or energy, of interaction and dynamics; the basic elements are ``species'', which may be biological, chemical, and environmental media such as air, water or sediment.

The overlap and relationship is apparent, and thus the integration of these two fields of research, technologies, or sets of methods, that is, their paradigms, is an obvious and promising idea (Fedra 1994,1995).

Simulation and optimization modelling is a powerful tool of analysis, and forecasting. Another example of possible integration is provided by quantitative numerical models and rule-based, qualitative expert systems. *Expert systems* can be used just as any other model to assign a value to an output variable given a set of input variables; they do this, however, by using rules and logical inference rather than numerical algorithms.

In the context of models, expert systems are often used to help configure models (implementing an experienced modellers know-how to support the less experienced user) and estimate parameters. A number of these ``intelligent front end systems" or model advisors have been developed in the environmental domain (Fedra, 1993b, 1992).

A rule-based approach can also be a substitute for a numerical model, in particular, if the processes described are not only in the physical and chemical, but also in the biological

and socio-economic domain. An example could be environmental impact assessment based on a checklist of problems, which can be understood as a diagnostic or classification task. A qualitative label is assigned to potential problems, based on the available data on environment and planned action, and a set of generic rules assessing and grading the likely consequences. An example of such a rule-based system for impact assessment is described in Fedra et al., 1991.

And finally, a model can be integrated into the inference chain of an expert system (Fedra, 1992). Recent examples of environmental expert systems are given in Hushon (1990) and Wright et al. (1993).

The flexibility to use, alternatively or conjunctively, both symbolic and numerical methods in one and the same application allows the system to be responsive to the information at hand, and the user's requirements and constraints. This combination and possible substitution of methods of analysis, and the integration of data bases, geographical information systems, and hypertext, allows to efficiently exploit whatever information, data and expertise is available in a given problem situation.

The approach is based on a model of human problem solving that recursively refines and redefines a problem as more information becomes available or certain alternatives are excluded at a screening level. Learning, i.e., adaptive response to the problem situation and the information available, and the ability to modify function and behaviour, as more information becomes available, is a characteristic of intelligent systems.

- Simulation and Optimization

Another example of the integration of different methods is the coupling of simulation and optimization: optimization usually requires a (often-gross) simplification of the problem representation to become tractable. Simulation models, on the other hand, while capable of representing almost arbitrary levels of detail and complexity, are rarely capable of solving inverse problems, i.e., determining the necessary set of inputs or controls to reach a desired outcome.

One can, however, combine the approaches in that a simplified model (e.g., steady state and spatially aggregated) is used as the basis for optimization; the result of the optimization is then used as the basis for a more detailed, e.g., dynamic and spatially distributed simulation model, that also keeps track of the criteria, objectives and constraints used for the optimization, but with a higher degree of spatial and temporal resolution, and possibly a more refined process description. If, in the simulation run, constraints are violated or objectives not met, the corresponding values can be tightened or relaxed in the optimization to obtain a new solution which again is subjected to more detailed examination with the simulation model.

Modelling of marine systems has a considerable tradition in both physical and biological oceanography. For summary treatments, more recent model comparisons, and some classics in the field see, for example, Riley et al., 1949; Nihoul, 1975; Goldberg et al., 1977; Kremer and Nixon, 1978; Falconer et al., 1989; Fransz et al., 1991.

However, most of these approaches are designed to improve scientific understanding of physical and biological processes, with the notable exception of some of the fisheries models that have obvious management implications (e.g., Andersen and Ursin, 1977). Other management oriented and model based studies of marine systems include the RAND Corporation classic on the Oosterschelde (Bigelow et al., 1977) or the recent study of the North Sea MANS, Management Analysis North Sea (Rijkswaterstaat, 1988, 1989; and Klomp, 1990). Other important recent models for coastal zone management are ECOPATH, ECOSIM and ECOSPACE (Pauly et al., 2000) that are offering a set of tools for evaluating ecosystem impact of fisheries based on modelling the mass balance fluexes in the fish food-webs both in homogeneous and heterogeneous space conditions.

- Multi-criteria Decision Analysis

Criterion is some basis for a decision that can be measured and evaluated. It is the evidence upon which a decision is based. Criteria can be of two kinds: factors and constraints.

Factor is a criterion that enhances or detracts from the suitability of a specific alternative for the activity under consideration. It is therefore measured on a continuous scale.

Constraint serves to limit the alternatives under consideration. In many cases constraints will be expressed in the form of a Boolean (logical) map: areas excluded from consideration being coded with a 0 and those open for consideration being coded with a 1.

Decision Rule is the procedure by which criteria are combined to arrive to a particular evaluation and by which evaluations are compared and acted upon.

A criterion is therefore a measure on a continuous scale of the effect on the activity on a given indicator commonly named impact (Patrono, 1997).

A decision rule is the procedure by which criteria are combined to arrive at particular evaluation and by which evaluations are compared and acted upon. Decision rules are structured in the context of a specific objective for example to determine area suitable for a given activity. To meet a specific objective it is frequently the case that several criteria will need to be evaluated (Multi-Criteria Evaluation). It also happens that decisions, which satisfy several objectives, are required (Multi-Decision Evaluation).

- Multi-Criteria Evaluation

To meet a specific objective, it is frequently the case that several criteria will need to be evaluated. Such procedures are called Multi-Criteria Evaluations (MCE) (Eastman ET al., 1995). Two of the most common procedures for multi-criteria evaluation are weighted linear combination and concordance-discordance analysis (Carver, 1991). Weighted linear combination is very straightforward in a raster GIS. Indeed, it is the derivation of the weights, within the context of the decision objective that provides the major challenge. The weights are developed by providing a series of PAIRWISE comparisons of the relative importance of factors to the suitability of pixels for the activity being evaluated (Eastman, 1997).

In a Multi-Criteria Evaluation, an attempt is made to combine a set of criteria to achieve a single composite basis for a decision according to a specific objective (Eastman et al., 1995). During the last two decades, further support has emerged for the view that a decision is a multidimensional concept (Nijkamp et al., 1990; Paruccini, 1994). Decisions about the allocation of land typically involve the evaluation of multiple criteria according to several, often conflicting-objectives (Eastman et al., 1995). Making-decisions about the allocation of land is one of the most fundamental activities of resource development (FAO, 1976). With the development of GIS, we now have the opportunity, for a more explicitly reasoned process of land-use evaluation. However, despite the wide range of analytical tools these systems provide, they are typically weak in the provision of decision support procedures (Honea et al, 1991).

The advantage of MCE is that it provides a flexible way of dealing with qualitative multi-dimensional environmental effects of decisions (Munda, 1995). Even in the absence of monetary information, the point of departure of any multi-criteria analysis is the generation of a discrete set of alternatives, the formulation of a set of criteria, and the evaluation of an impact of each criterion for every alternative. Janssen (1992) and Munda (1995) provide a comprehensive review of multi-criteria decision support for environmental management.

Although a variety of techniques exist for the development of weights, one of the most

promising would appear to be that of PAIRWISE comparisons developed by Saaty (1977) in the context of a decision making process known as the Analytical Hierarchy Process (AHP). In the PAIRWISE comparison method the decision-maker is asked to select the most important of each possible pair effects (Janssen & Van Herwijen, 1992). Subsequently comparison must be established in qualitative terms to what extend one effect is more important than the other one (Saaty, 1980) to express the differences of importance.

- Multi-Objective Evaluation

While many decisions we make are prompted by a single objective, it also happens that we need to make decisions that satisfy several objectives. A Multi-Objective problem is encountered whenever we have two candidate sets (i.e., sets of entries) that share members. These objectives may be complementary or conflicting in nature (Carver, 1991:332).

a) Complementary Objectives

With complementary or non-conflicting objectives, land areas may satisfy more than one objective, i.e., an individual can belong to more than one decision set. Desirable areas will thus be those, which serve these objectives together in some specified manner. For example, we might wish to allocate a certain amount of land for combined recreation and wildlife preservation uses. Optimal areas would thus be those that satisfy both of these objectives to the maximum degree possible.

b) Conflicting Objectives

With conflicting objectives, objectives compete for the available land since it can be used for one or the other, but not both. For example, we may need to resolve the problem of allocating land for timber harvesting and wildlife preservation. Clearly the two cannot coexist. Exactly how they compete, and on what basis one will win out over the other will depend upon the nature of the decision rule that is developed.

In case of complementary objectives, multi-objective decisions can often be solved through a hierarchical extension of the multi-criteria evaluation process. For example, we might assign a weight to each of the objectives and use these along with the suitability maps developed for each to combine them into a single suitability map indicating the degree to which areas meet all of the objectives considered (see Voogd, 1983). However, with conflicting objectives the procedure is more involved.

With conflicting objectives, it is sometimes possible to rank order to objectives and reach a *prioritised* solution (Rosenthal, 1985). In these cases, the needs of higher ranked objectives are satisfied before those of lower ranked objectives are dealt with. However, this is often not possible, and the most common solution to conflicting objectives is the development of a *compromise* solution. Undoubtedly the most commonly employed techniques for resolving conflicting objectives are those involving optimization of a choice function such as mathematical programming (Fiering, 1986) or goal programming (Ignizio, 1985). In both, the concern is to develop an allocation of the land that maximises or minimises an objective function subject to a series of constraints.

- The User Interface

Analysis in GIS is based on the map paradigm; maps in inherently static objects, and the basic level of analysis is (Boolean) map overlays. Complex relationships, and dynamic

processes, may require another layer of tools, namely simulation models, to be integrated with a GIS.

The integration of GIS and environmental models can come in many forms. In the simplest case, two separate systems, the GIS and the model, just exchange files: The model obtains some of its input data from the GIS, and produces some of its output in a format that allows import and further processing and display with the GIS. This seems to be a rather common approach, since it requires little if any software modifications. Only the file formats and the corresponding input and output routines, usually of the model, have to be adapted. Depending on the implementation however, a solution based on files shared between two separate applications, usually with a different user interface. is cumbersome and possibly error prone if it involves a significant amount of manual tasks. Deeper integration provides a common interface and transparent file or information sharing and transfer between the respective components. One possible way is the use of higher-level application language or application generators increasingly common as built-in features of commercial GIS packages as the basis of numerous integrated applications. Application generators and modelling capabilities with commercial GIS also offer the possibility of tight integration within the limits of the respective package options.

An alternative is to use an open GIS tool kit that uses a standardised interface, such as GRASS (< biblio >). Modules of the overall GIS system (which really is a set of tools with a standardised pipeline-type flexible coupling) can be included in modelling applications. The X Windows system and a number of interface building toolkits make this a rather efficient integration strategy. Any integration at this level, however, requires a sufficiently open GIS architecture, that provides the interface and linkages necessary for tight coupling.

Using such predefined tools and components, while efficient for fast prototyping, can also be restrictive: investments in toolkits always carry the temptation to reformulate problems in terms of the available tools rather than the other way round.

Another alternative is the use of do-it-yourself tool kits that provide both customised GIS functionality as well as interface components for simulation models. A relatively recent example of integration that draws together GIS, models, spreadsheet, and expert systems in a programmable system is RAISON (Lam and Swayne, 1991).

For having an efficient user interface we need:

An interactive, menu-driven user interface that guides the user with prompt and explain messages through the application. No command language or special format of interaction is necessary, the computer assists the user in its proper use; help and explain functions can be based on hypertext and possibly include multi-media methods to add video and audio technology to provide tutorial and background information; in other words, they must be easy to use even for the non-specialist.

Dynamic colour graphics for the model output and a symbolic representation of major problem components that allow easy and immediate understanding of basic patterns and relationships. Rather than emphasising the numerical results, symbolic representations and the visualisation of complex patterns support an intuitive understanding of complex systems behaviour; the goal is to translate a model's state variables and outputs into the information requirements of the decision making process; the coupling to one or several data bases, including geographical information systems, and distributed or remote sources of information in local or wide area networks, that provide necessary input information to the models and the user. The user's choice or definition of a specific scenario can be expressed in an aggregated and symbolic, problem-oriented manner without concern for the technical details of the computer implementation; data on chemical properties of the substances of concern, used in pollution fate and transport models, are one obvious example;

Embedded Artificial Intelligence (AI) components such as specific knowledge bases allow user specifications in allowable ranges to be checked and constrained, and ensure the consistency of an interactively defined scenario; as a special case this could also include QSAR methods that can be implemented in a combination of rule-based methods and algorithmic methods;

Interaction is a central feature of any effective man--machine system: a real-time dialogue, including explanation, allows the user to define and explore a problem incrementally in response to immediate answers from the system. Fast and powerful systems with modern processor technology can offer the possibility to simulate dynamic processes with animated output, and they can provide a high degree of responsiveness that is essential to maintain a successful dialogue and direct control over the software.

Visualisation provides the band-width necessary to communicate and understand large amounts of highly structured information, and permits the development of an intuitive understanding of processes and interdependencies, of spatial and temporal patterns, and complex systems in general. Also, many of the problem components in a real-world planning or management situation, such as risk or reliability, are rather abstract: a graphical representation of such concepts makes them tangible objects that can literally be manipulated and understood intuitively.

Intelligence requires software to be knowledgeable not only about its own possibilities and constraints, but also about the application domain and about the user, i.e., the context of its use. Defaults and predefined options in a menu system, sensitivity to context and history of use, built-in estimation methods, learning, or alternative ways of problem specification depending on the user can all be achieved by the integration of expert systems technology in the user interface and in the system itself.

Customisation is based on the direct involvement of the end-user, and the consideration of institutional context and the specifics of the problem domain in systems design and development. It is the users view of the problem and their experience in many aspects of the management and decision making process that the system is designed to support. This then must be central to a system's implementation to provide the basis for user acceptance and efficient use.

GIS and Spatial Data Integration

by

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Remote Sensing Techniques by A. Altobelli

Basics of Remote Sensing from Satellite by A. Altobelli

What is remote sensing?

" Remote Sensing" refers to the science of measuring properties of the Earth' s surface (land and sea), and atmosphere from a distance, using sensors onboard airplanes or satellites. The information needs a physical carrier to travel from the object/phenomena to the sensor through a medium. <u>Electromagnetic radiation</u> is normally used as an information carrier in remote sensing.

The electromagnetic spectrum

The entire array of electromagnetic waves comprises the electromagnetic (EM) spectrum. The EM spectrum has been arbitrarily divided into regions or intervals to which descriptive names have been applied. At the very energetic (high frequency; short wavelength) end are gamma rays and x-rays. Radiation in the ultraviolet region extends from about 1 nanometer (nm) to about 0.40 micrometers. The <u>visible</u> region occupies the range between 0.4 and 0.7 μ m, or its equivalents of 400 to 700 nm. The infrared (IR) region, spans between 0.7 and 100 μ m, more than 100 times as wide as the visible portion!

The infrared region can be divided into two categories based on their radiation properties the reflected IR, and the emitted or thermal IR.

Radiation in the reflected IR region is used for remote sensing purposes in ways very similar to radiation in the visible portion. The reflected IR covers wavelengths from approximately 0.7 μ m to 3.0 μ m. There are two different categories of IR waves: <u>near-IR</u> (from 0.7 to 1.3 μ m) and <u>mid-IR</u> (from 1.3 to 3 μ m). The thermal IR region is quite different than the visible and reflected IR portions, as this energy is essentially the radiation that is emitted from the Earth's surface in the form of heat. The thermal IR covers wavelengths beyond 3.0.

Longer wavelength intervals are measured in units ranging from millimetres (mm) through meters (m). The microwave region spreads across 1 mm to 1 m; this includes all of the intervals used by man-made radar systems.

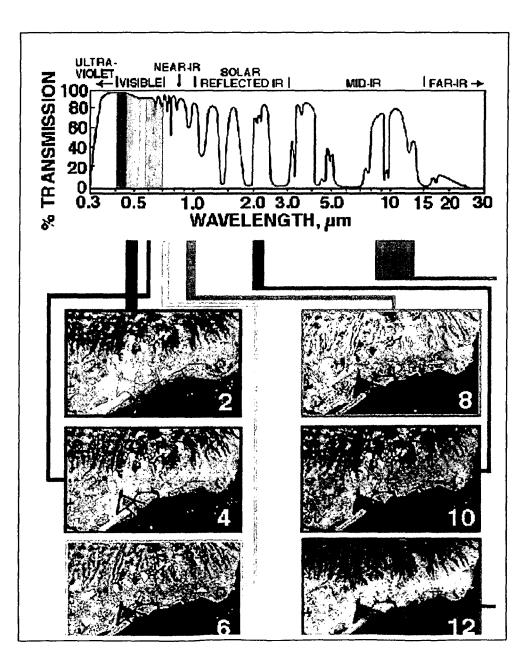
The lowest frequency (longest wavelength) region beyond 1 m is associated with radio waves.

Effects of Atmosphere

Atmospheric constituents absorption and scattering of radiation. These affects degrade the quality of images. Some of the atmospheric effects can be corrected before the images are subjected to further analysis and interpretation.

A consequence of atmospheric absorption is that certain wavelength bands in the electromagnetic spectrum are strongly absorbed and effectively blocked by the atmosphere. The wavelength regions in the electromagnetic spectrum usable for remote sensing are determined by their ability to penetrate atmosphere.

These regions are known as the <u>atmospheric transmission windows</u>. Remote sensing systems are often designed to operate within one or more of the atmospheric windows. These windows exist in the microwave region, some wavelength bands in the infrared, the entire visible region and part of the near ultraviolet regions.



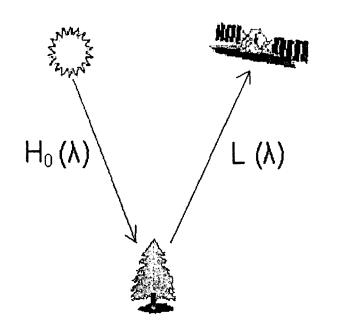
These images are registered by an electronic camera at the same time in different parts of the spectrum.

- Notice that the vertical axis (called % transmission) shows where our atmosphere allows a large quantity of the Sun's rays to reach the ground. It is easy to see why our eyes work in the visible part of the spectrum, because it is here that almost 100% of the Sun's energy reaches the surface. In contrast, due to water vapour in the atmosphere, almost no energy (%0 transmission) reaches the ground at 1.4 and 1.9 microns
- The images are all obtained in between 0.4 and 12.0 microns as a set of 12 individual black and white images. All 12 images are obtained at exactly the same time, and each is a computer image where we can assign an 8-bit binary number between 0 (black) and 255 (white). In this diagram, we show six of these images (or "bands"), along with the wavelength of the band. We show bands 2, 4, 6, 8, 10, and 12 just to cover the whole range, and to show that not all bands have the same wavelength range (the width of the colour bar beneath the spectrum). See how Band 12 is very wide (about 4 microns) while band 4 is very narrow (about 0.05 microns).

The various dimensions of remote sensing Three typical measurement techniques: Reflected (Passive) in visible and near IR

 $R(\lambda) = L(\lambda) / H_0(\lambda)$

(reflected) / (sun irradiation)

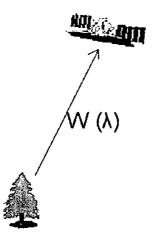


Emission (Passive) in thermal IR

Measured as a function of blackbody temperature (T)

 $W = \varepsilon \sigma T^4$

Where W is the exitance from the surface of a material, ϵ is the emissivity factor, σ is the Stefan-Boltzmann constant and T the absolute temperature (K) of the emitting material.



Scattering/dieletric (active) in microwaves

Measured returned signal (\underline{W}_R) s attenuated by distance (D) and σ_0 (roughness, dieletric) $W_R = A \times (\sigma_0 / D^4) \times W_E$ Where W_E is the emitted signal.

Radar D

Optical - infrared remote sensing

In optical remote sensing, optical sensors detect solar radiation in the visible and infrared wavelength region.

When electromagnetic energy from the sun hits the earth's surface three fundamental energy interactions are possible. Various fractions of energy are:

Reflected

Absorbed

Transmitted

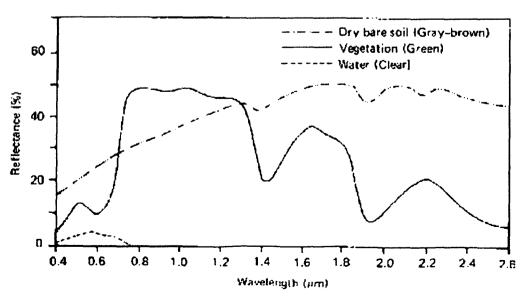
The proportions of energy reflected, absorbed, and transmitted will vary for different earth features. These differences permit us to distinguish different features of an image.

The interpretation of optical images requires the knowledge of the <u>spectral curve (spectral signature)</u> of the various materials (natural or man-made) covering the surface of the earth.

Spectral reflectance curves

The figure below shows typical spectral reflectance curves for three basic types of earth features: healthy green vegetation, dry bare soil, and clear lake water. These curves indicate how much incident energy would be reflected from the surface, and subsequently recorded by a remote sensing instrument. At a given wavelength, the higher the reflectance, the brighter the object appears in an image.

Note that vegetation reflects much more energy in the near-infrared (0.8 to 1.4 microns) than it does in visible light (0.4 to 0.7 microns). The amount of energy that vegetation reflects is related to the internal structure of the plant, and the amount of moisture in the plant. Another feature to notice is that clear water reflects visible light only, so it will appear dark in infrared images.

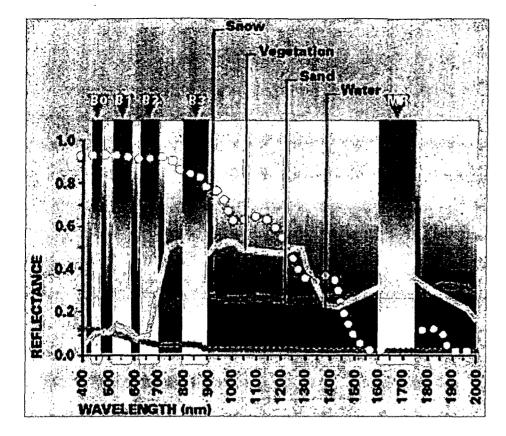


The <u>red edge</u> is the reflectance inflection point observed in the spectrum of green plants at the transition between the visible and near-infrared wavelengths. It is the long-wavelength edge of the chlorophyll absorption feature and has been used to describe the variation in leaf and canopy chlorophyll concentration. This boundary point in reflectance typically occurs between 690 and 740 nm in fresh leaves and is determined by the interaction between chlorophyll absorption of red light and the internal scattering process in the leaf. Increased chlorophyll concentration causes a deepening and broadening of the chlorophyll absorption feature, while stressed causing chlorosis usually increase reflectance and narrow the absorption feature, making the inflection point appear to move to longer or shorter wavelength, respectively. The relationship has been utilized to detect symptoms of mineral deficiency and air pollution damage.

Satellites characteristics

[Bits per
	Spectral bands	Nadir IFOV	Swath Width	Revisit at equator	pixels
					16 MS, 8
NOAA/AVHRR	Red-IR-TIR	1 Km	2700 Km	2 passes/day	NDVI
		30m MS			
Landsat TM	VIS-IR-TIR	120m TIR	185 Km	16 days	8
		30m MS			
	VIS-IR-TIR	120m TIR			
Landsat ETM+	PAN	15m PAN	185 Km	16 days	8
		20m MS			
SPOT	VIS-IR-PAN	10m PAN	60 Km	26 days	8
ERS1/2	C-band	30 m	100 Km	35 days	
		4m MS			
Ikonos	VIS-IR-PAN	1m PAN	11 Km		11
		2,44 m MS			
QuickBird	VIS-IR-PAN	0,61 m PAN	16.5 Km		11

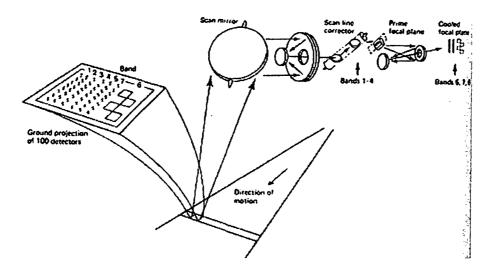
IFOV Instantaneous Field of View, Ms multispectral, VIS Visible, IR reflected infrared, TIR thermal infrared, PAN panchromatic.



"Xi" multispectral mode (Spot 4). The bands used are band B1 covering 0.50 to 0.59 μ m (green), band B2 covering 0.61 to 0.68 μ m (red), band B3 covering 0.79 to 0.89 μ m (near infrared). plus a Short Wave Infrared band B4, covering 1.58 to 1.75 μ m.

The Landsat program

The Landsat program consists of a series of optical-infrared remote sensing satellites for land observation. The program was first started by The National Aeronautics and Space Administration (NASA), in 1972. The first satellite in the series, Landsat-1 (initially named as the Earth Resource Technology Satellite ERTS-1) was launched on 23 July 1972. Landsat-2 was launched on 22 January 1975 and three additional Landsat satellites were launched in 1978, 1982, and 1984 (Landsat-3, 4, and 5 respectively). Landsat-6 was launched on October 1993 but the satellite failed to obtain orbit. A new satellite LANDSAT-7 was launched on 15 April 1999. Currently, only Landsat-5 and 7 are operational.



The Landsat satellites operate in near-polar, sun synchronous orbits. The altitude of the orbit of Landsat 5 (with Thematic Mapper) is 705 Km, Each orbit takes approximately 99 minutes, with just over 14.5 orbits being completed in one day. The orbit being sun synchronous implies that the satellite when travelling from north to South, passes the equator always at the same local time, 9:45 am for Landsat 5.

Between two successive orbits the earth rotates approximately $1/14.5 \times 360^\circ = 24.8^\circ$, which corresponds to 2754 Km at the equator. With a trace width of only 185 Km each day only a small portion of the earth surface can be scanned. The orbit is chosen in such a way that every 16 days the whole earth surface is covered.

To keep an orbit sun synchronous it has to be near polar instead of exactly polar.

The oscillation mirror acquires data of 16 scan lines (474 m) within each sweep in both directions, since there are 16 detectors for each of the six bands (4 for the thermal band). At receiving stations, TM imagery is resample to 30×30 m ground resolution (120 for the thermal band).

Band	Wavelength (µm)	Spectral name	Principal Applications
1	0.45 - 0.52	Blue	Designed for water body penetration, Making it useful for coastal water mapping. Also useful for soil/vegetation discrimination and forest type mapping.
2	0.52 – 0.60	Green	Designed to measure green reflectance peak of vegetation for vegetation discrimination and vigour assessment.
3	0.63 – 0.69	Read	Designed to sense in a chlorophyll absorption region aiding in plant species differentiation.
4	0.76 – 0.90	Near infrared	Useful for determining vegetation types, vigour and biomass content, for delineating water bodies and for soil moisture discrimination.
5	1.55 – 1.75	Mid infrared	Indicative of vegetation moisture content and soil moisture. Also useful to differentiate snow from clouds.
6	10.5 – 12.5	Thermal Infrared	Useful in vegetation stress analysis, soil moisture discrimination, and thermal mapping applications.
	2.08 – 2.35		Useful for discrimination of mineral and rock types. Also sensitive to vegetation moisture content.

Panchromatic and multispectral images

A <u>panchromatic image</u> consists of only one band. It is usually displayed as a <u>grey scale</u> <u>image</u>, i.e. the displayed brightness of a particular pixel is proportional to the pixel digital number which is related to the intensity of solar radiation reflected by the targets in the pixel and detected by the detector. Thus, a panchromatic image may be similarly interpreted as a black-and-white aerial photograph of the area, though at a lower resolution.

A <u>multispectral image</u> consists of several bands of data. For visual display, each band of the image may be displayed one band at a time as a grey scale image, or in combination of three bands at a time as a colour composite image.

In displaying a colour <u>composite image</u>, three primary colours (red, green and blue) are used. When these three colours are combined in various proportions, they produce different colours in the visible spectrum. Associating each spectral band (not necessarily a visible band) to a separate primary colour results in a colour composite image. When displaying a <u>natural colour composite</u> image, the spectral bands (some of which may not be in the visible region) are combined in such a way that the appearance of the displayed image resembles a visible colour photograph, i.e. vegetation in green, water in blue, soil in brown or grey, etc. Many people refer to this composite as a "<u>true colour</u>" composite. However, this term may be misleading since in many instances the colours are only simulated to look similar to the "true" colours of the targets. For example, the bands 3 (red band), 2 (green band) and 1 (blue band) of a Landsat TM image can be assigned respectively to the R, G, and B colours for display. In this way, the colours of the resulting colour composite image resembles closely what would be observed by the human eye.

The display colour assignment for any band of a multispectral image can be done in an entirely arbitrary manner. In this case, the colour of a target in the displayed image does not have any resemblance to its actual colour. The resulting product is known as a <u>false</u> <u>colour composite</u> image. There are many possible schemes of producing false colour composite images. However, some scheme may be more suitable for detecting certain objects in the image. R = (NIR band), G = (red band), B = (green band).

This false colour composite scheme allows vegetation to be detected readily in the image. In this type of false colour composite images, vegetation appears in different shades of red depending on the types and conditions of the vegetation, since it has a high reflectance in the NIR band. Clear water appears dark-bluish (higher green band reflectance), while turbid water appears cyan (higher red reflectance due to sediments) compared to clear water. Bare soil, roads and buildings may appear in various shades of blue, yellow or grey, depending on their composition.

In the RGB colours combination 7 (mid IR band), 5 (mid IR band) and 2 (green band) of a Landsat TM the inclusion of the mid-infrared bands provide additional information on rock morphology.

Vegetation indices

Different bands of a multispectral image may be combined to accentuate the vegetated areas. One such combination is the ratio of the near-infrared band to the red band. This ratio is known as the Ratio Vegetation Index (RVI)

RVI = NIR/Red

Since vegetation has high NIR reflectance but low red reflectance, vegetated areas will have higher RVI values compared to non-vegetated aeras. Another commonly used vegetation index is the <u>Normalised Difference Vegetation Index (NDVI)</u> computed by

NDVI = (NIR - Red)/(NIR + Red)

In the NDVI map shown above, the bright areas are vegetated while the nonvegetated areas (buildings, clearings, river, sea) are generally dark. Note that the trees lining the roads are clearly visible as grey linear features against the dark background. The NDVI band may also be combined with other bands of the multispectral image to form a colour composite image which helps to discriminate different types of vegetation, for example:

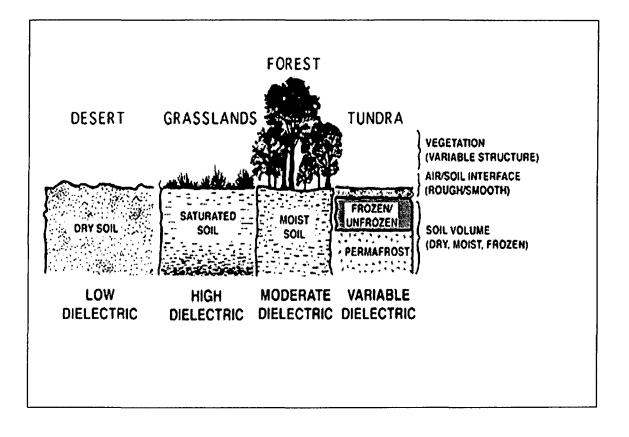
R = (NIR band), G = (NDVI band), B (green band)

Microwave Remote Sensing

The active sensor emits pulses of microwave radiation to illuminate their targets. Images of the earth surface are formatted by measuring the microwave energy scattered by the ground or sea back to the sensors. The images can thus be acquired day and night. Microwaves have an additional advantage as they can penetrate clouds. Images can be acquired even when there are clouds covering the earth surface.

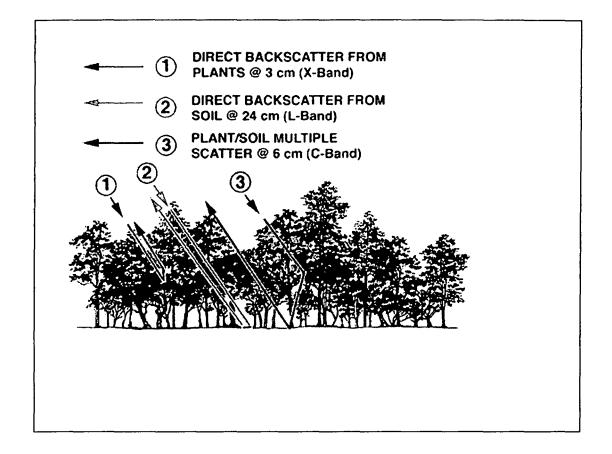
Real researchers who use radar data have to worry about many different aspects of the surface. Different vegetation types (e.g., desert, grasslands, forests or frozen tundra) will all have different backscatter properties. In addition, the basic reflectivity of the soil, called the "dielectric constant" will change depending on the amount of water that the soil contains. Dry soil has a low dielectric constant, so that little radar energy will be reflected. Saturated soil will have the opposite effect, and will be a strong reflector. Moist and partially frozen soils will have intermediate values.

Thus, if we keep the radar incidence angle constant, we can use radar to study large differences in moisture content of the soil. This is often done in deserts (looking for subsurface water) or in tropical rain forests (looking for flooded rivers beneath the tree canopies).



Radar signals can be generated at several different wavelengths, which is useful because energy has an ability to travel through vegetation or soil to different amounts that are controlled by the dielectric constant of the material. As this diagram shows, short wavelength radar (3 cm) will be reflected from the tops of trees. Long wavelength radar (24 cm) data will normally go right down to the ground and be reflected off of the surface. Intermediate wavelength radars (say, 6 cm) will sometimes experience multiple scattering events within the canopy.

If we had a set of different wavelength radar images over a forest, it should be possible to use this changing penetration capability to study the structure of the trees and the total amount of material ("biomass") in the forest



Projection and datum

Maps are flat, but the surfaces they represent are curved. Transforming the threedimensional space onto a two-dimensional map is called "projection". There are many examples that can be used to describe the projection process. One of the most common describes the result of trying to flatten an orange peel. Take an orange and remove the peel, as much as possible, in one piece. When you try to flatten the peel, the edges crack, pieces break off, and parts of the peel remain raised and distorted. Projections make it possible to create maps of areas of the earth with as little distortion as possible. The projection process affects four properties: area, shape, distance, and direction. There is no projection that maintains the integrity of all four properties at the same time. A particular projection should be chosen based on the importance to your project of one of the affected properties. For example, if you want to analyse land use with respect to the percentage of area used for different purposes (i.e. agricultural versus residential) the reliability of your results rest with an accurate estimate of area. Therefore, you would choose an "equal area" projection such as Lambert Azimuthal Equal Area. An "equal area" projection is one that reports accurate area measurements while incurring some distortion of the remaining three properties-shape, distance, and direction.

If you want to optimise for shape, you would choose a "conformal projection" such as Lambert Conformal Conic. A conformal projection maintains shapes such as rectangles (buildings) at the expense of area, distance, and direction. The Universal Transverse Mercator (UTM) projection tries to maintain a happy medium among all four properties. A Geodetic datum is a reference point that describes the position, orientation and scale relationships of a reference ellipsoid to the Earth. The North American Datum of 1927 (NAD 27) uses Clarke spheroid of 1866 to represent the shape of the earth.

Lambert Azimuthal Equal-Area Projection

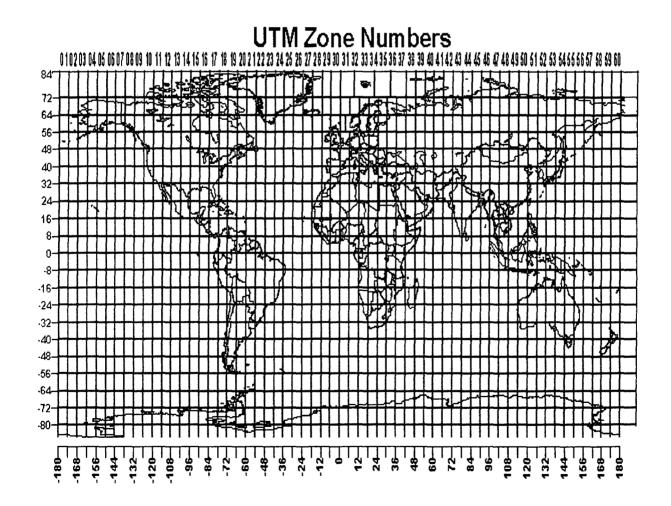




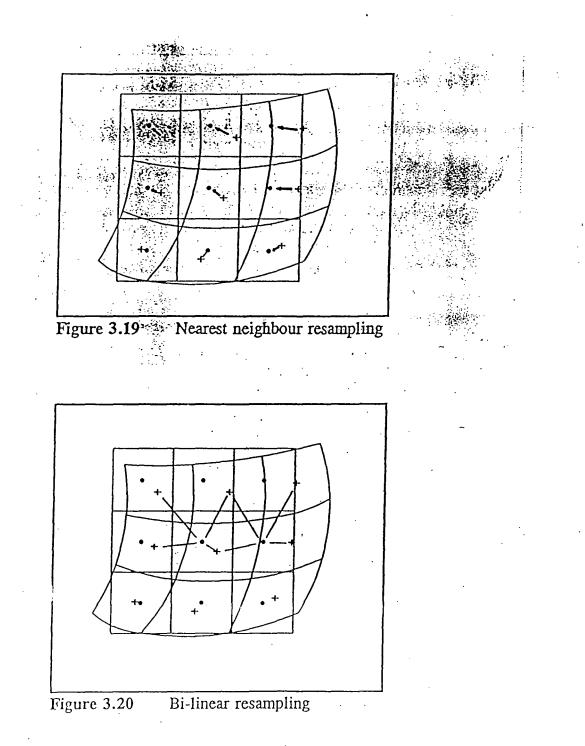
Geographic Coordinate System

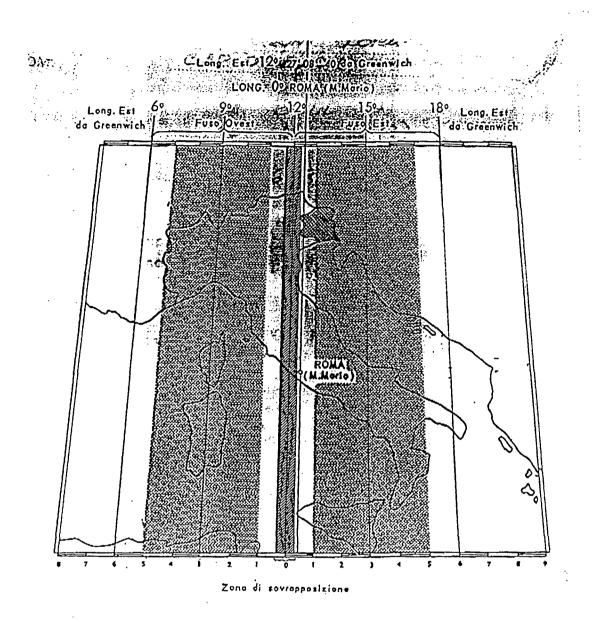


Universal Transverse Mercator (UTM) System



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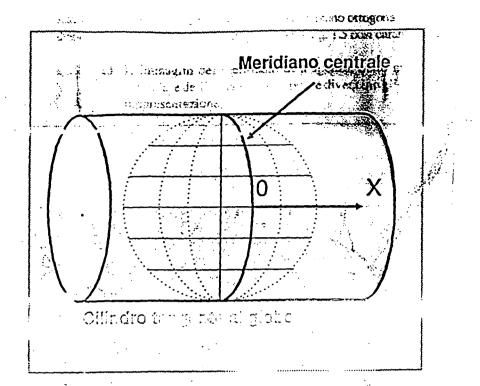


Figura 1.4 - Proiezione trasversa di Mercatore: il cilindro è tangente al globo lungo un meridiano acuo una mina esterio di .

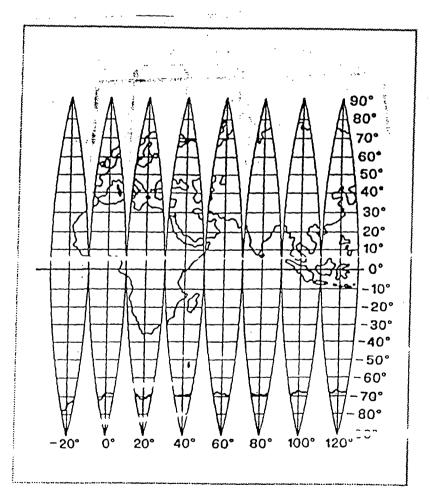
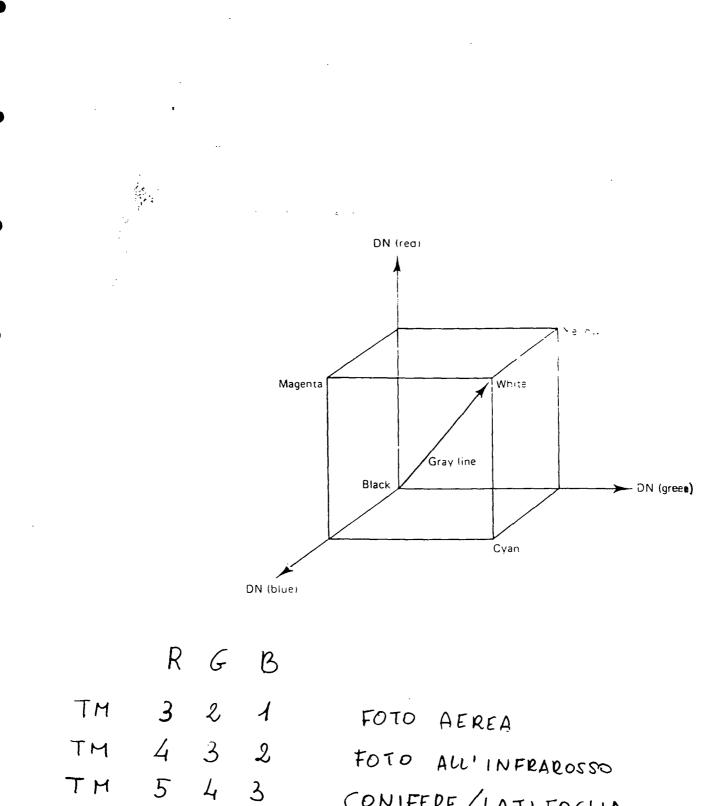


Figura 1.1.6 - Fusi nella carta di Gaussi i fusi rappresentati hanno ampiezza di 90 144 G. Bezoari e altri, op. cit.).

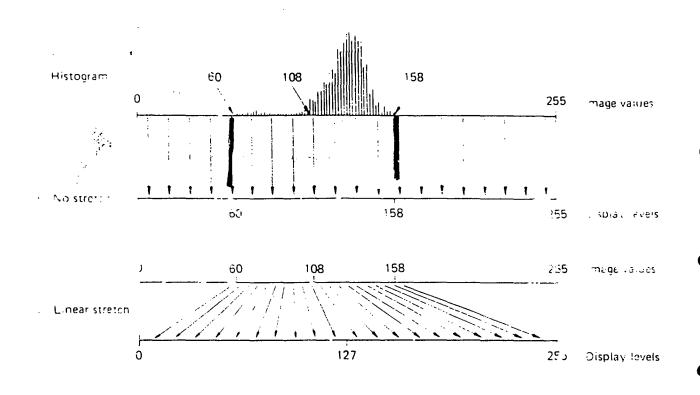
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CONTRAST STRECHING



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- HISTOGRAM EQUALIZATION

ORIGINAL BANDS FEATURES SELECTION (FEATURE SPACE) CLASSIFICATION (1) UNSUPERVISED 2) SUPERVISED (1) CLUSTERING ALGORITHMS of SINGLE PASS ALGORITHM ITERATIVE TYPE - ISODATA HAXIMUM LIKELIHJOD JMultiSpec ECHO SPECTRAL-SPATIAL JMultiSpec 2

NEURAL NETWORK

MultiSpec: free Max and PC image processing Software PORDOE University, IN RHp: //olynamis.ecn.purolue.eolus/~biehl/HultiSpec/finder.html Touisi 32 Release 2: Geographic Analysis and Image. Processing Software GLARK deb. - Clark Univerity, HA Acaslemic single reser license Non-profit single reser license Unlimitest strustent lab. Kit 600 800 \$ 4000 http://www.clarklabs.org/ ERASS (Geographic Resources Amalysis Support System) Open source, free 615 (direcs) http:// gran.itc.it

Data Processing and Analysis

Kauth-Thomas transformations were applied to strengthen the correlation between spectral response and biophysical reality. These transformations resulted in six new features, three of which were considered directly relevant to this research: brightness, greenness, and wetness, related to soil brightness, green biomass reflectance, and moisture condition, respectively. The model coefficients for brightness, greenness, and wetness were obtained from Crist and Cicone (1984) and $B_1, B_2, B_3, B_4, S_5,$ and B_7 are the reflectance values of TM bands 1, 2, 3, 4, 5, and 7, respectively.

$$Brightness = 0.33183 \times B_1 + 0.33121 \times B_2 + 0.55177 \times B_3 + 0.42514 \times B_4 + 0.48087 \times B_5 + 0.25252 \times B_7$$
(3)

$$Greatiness = -0.24717 \times B_1 - 0.162663 \times B_2 - 0.40639 \times B_3$$

$$+ 0.85468 \times B_4 + 0.05493 \times B_5 - 0.11749 \times B_7$$
(4)
$$W_{\text{SUMASS}} = 0.13929 \times B_5 + 0.22490 \times B_5 + 0.40359 \times B_5$$

$$H: Mess = 0.13929 \times B_1 + 0.22490 \times B_2 + 0.40359 \times B_3 + 0.25178 \times B_4 - 0.70133 \times B_5 - 0.45732 \times B_7$$
(5)

of the TM6 must be converted into thermal radiance L_{TM6} (expressed in mW cm⁻²sr⁻¹ μ m⁻¹) by means of the pre-launch calibration coefficients (Markham and Barker, 1986):

$$L_{TM6} = 0.0056322 \times DN6 + 0.1238 \tag{113}$$

Using the inverse Planck's law, the thermal radiances are converted to radiant temperature:

$$T = \frac{k_2}{\ln\left(\frac{k_1}{L_{THG}} + 1\right)}$$
(K) (114)

C.

where: k_1 and k_2 are the Landsat TM thermal band calibration constants, respectively for Landsat-4: 67.162 and 1284.3; for Landsat-5: 60.776 and 1260.56.

Ben Gorte

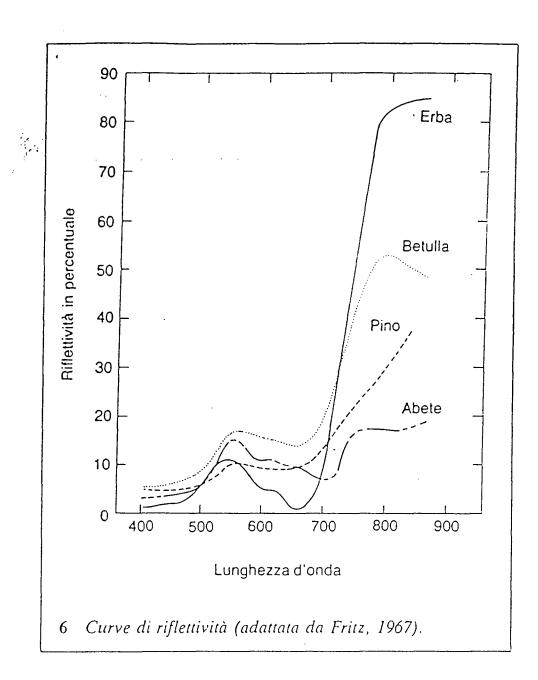
Band	Wavelength (μm)	Spectral Name	Principal Applications
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3	0.63 - 0.69	Red	Designed to sense in a chlorophyll absorption region aiding in plant species differentiation.
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5	10.4 - 12.5	Thermal Infrared	Useful in vegetation stress analysis, soil moisture discrimination, and thermal mapping applications.
,	2.08 ~ 2.35	Mid-infrared	Useful for discrimination of mineral and rock-types. Also sensitive to vegetation moisture content.

5	1.55 - 1.75	Mid-infrared	Indicati content to differ
6	10.4 - 12.5	Thermal Infrared	Useful i soil moi thermal
7	2.08 ~ 2.35	Mid-infrared	Useful f and_roc vegetati
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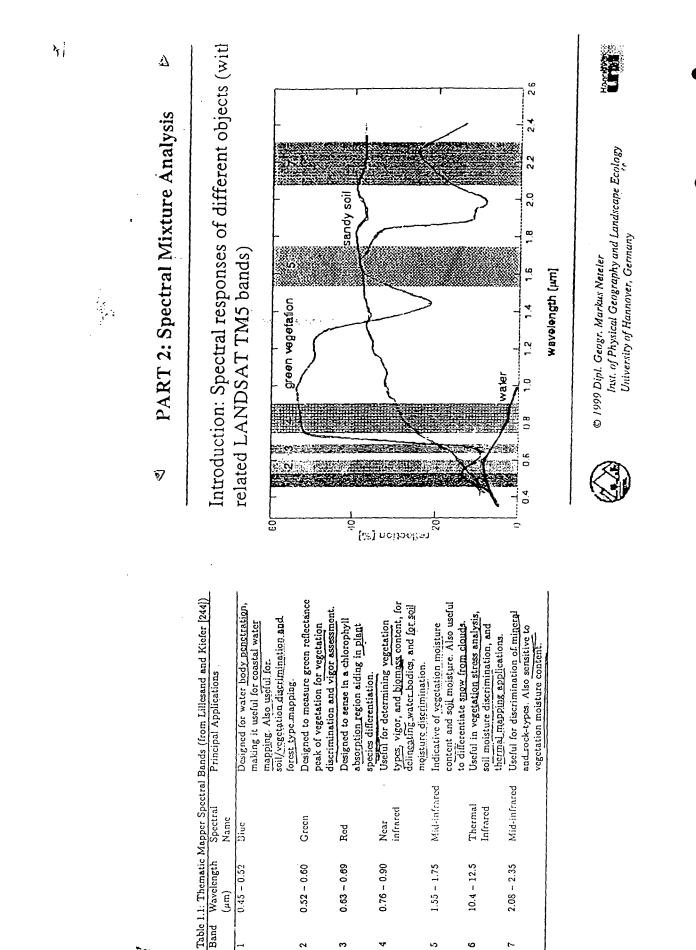
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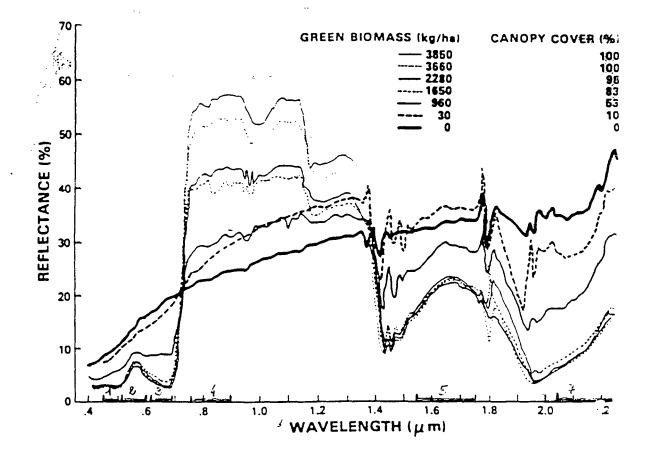
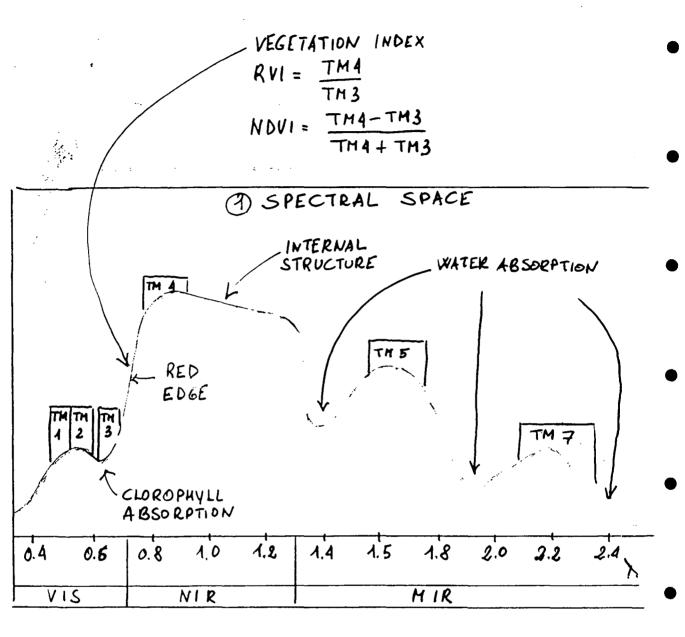
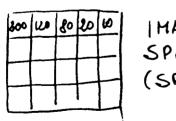


FIGURE 3.2: CROP CANOPY SPECTRAL SIGNATURE (From Deering, 1987).

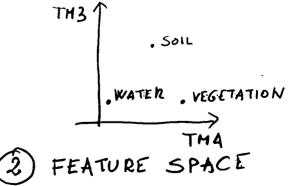
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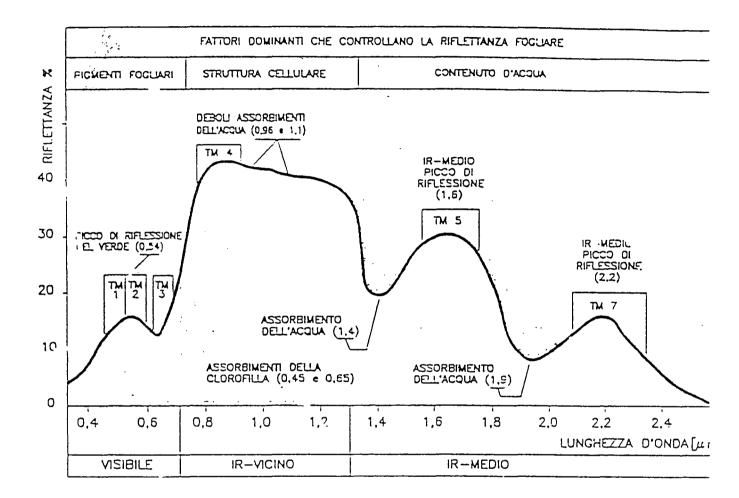
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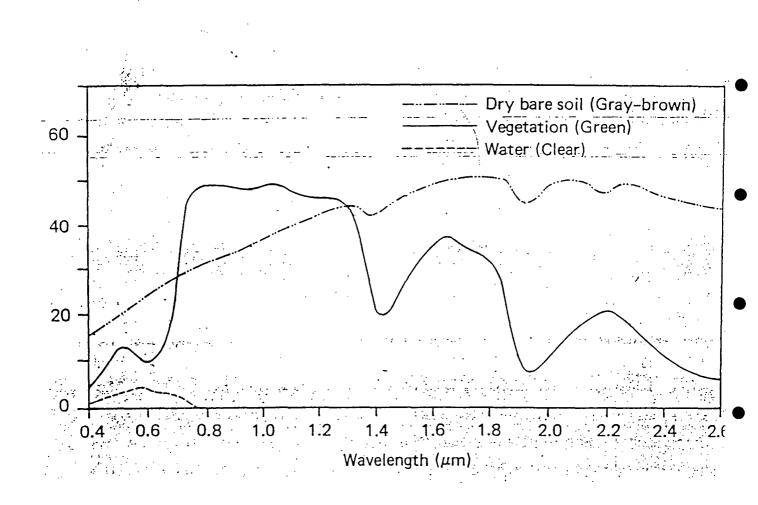




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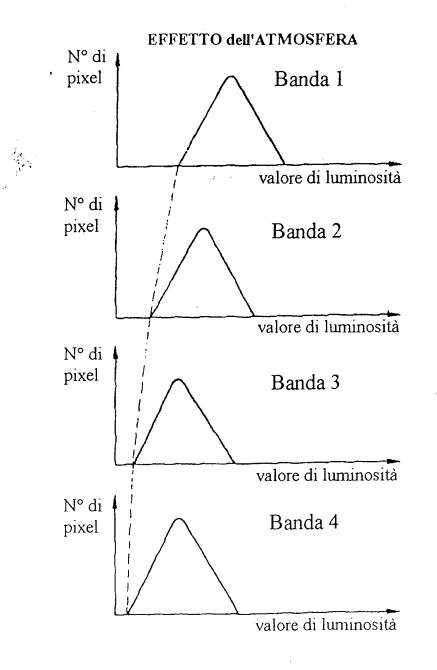
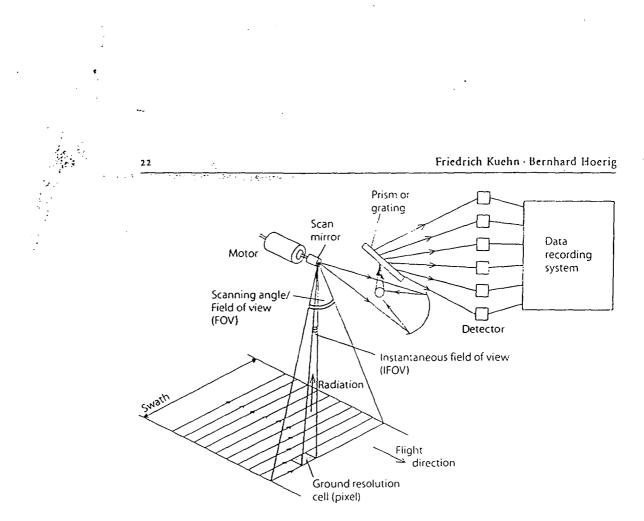


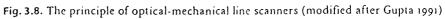
Fig. 3.6. Effetti dell'atmosfera sugli istogrammi delle immagini satellitari



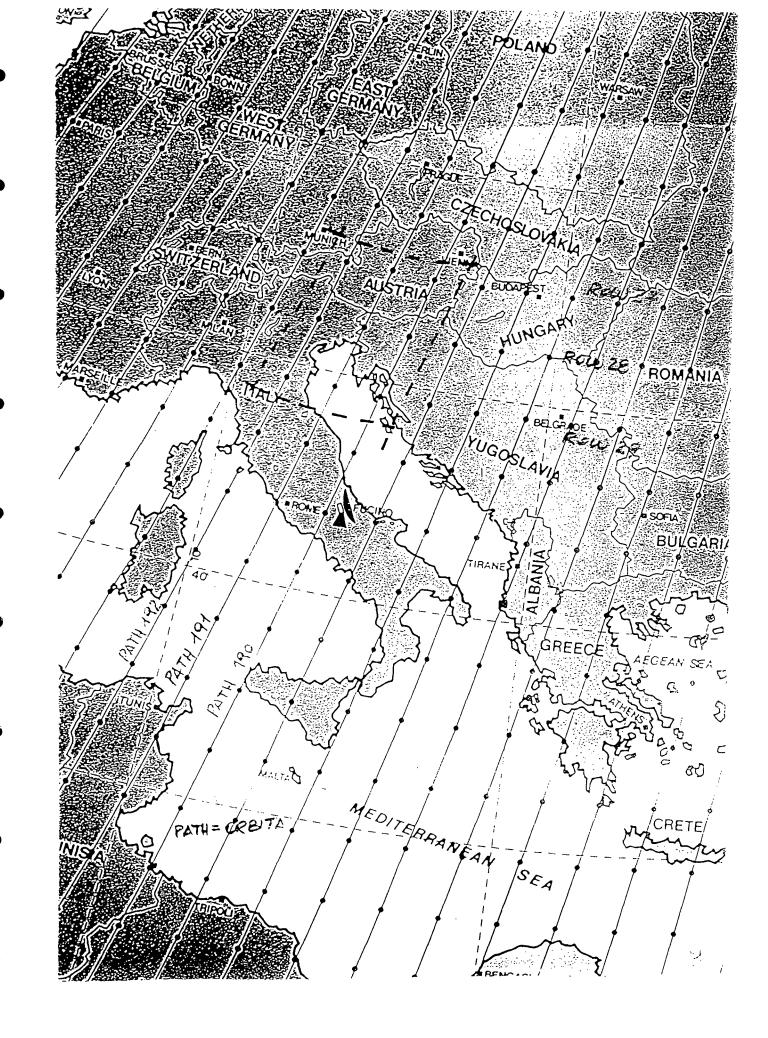
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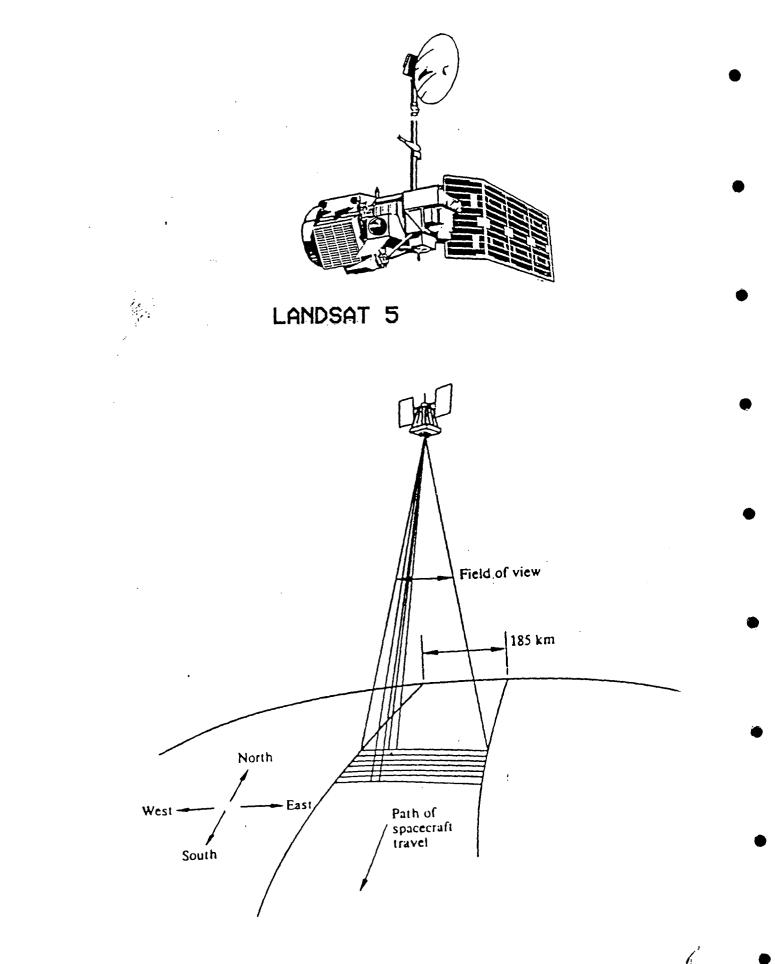




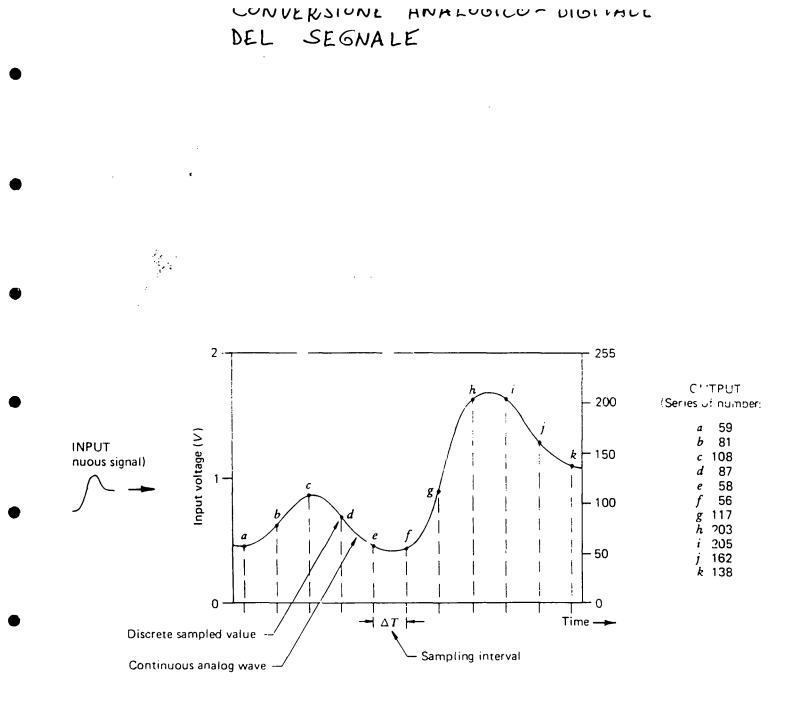




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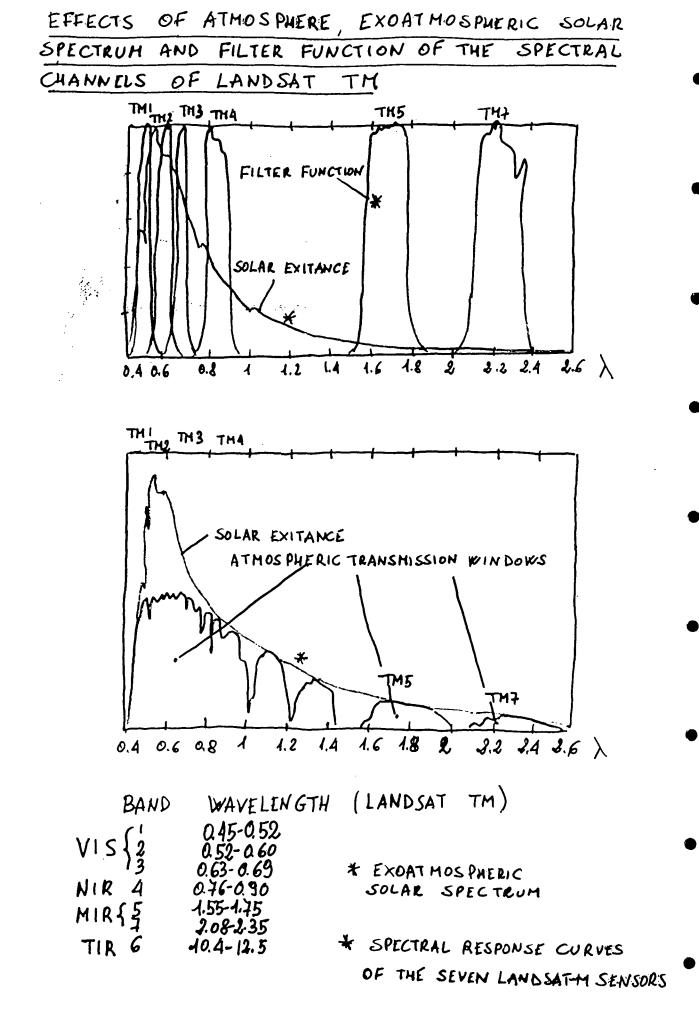


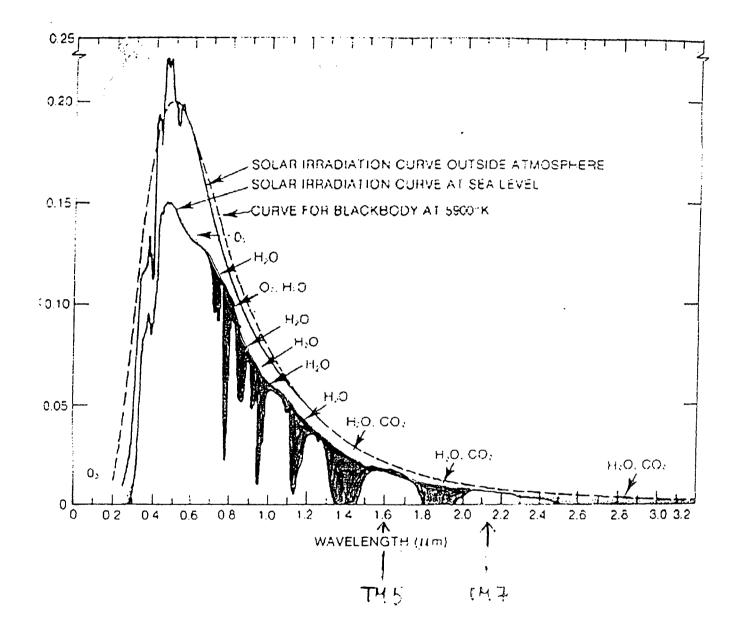
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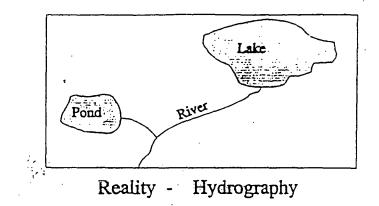
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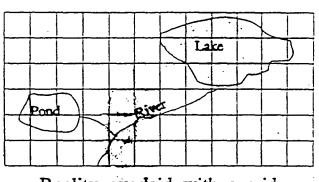




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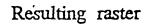




Reality overlaid with a grid

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0 = No Water Feature 1 = Water Body 2 = River



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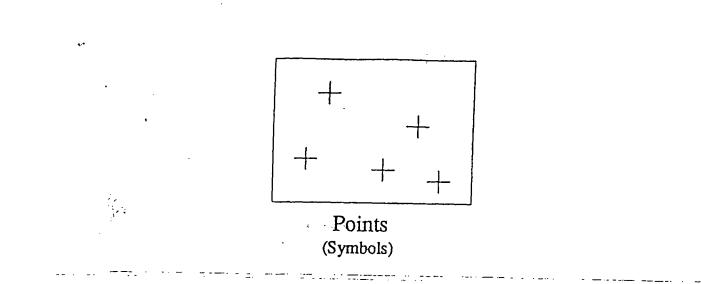


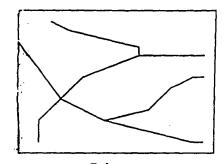
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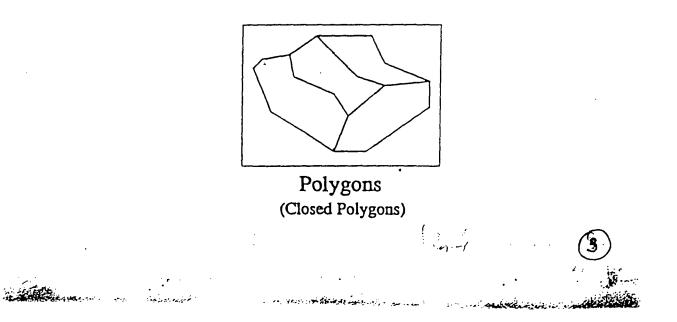
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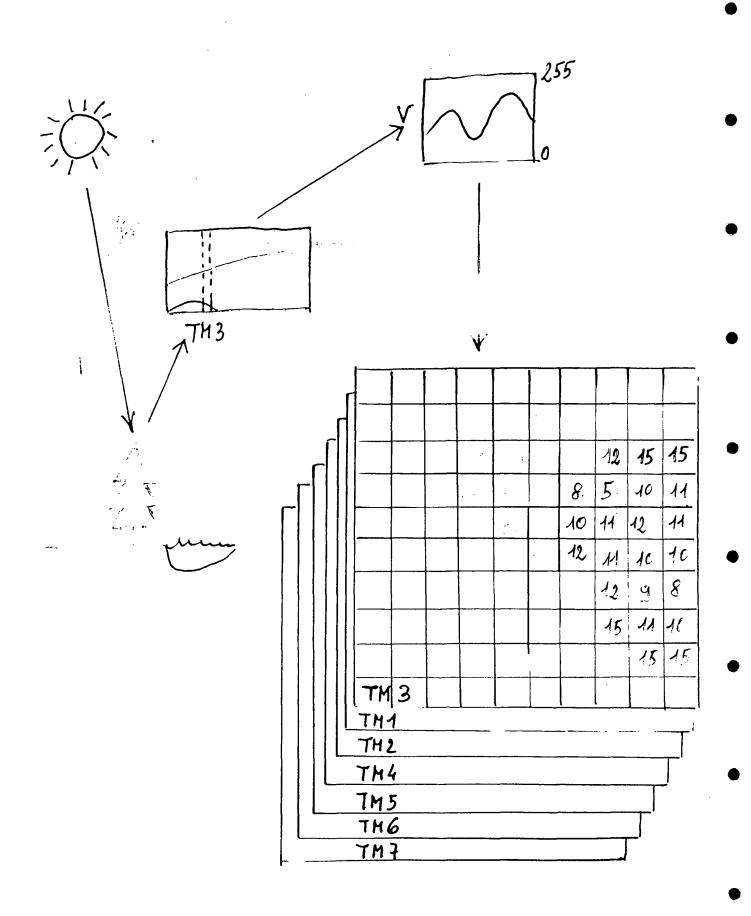
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Lines (Vectors or Open Polygons)





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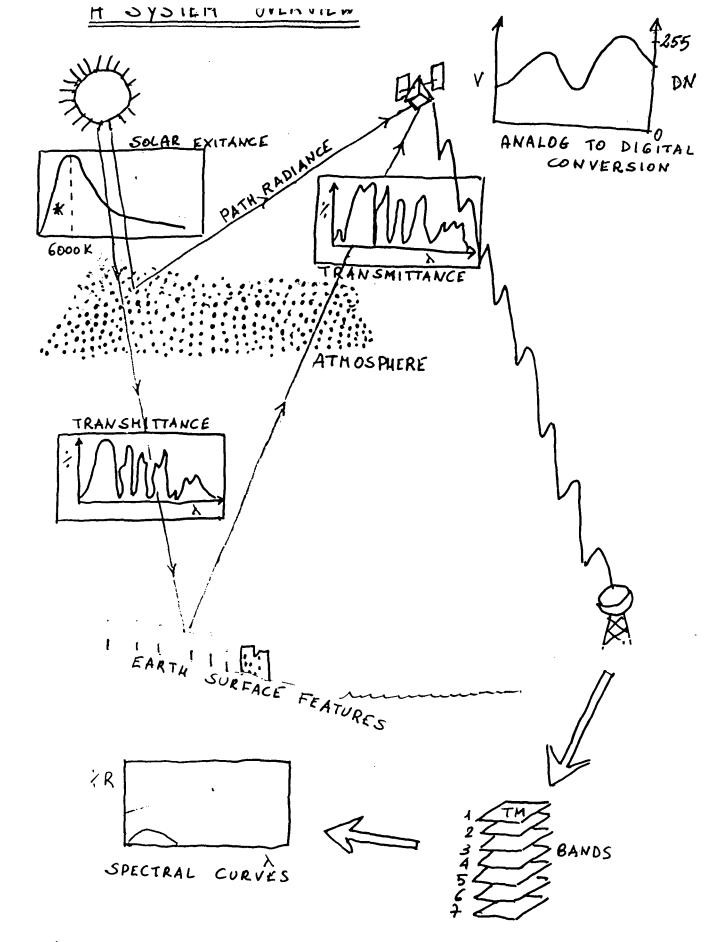
3) SCA MERING, DIELETRIC (ACTIVE) IN MICROWAVES

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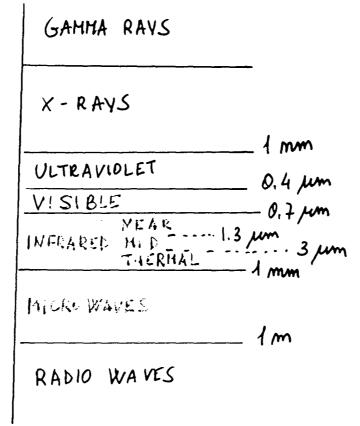


* EXO ATHOSPHERIC SOLAR SPECTRUM

REMOTE SENSING _____ THE INFORMATION CARRIER IS ELECTROMAGNETIC RADIATION

BASIC PRINCIPLE OF MULTISPECTRAL REMOTE SENSING: OBJECT \iff SHAPE OF REFLECTANCE CURVE

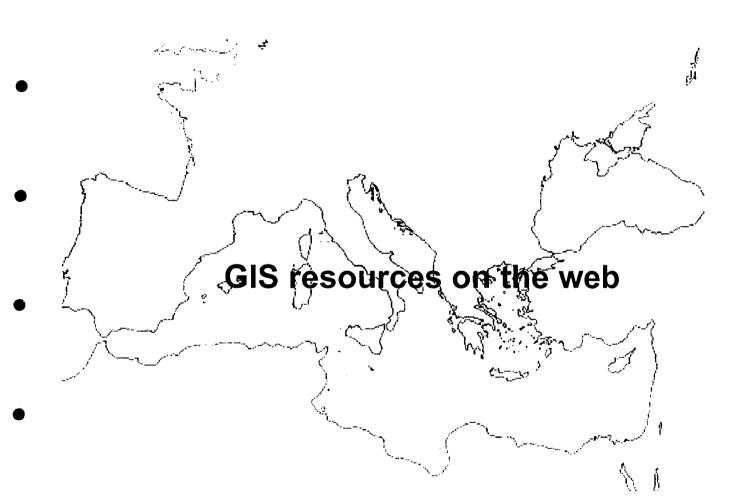
THE ELECTROMAGNETIC SPECTRUM



Practical Exercises on "Satellite Images Acquisition, Data Analysis, Information Extraction, Correction Techniques and Image Processing by A. Altobelli and M. Ghribi



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Mounir Ghribi Environmental Technologies Engineer Earth, Environmental & Marine Sciences and Technologies ICS-UNIDO AREA Science Park, Building L2 Padriciano 99, 34012 Trieste, Italy Tel. +39-040-9228105, Fax. +39-040-9228136 E-mail: mounir.ghribi@ics.trieste.it Exercises

Geographic Information Systems (GIS) MAPPING APPLICATIONS ON THE WEB

The URLs below provide information about Geographic Information Systems (GIS) and a variety of examples of how GIS has been applied to communicate information and ideas.

http://www.nativemaps.org/

The Aboriginal Mapping Network Web site is a collection of resource pages to help share information throughout the aboriginal mapping community. It is a source for both basic information on cartography, to technical information on GIS mapping.

http://www.epa.gov/enviro/html/mod/index.html

U.S. Environmental Protection Agency Maps on Demand-Interactive mapping applications and a storehouse of Environmental data.

http://www.esri.com/

Home Page for ESRI, the world's leading provider of GIS software, according to computer industry analysts. http://www.esri.com/industries/k-12/k-12form.html

ArcVoyager Special Edition, available on the CD in the 1998 Geography Awareness Week package, or on the ESRI K12 CD#4. If you don't have ArcView or the other items, you can request the ESRI K12 CD#4 (containing lots of good introductory and instructional tools)

http://www.esri.com/data/online/esri/wobmselect.html

ArcVoyager Special Edition includes a basic set of country data. There is a lot of detailed data available for free and easy download.

http://www.maproom.psu.edu/dcw/

Pick and choose your area, and you can request to download data, which will arrive as a compressed set of ".e00" files. These require use of ESRI's Import utility (converts .e00 files into ARC/INFO coverage), freely available from http://www.esri.com/company/free.html

http://www.esri.com/data/online/tiger/index.html

The full 1995 TIGER data from the U.S. Census Bureau, including streets blocks, streams, and so forth, is waiting to be downloaded, already in a shapefile format. After registering, choose the county, choose the desired layers, and download a ZIP file containing ZIP files you've selected plus readme and license in HTML format.<u>http://www.esri.com/data/online/tiger/index.html</u>

http://www.epa.gov/ostwater/BASINS/

US Watershed Data. The US Environmental Protection Agency has created a site containing vast quantities of downloadable data, for the conterminous 48 states. Click "Downloads", then, after registering, go to "BASINS GIS Data". Choose the state, pick the watershed, and, for starters, just download the elevation (DEM) and stream (RF3) data, already in compressed shapefiles.<u>http://www.esri.com/data/online/tiger/index.html</u>

http://andes.esri.com/events/seminarlist.cfm

A listing of seminars and training sessions provided by ESRI, a producer and distributer of GIS software. http://alexandria.sdc.ucsb.edu

The Alexandria Digital Library - Links to significant collections of geospatially referenced information. http://mapfinder.nos.noaa.gov

National Ocean Service Map Finder Home Page - A one-stop World Wide Web service that provides direct Internet access to primary NOS imagery and data holdings.

http://gis.mit/edu/projects

Computer resource lab at MIT - Links to Internet GIS research projects in progress.

http://www.weather.com/weather/maps/index.html

The Weather Channel - Interactive Weather-mapping site.

http://plue.sedac.ciesin.org/plue/ddviewer

Demographic data viewer home page Java-Based interactive census data viewer.

http://www.fmri.usf.edu/surf

Florida Marine Research Institute-Interactive mapping application for browsing the conditions and vulnerability of any of the 2,111 watersheds within the continental United States.

http://www.ramaker.com/gis/pds/pdsmapfrm.htm

Village of Prairie Due Sac, Wisconsin - Interactive GIS Web site where citizens can access local GIS data online.

http://crusty.er.usgs.gov:80/mapit U.S. Geological Survey Map-It-Form-based map generator uses latitude/longitude coordinates to create local maps. http://thoth.sbs.ohio-state.edu/osumap/ The OSU MAP-for-the-PC is primarily intended for instructional use in colleges and universities, although its enhanced capabilities may also make it useful in other situations (e.g., small planning agencies or in the third world) where professional staff need some introduction to GIS prior to the acquisition and utilization of a larger. commercial GIS. http://www.hdm.com/gis3.htm Great GIS net sites http://www.mrsc.org/planning/gis/gis.htm This information on geographic information systems (GIS) was compiled by Carol Shenk, a graduate student in library and information science at the University of Washington and a former MRSC employee. http://www.mapsonus.com Maps on Us - A map, route and yellow pages service http://www.gis.umn.edu/snf/info/brochures/maps.html Superior National Forest Map - Interactive Map Server http://www.gisca.adelaide.edu.au/cgi-bin/eco/ecogis Eco-tourism Interactive GIS- An interactive map design. http://www.visa.com Visa-ATM Locator-Interactive bank and ATM locator http://www.realtor.com Realtor.com - Search over one million new and resale homes. http://www.co.cabarrus.nc.us Cabarrus County - A county government public-access property ownership and tax records database. http://www.inforain.org Interrain Pacific INFORAIN Web Site-Bioregional Geographic Information Server for the North American coastal rain forest. http://www.gulasidorna.se/e_gula.html GulaSidorna/2 English- Sweden's largest online business directory. http://icg.fas.harvard.edu/~maps Harvard University Map collection - Links to the Massachusetts Electronic Atlas and other Internet GIS resources. http://gis.ci.ontario.ca.us/gis City of Ontario, California Geographic Information Server - A parcel search and site locator. http://www.macavsat.org Metropolitan Airports Commission Department of Environment Web site - Web-based airport noise and operations monitoring system at the Minneapolis Metropolitan Airports Commission. http://godiva.geoserve.com/forms/godiva.htm Godiva Chocolate - Godiva Chocolatier U.S. store locator http://www.nursinghomeinfo.com Nursing Home INFO-Nationwide Nursing Home Directory. Click on "Find a Nursing Home." http://openspace.ci.boulder.co.us/gis/osmap.html City of Boulder Open Space Interactive Map - Public Access GIS site established to promote discussion Of open space issues in Boulder, Colorado. http://poca.osmre.gov West Virginia Division of Environmental Protection-Create customized maps of environmental issues affecting the state of West Virginia. http://www.stickmap.com Seismic Geophysical Exploration Data from @stickmap.com-Commercial data vendor site. http://www.gis.umn.edu/snf/info/brochures/maps.html Superior National Forest Map - Interactive Map Server http://www.maxwell.com/caltrans Southern California Traffic Report - Real-time traffic conditions for Southern California. http://www.kennebunk.maine.org/mapserver/home.html Town of Kennebunk - Mapserver Index - Links to several interactive street, parcel and other maps. http://www.cast.uark.edu/products/MAPPER University of Arkansas - Arkansas Interactive Mapper.

http://www.unh-eos-explorer.sr.unh.edu

UNH EOS Explorer - Prototype Java application features data from Brazil's Legal Amazon region and contains a total of thirty-seven themes.

http://www.geoinformatics.com/

GeoInformatics expresses the vision of a continuing integration of the separate disciplines dealing with spatial information. Therefore, the editorial content of the magazine addresses all links in the spatial data production chain. From acquisition (geodesy, surveying, photogrammetry, remote sensing), to processing (mapping, GIS, database management), presentation and distribution (cartography, reporting, networking).

http://www.gisdatadepot.com/catalog/

This is the place to access FREE data from online inventory of more than 2 terabytes of geospatial data.

http://www.geog.uni-hannover.de/grass/index2.html

GRASS GIS (Geographic Resources Analysis Support System) home page.

http://edcdaac.usgs.gov/1KM/comp10d.html

FTP access to Global AVHRR 10-day composite data.

http://www.pes.com/pes/demprog.html

1-Degree DEM File Finder.

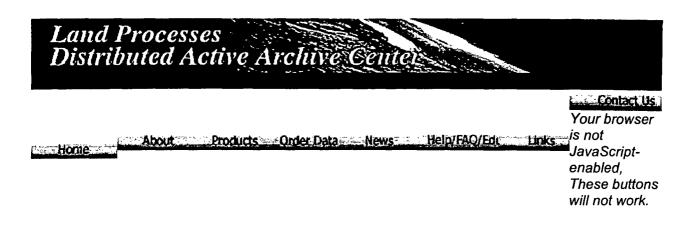
http://www.gis.umn.edu/rsgisinfo/rsgis.html

All sorts of information regarding remote sensing and geographic information systems (GIS) are provided here as a reference for RSL students and staff.

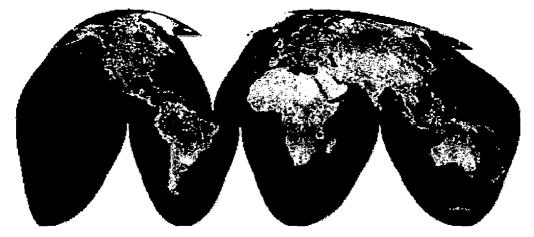
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http://edcdaac.usgs.gov/1KM/comp10d.html



FTP access to Global AVHRR 10-day composite data



New Notice to Global 1-km AVHRR Data Users

Production of the Global AVHRR composites has been temporarily suspended while funding and new processing techniques are under consideration.

Questions? Please be sure to check the detailed information available from the Global Land 1-KM AVHRR Project web site by clicking <u>here</u>.

Past Notices

In order to better serve our users, the most recent three global 10-day composites will also be made available via FTP at <ftp://edcdaacftp.cr.usgs.gov/pub/avhrr/global>. See the README file in that directory for more information.

Prior to April 15, 1995 the data *provided via this interface* did not conform to IGBP processing standards. The primary difference is in the scaling method applied to byte and integer data.

10-day composite period	od:
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May 1-10, 1992	
May 11-20, 1992	-1
May 21-30, 1992	
June 1-10, 1992	-

• The original data resolution is 1-km. Subsampled data is also available. Select the desired resolution, in kilometers.

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You can select the complete global image (default) or a geographic subset.
 Enter the coordinates of the desired subset. Coordinates can be in either decimal

degrees or degrees and minutes (see <u>examples</u>). The actual area retrieved will be <u>slightly</u> <u>larger</u>.

Select the complete global image or a geographic subset Northernmost Latitude:

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Westernmost Longitude:
Southernmost Latitude:
Easternmost Longitude:
 Each band must be retrieved separately. Select the desired band. Band: 6NDVI - For band 10 the date index files are available.
 The data can be compressed with the standard Unix compress program or the <u>GNU gzip</u> Compress gzip
 Byte Order Most significant byte first
Retrieve data
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LP DAAC EDC Home About Products Order Data News Help/FAQ/Edu Links Con
This site is hosted by the <u>USGS</u> - <u>NASA</u> Distributed Active Archive Center <u>Disclaimers, Statements and Accessibility</u>
URL: http://edcdaac.usgs.gov/1KM/comp10d.html FIRSTGOV Technical Contact: edc@eos.nasa.gov First Gut is the U.S. Generative Last Update: Wednesday, 29-May-2002 10:36:12 CDT First Gut is the U.S. Generative

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GIS Applications Using MultiSpec for Unsupervised and Supervised Classification of LANDSAT Images by A. Altobelli

MultiSpec[©]

MultiSpec is being developed at <u>Purdue University</u>, <u>West Lafayette</u>, <u>IN</u>, by <u>David Landgrebe</u> and <u>Larry Biehl</u> from the <u>School of Electrical and Computer Engineering</u> and <u>LARS</u>. It results from an on-going multiyear research effort which is intended to define robust and fundamentally based technology for analyzing multispectral and hyperspectral image data, and to transfer this technology to the user community in as rapid a manner as possible. The results of the research are implemented into MultiSpec and made available to the user community via the download pages. MultiSpec@ with its documentation@ is distributed without charge.

Description	Latest Releases	Other Information	
	Macintosh: 5-20-2002 Windows: 5-20 -2002	Download	
<u>What's New</u> <u>to 5.20.02</u>	Note: Documentation Revised on 5-01	<u>Macintosh</u>	
	٠	<u>Windows</u>	
Other Sites	Documentation	LARS	

These pages look best when viewed using <u>Netscape 1.1 or later</u>. <u>Mosaic 2.0 or later</u> and <u>Internet Explorer</u> can also be used.

Updated: May 20, 2002. Please send questions or comments to Larry Biehl, webwriter.

Copyright (c) 1994-2002 Purdue Research Foundation. Macintosh is a registered trademark of Apple Computer. Windows is a registered trademark of Microsoft Corporation.

MultiSpec© A Multispectral Image Data Analysis System

MultiSpec (@Purdue Research Foundation) is a processing system for interactively analyzing Earth observational multispectral image data such as that produced by the Landsat series of Earth satellites and hyperspectral image data from current and future airborne and spaceborne systems such as AVIRIS. The primary objective of MultiSpec is as an aid to export the results of our research into devising good methods for analyzing such hyperspectral image data. It has also found significant use in other applications such as multiband medical imagery and in K-12 and university level educational activities. There are currently in excess of several thousand known, registered users.

MultiSpec satisfies the following design goals:

- The implementation should be on a readily available computer platform which has adequate processing power, but is financially within the reach of any Earth science researcher (i.e., computer platforms < \$2000).
- The system should be easy to learn and easy to use, even for the infrequent user, using the most modern of software environments.
- The system should provide for easy import of data in a variety of formats, and easy export of results, both in thematic map and in tabular form.

The work of building the current capability began by implementing an upgraded version of the LARSYS multispectral image data analysis system. LARSYS is one of the first remote sensing multispectral data processing systems, originally created during the 1960's. A number of systems in government laboratories, university research labs, and several commercially offered products are descendants of LARSYS. The current system, called MultiSpec, has been implemented for the Apple Macintosh and PC-Windows personal workstations. A reasonably current generation, middle range machine and color display, would have a street price of less than \$2000 at the present time. Such a system is capable of classifying in excess of 6 million pixel-classes per minute using 12 bands and a Gaussian maximum likelihood scheme. Given the current cost/performance trends, even more cost-effective systems are likely to be available in the future.

New capabilities are continually added to MultiSpec as they emerge from our research on hyperspectral processing. Capabilities of the current version of MultiSpec include the following.

- Import data in either Binary or ASCII format with or without a header, and in Band Interleaved by Line (BIL), Band Sequential (BSQ), or Band Interleaved by Sample (BIS) formats. The data may have either one or two bytes per data value, and may have 4, to 16 bits per data value. In the case of two bytes per sample, the two bytes may be in either order.
- Display multispectral images in a variety of B/W or color formats using linear or equal area gray scales; display (internally generated) thematic images also in B/W or color, with an ability to control the color used for each theme. ArcView Shape Files may be overlain on the images.
- Histogram data for use in determining the gray scale regime for a display or for listing and graphing.
- **Reformat** the data file in a number of ways, e.g., by adding a standard header, changing from any one of the three interleave formats to either of the other two, editing out channels, combining files, adding or modifying channel descriptions, mosaicing data sets, changing the geometry of a data set, and a number of other changes.

http://www.ece.purdue.edu/~biehl/MultiSpec/description.html

- Create new channels of data from existing channels. The new channels may be the result of a principal components or feature extraction transformation of the existing ones, or they may result from the ratio of a linear combination of existing bands divided by a different linear combination of bands.
- Cluster data using either a single pass or an iterative (isodata) clustering algorithm. Save the results for display as a thematic map. Cluster statistics can also be saved as class statistics. Use of clustering followed by ECHO spectral/spatial classification provides an effective multivariate scene segmentation scheme.
- Define classes via designating rectangular or polygonal training fields or mask image files, compute field and class statistics, and define test fields for use in evaluating classification results quantitatively. A feature called "Enhance Statistics" also allows one to improve the extent to which the defined class statistics fit the composite of all data in the data set. A covariance estimation scheme (LOOC) can optimize that estimate for small training sets.
- Determine the best spectral features to use for a given classification using (a) searching for the best subset of features using any of five statistical distance measures, (b) a method based directly upon decision boundaries defined by training samples, or (c) a second method based directly upon the discriminant functions. Also included are methods especially designed to search for narrow spectral features such as spectroscopic characteristics, and for use of projection pursuit as a means of further improving the features extracted.
- Classify a designated area in the data file. Six different classification algorithms are available: use of minimum distance to means, correlation classifier (SAM), matched filter (CEM), Fisher linear discriminant, the Gaussian maximum likelihood pixel scheme, or the ECHO spectral/spatial classifier. Save the results for display as a thematic map, with or without training and test fields being shown. Apply a threshold to a classification, and generate a probability/threshold map showing the degree of membership of each pixel to the class to which it was assigned.
- List classification results of training or test areas in tabular form on a per field, per class, or groups of classes basis.
- Show a graph of the spectral values of a currently selected pixel or the mean ± s for a selected area. Show scatter diagrams of data from pairs of bands and ellipses of concentration for training sets and selected areas. Show a graph of the histograms of the class or field data values used for training. Show the coordinates of a currently selected area.
- Show a color presentation of the correlation matrix for a field or class as a visualization tool especially for hyperspectral data.
- Several additional **utility functions** including listing out a subset of the data e.g., for use externally, conducting principal component analysis, etc.
- Transfer intermediate or final results, be they text, B/W image or color image, to other application programs such as word processors, spreadsheet, or graphics program by copying and pasting or by saving and then opening the saved file within another application.

The MultiSpec implementation is carried out in such a way that the primary limit on the number of lines or columns of the data, the number of spectral bands, etc., are those determined by the available disk and memory space. Taken together, these capabilities provide a state-of-the-art capability to analyze moderate and high dimensional multispectral data sets of practical size. All versions, along with a 170+ page document listing its capabilities in more detail and providing tutorial exercises in its use, is available, along with substantial additional documentation, via the World Wide Web.

| <u>MultiSpec</u> | Description | <u>Other Information</u> | <u>What's New</u> | <u>Download Mac</u>, <u>Win</u> | <u>Documentation</u> | Introduction to Multivariate Data Analysis by E. Feoli

Introduction to Multivariate Data Analysis

by

Enrico Feoli

Draft for ICS-UNIDO Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas" Iglesias-Cagliari, Italy, 4-8 June 2002

Multivariate data analysis deals with matrices of data of different nature. The matrices contain descriptions of objects, at the same time or in different times, by variables that can be measured in different scales. Data matrices generate multidimensional spaces in which the row and column vectors find their mutual position. In order to represent the multidimensional space it is useful to standardize and/or normalise the variables and the objects.

Some basic formulas:

Translation	$x_{t} = x - x_{\min}$	(0.1)
Centering	$x_{t} = x - \overline{x}$	(0.2)
Relative centering	$x_{t} = \frac{x - \overline{x}}{\sqrt{n - 1}}$	(0.3)
Standardisation	$z = x_r = \frac{x - \overline{x}}{\sqrt{\frac{\sum (x - \overline{x})^2}{n}}}$	(0.4)
Standardisation with max	$x_{t} = \frac{x}{x_{\max}}$	(0.5)
Standardization and translation	$x_{t} = \frac{x - x_{\min}}{x_{\max} - x_{\min}}$	(0.6)
Normalization	$x_{i} = \frac{x}{\sqrt{\sum x_{i}^{2}}}$	(0.7)
Normalization of centred variables	$x_{t} = \frac{x - \overline{x}}{\sqrt{\sum (x - \overline{x})^{2}}}$	(0.8)
Standardization by total	$x_{i} = \frac{x}{\sum x}$.9)

Double centering
$$x_t = x - \overline{x_r} - \overline{x_c} + \overline{x}$$
 (0.10)

Deviation from
expectation
$$x_t = x - \frac{\sum x_r \sum x_c}{\sum \sum x}$$
 (0.11)

$$norma_{y} = \sqrt{x_{1j}^{2} + x_{2j}^{2} + \dots + x_{kj}^{2}} = \sqrt{\sum_{i=1}^{k} x_{ij}^{2}}$$
(0.12)

An example of a data matrix

Transformation parameters

	1	2	3	4
ph	6.8	5.9	7.5	7.1
A (m)	1200	1000	950	1150
T (C°)	18	21	22	19

Transormation by Eq. (0.5)

-		2	3	4
ph	.907	.787	1.00	.947
A (m)	1.00	.833	792	.958
<i>Т (С</i> °)	.818	.955	1.00	.864

	x _{min}	x _{max}	\overline{x}	S
ph	5.9	7.5	6.83	.68
A (m)	950	1200	1075	119
T (C°)	18	22	20	11.83

Transformation by Eq. (0.6)

	1	2	3	4
ph	.563	.000	1.00	.750
A (m)	1.00	.200	.000	.800
Т (C°)	.000	.750	1.00	.750

Transformation by Eq. (0.4)

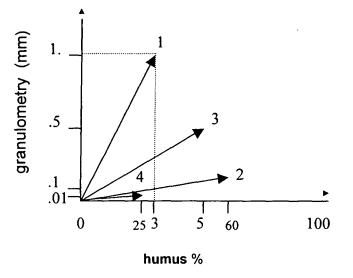
	1	2	3	4
ph	037	-1.36	.993	.404
A (m)	1.05	630	-1.05	.630
T (C°)	-1.10	.548	1.10	548

Transformation by Eq. (0.8).

<u> </u>				
	1	2	3	4
ph	.496	.431	.547	.518
A (m)	.556	.463	.440	.532
T (C°)	.449	.523	.548	.474
T (C°)	.449	.523	.548	

An example of another data matrix

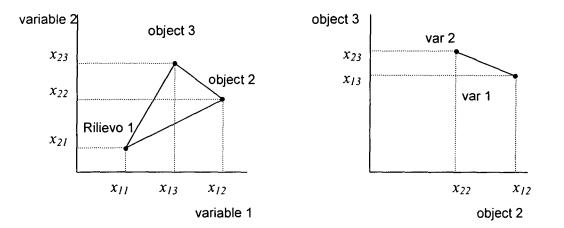
	1	2	3	4
Humus (peso %)	30	60	50	25
Granulometry (mm)	1	0.1	0.5	0.01
ph	6.5	7.7	8.1	7.3
Potassium (p.m.)	52	102	78	120



DUALITY OF THE MULTIDIMENSIONAL SPACE:

Modality Q

Modality R



The mutual position of the vectors in the multidimensional space can be defined by the resemblance functions. They are scalars that are functions of the values of the variables in the objects.

The application of resemblance functions to the data matrices gives origin to RESEMBLANCE Matrices:

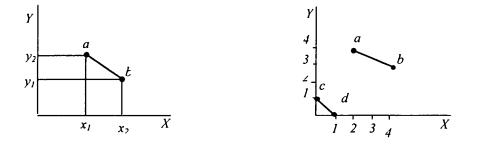
		·				
		1	2	3		n
	1	s ₁₁	s ₁₂	S ₁₃		s _{in}
	2	s ₂₁	S ₂₂	S 2 3	S _{ji}	S _{2n}
<i>S</i> =	3	\$31	S ₃₂	S33	•••	S _{3n}
			S _{ij}		····	
	n	S _{n1}	S _{n2}	S _n 3		S _{nn}

A symmetric matrix

DISTANCES :

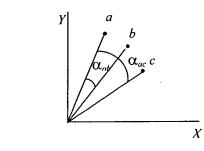
$$D_{euclidea(a,b)} = \sqrt{\sum_{i=1}^{m} (x_{ia} - x_{ib})^2}$$
(0.13)

$$D_{corda(a,b)} = \sqrt{2 \left(1 - \frac{\sum_{i=1}^{m} x_{ia} x_{ib}}{\sqrt{\sum x_{ia}^{2} \sum x_{ib}^{2}}} \right)}$$
(0.14)

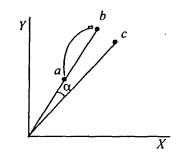


Scalar products:

$$S_{PS(a,b)} = \sqrt{\sum_{i=1}^{m} x^{2}_{ia}} \sqrt{\sum_{i=1}^{m} x^{2}_{ib}} \cos \alpha$$
(0.15)



$$\cos \alpha = \frac{\sum_{i=1}^{m} x_{ia} x_{ib}}{\sqrt{\sum_{i=1}^{m} x^{2}_{ia}} \sqrt{\sum_{i=1}^{m} x^{2}_{ib}}}$$



$$S_{PS(a,b)} = \frac{\sum_{i=1}^{m} x_{ia} x_{ib}}{\max(\sum_{i=1}^{m} x^{2}_{ia}, \sum_{i=1}^{m} x^{2}_{ib})}$$
(0.16)

$$S_{SR(a,b)} = \frac{\sum_{i=1}^{m} x_{ia} x_{ib}}{\sum_{i=1}^{m} x^{2}_{ia} + \sum_{i=1}^{m} x^{2}_{ib} - \sum_{i=1}^{m} x_{ia} x_{ib}}$$
(0.17)

$$r_{i,h} = \frac{\sum_{j=1}^{n} (x_{ij} - \overline{x_i})(x_{hj} - \overline{x_h})}{\sqrt{\sum_{j=1}^{n} (x_{ij} - \overline{x_i})^2} \sqrt{\sum_{j=1}^{n} (x_{hj} - \overline{x_h})^2}}$$
(0.18)

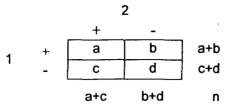
$$r_{i,h} = \frac{n \sum_{j=1}^{n} x_{ij} x_{hj} - \sum_{j=1}^{n} x_{ij} \sum_{j=1}^{n} x_{hj}}{\sqrt{n \sum_{j=1}^{n} x_{ij}^{2} - (\sum_{j=1}^{n} x_{ij})^{2}} \sqrt{n \sum_{j=1}^{n} x_{hj}^{2} - (\sum_{j=1}^{n} x_{hj})^{2}}}$$
(0.19)

$$S_{i,h} = \frac{\sum_{j=1}^{n} (x_{ij} - \overline{x_i})(x_{hj} - \overline{x_h})}{n-1}$$
(0.20)

RELATIONSHIP BETWEEN DISTANCE AND SCALAR PRODUCT:

$$D^{2}_{(a,b)} = \sum x_{ia}^{2} + \sum x_{ib}^{2} - 2\sum x_{ia} x_{ib}$$
(0.21)

ī.



$$S_{Jaccard} = \frac{a}{a+b+c}$$
(0.22)

$$S_{Sorensen} = \frac{2a}{2a+b+c}$$
(0.23)

$$S_{Ochiai} = \frac{a}{\sqrt{(a+b)(a+c)}}$$
(0.24)

Euclidean distance for binary data:

$$D_{euclidea} = \sqrt{b+c} \tag{0.25}$$

Cord distance for binary data (0.24):

$$D_{corda} = \sqrt{2\left(1 - \frac{a}{\sqrt{(a+b)(a+c)}}\right)}$$
(0.26)

Correlation coefficient:

$$r = phi = \frac{ad - bc}{\sqrt{(a+b)(a+c)(b+d)(c+d)}}$$
(0.27)

r is related to χ^2 :

$$r = \sqrt{\frac{\chi^2}{n}} \tag{0.28}$$

Simple matching" (Sokal & Michener),

$$S_{SM} = \frac{a+d}{n} \tag{0.29}$$

Yule's index:

$$S_{Yule} = \frac{ad - bc}{ad + bc} \tag{0.30}$$

Gower' index for mixed data::

$$S_{Gower(a,b)} = \frac{\sum_{i=1}^{m} w_{iab} S_{iab}}{\sum_{i=1}^{m} w_{iab}}$$
(0.31)

 $s_{i(a,b)}$ has values between 0 ed 1. For nominal data $s_{i(a,b)}=1$ if the objects agree for character i, $s_{i(a,b)}=0$ if they disagree, for continuous data:

$$s_{i(a,b)} = 1 - \frac{|x_{ia} - x_{ib}|}{R}$$
(0.32)

R is xmax -x min.

Transformation of similarity to dist:

 $D = 1 - S \tag{0.33}$

$$D = -\log S \tag{0.34}$$

$$D = 1 / (1 + S) \tag{0.35}$$

Examples of calculus:

a data table					
	1	2	3	4	media
Specie A	10	5	15	20	12.5
Specie B	12	20	0	0	8
Specie C	0	10	7	15	8

Euclidean distance [eq. (0.13)]:

$$D_{\text{euclides (1,2)}} = \sqrt{(10-5)^2 + (12-20)^2 + (0-10)^2} = 13.75$$

Cord distance [eq. (0.14)]:

$$D_{\text{corda}(1,2)} = \sqrt{2\left(1 - \frac{10 \times 5 + 12 \times 20 + 0 \times 10}{\sqrt{(10^2 + 12^2 + 0^2) \times (5^2 + 20^2 + 10^2)}}\right)} = \sqrt{2\left(1 - \frac{50 + 240}{\sqrt{(100 + 144) \times (25 + 400 + 100)}}\right)} = \sqrt{2\left(1 - \frac{290}{\sqrt{128100}}\right)} = \sqrt{2\left(1 - 0.81\right)} = 0.616$$

Scalar product:

 $S_{PS(1,2)} = 10 \times 5 + 12 \times 20 + 0 \times 10 = 290$

$$\cos\alpha_{1,2} = \frac{10 \times 5 + 12 \times 20 + 0 \times 10}{\sqrt{(10^2 + 12^2 + 0^2)}\sqrt{(5^2 + 20^2 + 10^2)}} = \frac{290}{\sqrt{244}\sqrt{525}} = \frac{290}{357.91} = 0.810$$

$$S_{PS(1,2)} = \frac{10 \times 5 + 12 \times 20 + 0 \times 10}{\max[(10^2 + 12^2 + 0^2), (5^2 + 20^2 + 10^2)]} = \frac{290}{\max(244, 525)} = \frac{290}{525} = 0.552$$

Similarity ratio [eq. (0.17)]:

$$S_{SR(1,2)} = \frac{10 \times 5 + 12 \times 20 + 0 \times 10}{(10^2 + 12^2 + 0^2) + (5^2 + 20^2 + 10^2) - (10 \times 5 + 12 \times 20 + 0 \times 10)} = \frac{290}{244 + 525 - 290} = 0.605$$

Coefficiente di correlazione [eq. (0.18)]:

$$r_{A,B} = \frac{(10-12.5)(12-8) + \dots + (20-12.5)(0-8)}{\sqrt{(10-12.5)^2 + \dots + (20-12.5)^2}\sqrt{(12-8)^2 + \dots + (0-8)^2}} = \frac{(-2.5)4 + (-7.5)(2+2.5)(-8) + 7.5(-8)}{\sqrt{6.25 + .56.25 + 6.25 + 56.25}\sqrt{16 + 144 + 64 + 64}} = \frac{-10-90-20-60}{\sqrt{125}\sqrt{288}} = \frac{-180}{189.74} = -0.949$$

Binary data

An example of	binary dat	a				
	1	2	3	4	5	6
Specie A	1	1	0	0	0	1
Specie B	1	0	1	0	1	1
Specie C	1	1	1	1	0	0
Specie D	1	0	1	1	1	1
Specie E	0	1	0	1	0	1

Contingency table between objects

		Objec							
presente assente									
e	presente	a=2	b=1	a+b=3					
bject	assente	c=0	d=2	c+d=2					
Ő		a+c=2	b+d=3	n =5					

Index of Jaccard [eq. (0.21)]:

$$S_{Jaccard(3,5)} = \frac{2}{2+1+0} = 0.667$$

Index of Sorensen [eq. (0.22)]:

$$S_{Sorensen (3.5)} = \frac{2 \times 2}{2 \times 2 + 1 + 0} = 0.8$$

Index of Ochiai [eq. (0.23)]:

$$S_{\text{Ochiai}(3,5)} = \frac{2}{\sqrt{3 \times 2}} = 0.816$$

Euclidean distance [eq. (0.24)]:

$$D_{euclidea(3,5)} = \sqrt{1+0} = 1$$

Cord distance [eq. (0.25)]:

$$D_{\text{corda}(3,5)} = \sqrt{2\left(1 - \frac{2}{\sqrt{(3 \times 2)}}\right)} = \sqrt{2(1 - 0.816)} = 0.606$$

Variables association:

Correlation coefficient [eq. (0.26)]:

$$r_{B,D} = \frac{4 \times 1 - 0 \times 1}{\sqrt{4 \times 5 \times 1 \times 2}} = 0.632$$

"Simple matching" [eq. (0.28)]:

$$S_{SM(B,D)} = \frac{4+1}{6} = 0.833$$

Contingency table between variables

		Specie	es D	
B	presente	a=4	b=0	a+b=4
Species	assente	c=1	d=1	c+d=2
Sp		a+c=5	b+d=1	n=6

Index of Yule [eq. (0.29)]:

$$S_{\text{Yule}(B,D)} = \frac{4 \times 1 - 0 \times 1}{4 \times 1 + 0 \times 1} = 1$$

Mixed data:

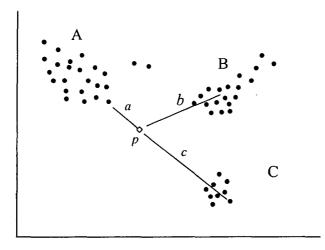
Example of a mixed data matrix											
	1	2	3	4							
A	+	+	-	-							
В	L	V	1	L							
С	r	r	W	w							
D (cm)	5	7	3	4							
E	10	7	5	7							

$$S_{Gower(1,2)} = \frac{1+0+1+\left[1-(7-5)/(7-3)\right]+\left[1-(10-7)/(10-5)\right]}{1+1+1+1+1} = 0.58$$

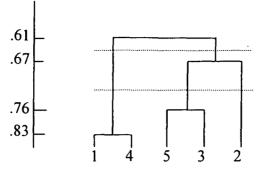
$$S_{Gower(3,4)} = \frac{1+0+1+\left[1-(4-3)/(7-3)\right]+\left[1-(7-5)/(10-5)\right]}{1+0+1+1+1} = 0.8375$$

$$S_{Gower(2,3)} = \frac{0+0+0+\left[1-(7-3)/(7-3)\right]+\left[1-(7-5)/(10-5)\right]}{1+0+1+1+1} = 0.15$$

Classification algorithms:



a= single linkage, b=average linkage, c= complete linkage



Example of dendrogram

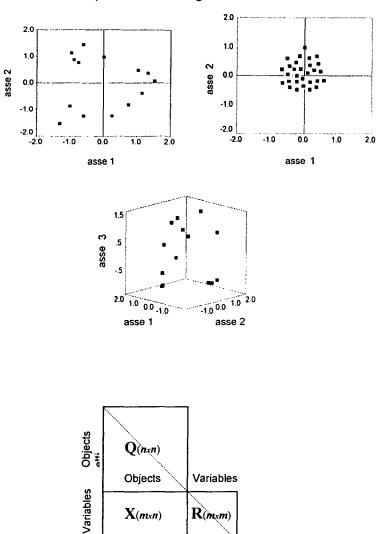
Cross validation

Objects	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25
I classif.	1	1	1	1	1	1	1	1	1	1	2	2	2	2	2	2	2	2	2	2	2	2	2	3	3
II classif.	1	1	1	1	1	3	3	3	3	3	1	1	1	2	2	2	2	2	2	2	2	2	2	2	2

Cross table of the two classificatios. Index of Cramer=0.59.

		1	2	3	
	1	5	0	5	10
I	2	3	10	0	13
	3	0	2	0	2
		8	12	5	25

ORDINATION



Objects and variables are disposed according to axes:

Given a matrix X two symmetric matrices may be obtained

 $(SB_i - \lambda_i B_i) = 0 \tag{0.1}$

$$(S - \lambda_i I_p) B_i = 0 \tag{0.2}$$

The scalars λ_i are eigenvalues of *S*, while the vector B_i associated to λ_i is called eigenvector.

$$\begin{pmatrix} S - \lambda_i I_p \end{pmatrix} & B_i & 0 \\ \begin{pmatrix} s_{11} - \lambda & s_{12} & \dots & s_{1p} \\ s_{21} & s_{22} - \lambda & \dots & s_{2p} \\ \dots & \dots & \dots & \dots \\ s_{p1} & s_{p2} & \dots & s_{pp} - \lambda \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \\ \dots \\ b_p \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \dots \\ 0 \end{pmatrix}$$

$$(s_{11}-\lambda)b_{1} + s_{12}b_{2} + \dots + s_{1p}b_{p} = 0$$

$$s_{21}b_{1} + (s_{22}-\lambda)b_{2} + \dots + s_{2p}b_{p} = 0$$

$$\dots + \dots + \dots + \dots = 0$$

$$s_{p1}b_{1} + s_{p2}b_{2} + \dots + (s_{pp}-\lambda)b_{p} = 0$$
(0.3)

$$\left|S-\lambda,I\right|=0\tag{0.4}$$

$$\begin{vmatrix} s_{11} - \lambda & s_{12} & \dots & s_{1p} \\ s_{21} & s_{22} - \lambda & \dots & s_{2p} \\ \dots & \dots & \dots & \dots \\ s_{p1} & s_{p2} & \dots & s_{pp} - \lambda \end{vmatrix} = 0$$
(0.5)

The solution of the polynomial of p degree gives the λ_{i} . The efficiency of the solution is:

$$\lambda_i^{\%} = \frac{\lambda_i}{\sum \lambda_i} 100 \tag{0.6}$$

Example

Give	en the r	natrix					
			(Object	5		
		1	2	3	4	5	Aver:
ij	1	1	2	5	0	5	2.6
Vari.	2	3	4	5	2	3	3.4

Let's compute the covariance between variables 1 and 2 and the respective variances:

$$s_{1,2} = \frac{(1-2.6)(3-3.4) + \dots + (5-2.6)(3-3.4)}{4} = 1.7$$

:
$$S_{mxm} = \frac{5.3 \quad 1.7}{1.7 \quad 1.3}$$

Let's compute now the eigenvalues:

$$|S - \lambda_{i}I_{2}| = \begin{vmatrix} 5.3 - \lambda & 1.7 \\ 1.7 & 1.3 - \lambda \end{vmatrix} = 0$$

(5.3 - λ) × (1.3 - λ) - 1.7 × 1.7 = 0

 $\lambda^2 - 6.6\lambda + 4 = 0$

$$\lambda_1 = 3.3 + \frac{\sqrt{3.3^2 - 4}}{1} = 5.925$$
 $\lambda_2 = 3.3 - \frac{\sqrt{3.3^2 - 4}}{1} = 0.675$

$$\sum \lambda_i = \lambda_1 + \lambda_2 = 5.925 + 0.675 = 6.6$$

$$\sum s_{ii} = s_{11} + s_{22} = 5.3 + 1.3 = 6.6$$

$$\lambda_{1}^{\%} = (5.925 / 6.6) \times 100 = 89.77$$

 $\lambda_{2}^{\%} = (0.675 / 6.6) \times 100 = 10.23$

Now let's calculate the eigenvectors

 $\begin{cases} (5.3 - 5.925)b_1 + 1.7b_2 = 0\\ 1.7b_1 + (1.3 - 5.925)b_2 = 0 \end{cases}$

$$\begin{cases} 0.625b_1 + 1.7b_2 = 0\\ -1.7b_1 + 4.625b_2 = 0 \end{cases}$$

let b_1 =1, then:

$$\begin{cases} 1.7b_2 = 0.625 \\ 4.625b_2 = 1.7 \end{cases}$$

*b*₂ is 0.3676.

By assigning arbitrary values the elements of eigenvectors remain always proportional, in fact if we put $b_1=3$, $b_2=3x(0.3676)=1.1028$, and if $b_2=1$, $b_1=1.7/0.625=2.72$; in all the cases the $b_1/b_2 = 2.72$ and $b_2/b_1 = 0.3676$. To avoid arbitrary values the eigenvectors are normalised:

$$b_1 = \frac{1}{\sqrt{1^2 + 0.3676^2}} = 0.9386$$
$$b_2 = \frac{0.3676}{\sqrt{1^2 + 0.3676^2}} = 0.345$$

Summarising:

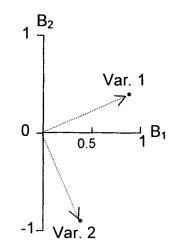
$$B_1 = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 0.3676 \end{pmatrix} \text{ normalising } = \begin{pmatrix} 0.939 \\ 0.345 \end{pmatrix}$$

With the similar procedure we find the second eigenvector B_2 corresponding to the second eigenvalue λ_2 :

$$B_2 = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \begin{pmatrix} 1 \\ -2.722 \end{pmatrix} \text{ normalising} = \begin{pmatrix} 0.345 \\ -0.939 \end{pmatrix}$$

Table of eigenvalues and eigenvectors of the example.

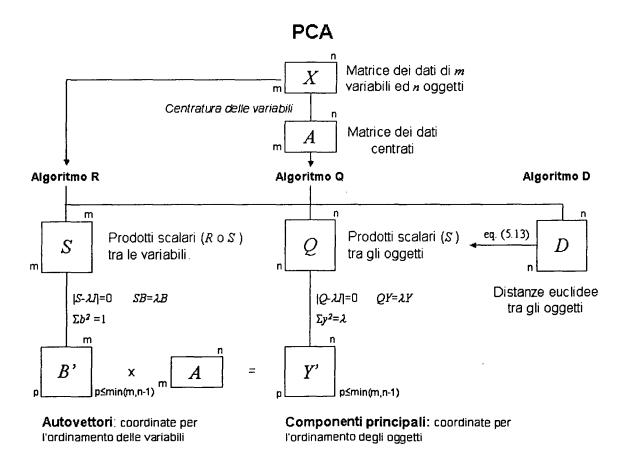
	Eigenval.	Variance %	Eigenvector	1	2
λ_1	5.925	89.77	<i>B</i> ₁	0.939	0.345
λ_2	0.675	10.23	B_2	0.345	-0.939

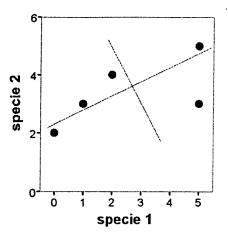


Scattergram of the two variables

Principal Component Analysis

(PCA) R, Q and D described by Orloci (1978).





Scattergram of the 5 points and PCA

$$Y_i = B_i X$$
 for $i = 1, m$ (0.7)

$$y_{ij} = \sum_{h=1}^{m} b_{hi} x_{hj} = b_{1i} x_{1j} + b_{2i} x_{2j} + \dots + b_{mi} x_{mj}$$
(0.8)

Algorithm R

$$(S - \lambda_i I_p) B_i = 0 \tag{0.9}$$

$$B_i' B_i = \sum_{j=1}^{m} b_{ij}^2 = b_{i1}^2 + b_{i2}^2 + \dots + b_{im}^2 = 1$$
(0.10)

The components are $(Y_i = B'_i A)$,

The elements in A are:



(0.11)

 F_i is the standardisation factor.

Table of F that allows different solutions of PCAautomatically normalised to the square root of eigenvalues.

Fi	Eq.	Indice di S	Eq.
1		Centred scalar product	
$\sqrt{n-1}$	(0.12)	Covariance	(0.20)
$\sqrt{\sum(x_{ij}-\overline{x})^2}$	(0.13)	Correlation coefficient	(0.18)

Anyway it is always possible to standardize the components by:

$$y_j = \frac{y_j}{\sqrt{\sum y_i^2}} \sqrt{\lambda_i}$$
(0.14)

The correlation coefficient r_{hi} between the original *h*-th variable and the *i*-th component is:

$$r_{hi} = \frac{\sqrt{\lambda_i} b_{hi}}{\sqrt{s_{hh}}} \tag{0.15}$$

where b_{hi} is the coefficient of the *i*-th component of the variable *h*-th, λ_i is the *i*-th eigenvalue and s_{hh} is the diagonal value of S of variable *h*-th. If S is a correlation matrix (*R*), s_{hh} is 1 and:

$$r_{hi} = \sqrt{\lambda_i b_{hi}} \tag{0.16}$$

D and Q algorithms

_

 d_{ij} of matrix D is transformed Q by:

$$q_{ij} = -0.5d_{ij}^2 + 0.5(D_i + D_j - D_i)$$
(0.17)

where:

$$D_{i} = \frac{\sum_{i=1}^{n} d_{i}^{2}}{n} \qquad D_{j} = \frac{\sum_{j=1}^{n} d_{j}^{2}}{n} \qquad D_{i} = \frac{\sum_{i=1}^{n} \sum_{j=1}^{n} d_{ij}^{2}}{n^{2}}$$

Data	matrix									•				
				object	S					_				
-i		1	2	3	4	5	-		1	2	3	4	5	$\sum y_{ij}^2$
Variabl.	1	-0.8	-0.3	1.2	-1.3	1.2	_	Y _I	-0.82	-0.18	1.4	-1.46	1.06	5.925
<a>	2	-0.2	0.3	0.8	0.7	-0.2	_	Y ₂	-0.09	-0.38	-0.34	0.21	.6	0.675
<i>Y</i> =	= B ['] A	$= \begin{pmatrix} 0.939\\ 0.345 \end{pmatrix}$	0.34 5 - 0.93	$\binom{5}{39} \times \begin{pmatrix} -\\ - \end{pmatrix}$	0.8 – 0 0.2 0.1	.3 1.2 3 0.8	- 1.3 0.7	1.: - 0	$\binom{2}{2} = \begin{pmatrix} -\\ - \end{pmatrix}$	- 0.82 - 0.09	- 0.18 - 0.38	1.4 - 0.34	- 1.46 0.21	1.06 0.6

The detailed calculations:

 $y_{11} = (0.939)(-0.8) + (0.345)(-0.2) = -0.819$ $y_{12} = (0.939)(-0.3) + (0.345)(0.3) = -0.178$ $y_{21} = (0.345)(-0.8) + (-0.939)(-0.2) = -0.088$ $y_{22} = (0.345)(-0.3) + (-0.939)(0.3) = -0.385$

To normalize the components of eigenvectors to $\sqrt{\lambda}$ (0.14):

$$b_{11} = \frac{0.939 \times \sqrt{5.925}}{1} = 2.2857$$

$$b_{21} = \frac{0.345 \times \sqrt{0.675}}{1} = 0.2834$$

$$b_{12} = \frac{0.345 \times \sqrt{5.925}}{1} = 0.8398$$

$$b_{22} = \frac{-0.939 \times \sqrt{0.675}}{1} = -0.7715$$

Correlations with components are:

$$r_{11} = \frac{0.939 \times \sqrt{5.925}}{\sqrt{5.3}} = 0.991$$
$$r_{21} = \frac{0.345 \times \sqrt{5.925}}{\sqrt{1.3}} = 0.737$$

Q matrix of scalar product of objects in the data	а
matrix	

	1	2	3	4	5
1	.68	.18	-1.12	1.18	92
2	.18	.18	12	.18	42
3	-1.12	12	2.08	-2.12	1.28
4	1.18	.18	-2.12	2.18	-1.42
5	92	42	1.28	142	1.48

<i>r</i> ₁₂	=	$\frac{0.345 \times \sqrt{0.675}}{\sqrt{5.3}} = 0.123$
<i>r</i> ₂₂	=	$\frac{-0.937 \times \sqrt{0.675}}{\sqrt{1.3}} = -0.675$

D matrix

	1	2	3	4	5
1	0	.707	2.24	.707	2
2	.707	0	1.58	1.41	1.58
3	2.24	1.58	0	2.92	1
4	.707	1.41	2.92	0	2.55
5	2	1.58	1	2.55	0

$$D_{1} = \frac{(0^{2} + 0.707^{2} + 2.236^{2} + 0.707^{2} + 2^{2})}{5} = 2$$
$$D_{2} = \frac{(0.707^{2} + 0^{2} + 1.58^{2} + 1.41^{2} + 1.58^{2})}{5} = 1.5$$
$$D_{r} = \frac{(0^{2} + 0.707^{2} + 2.236^{2} + ... + 2.539^{2} + 0^{2})}{5^{2}} = 2.64$$

to obtain q values by (0.17) :

$$q_{11} = -0.5 \times 0^2 + 0.5(2 + 2 - 2.64) = 0.68$$

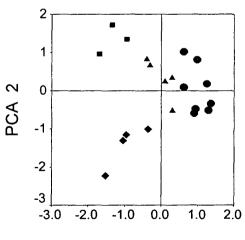
$$q_{12} = -0.5 \times 0.707^2 + 0.5(2 + 1.5 - 2.64) = 0.18$$

$$q_{22} = -0.5 \times 0^2 + 0.5(1.5 + 1.5 - 2.64) = 0.18$$

RESULTS of Q and D PCA:

Eigenvalues					Eigenvectors	;	
			1	2	3	4	5
1	5.9249	B_{I}	337	073	.576	601	.434
2	0.6751	B_2	107	469	410	.254	.732
					PCA		
			1	2	3	4	5
		Y_I	-0.820	-0.178	1.402	-1.462	1.057
		Y_2	-0.088	-0.385	-0.337	0.209	.602

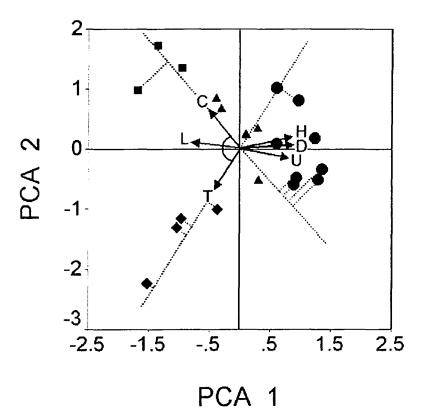
BIPLOT:





Euclidean biplot eigenvector normalized to 1 and components to the square root of the corresponding eigenvalue.

Covariance biplot eivenvecotors normalized to the square root of eigenvalues and components normalized to 1.



Correlation between variables and components					
	PCA I	PCA II			
U	.979	.139			
н	.977	.070			
G	.934	173			
L	948	.107			
т	522	787			
С	409	.860			

GIS, Remote Sensing and Spatial Decision-Support Systems (SDSS) for Coastal Zone Management by M. Ghribi

Spatial Decision-Support Systems for Integrated Coastal Area Management

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Abstract

This paper presents an overview of methods and tools of spatial data analysis useful for establishing computer based decision-support systems (DSS) for an integrated coastal area management (ICAM). The spatial DSS refer to those systems based on the use of GIS (Geographic Information Systems) technology. The integration of those information technologies provides an important source of methods that are usefully incorporated in a decision-support system and make use of geographically referenced information.

After a brief introduction on the peculiarities and unique management problems of coastal areas, the paper first describes the meaning and role of Decision Analysis (DA) and then the major tools and components for constructing a Spatial Decision-support Systems (SDSS) for Coastal Zone Management (CZM).

Key words: Methods, tools, spatial data analysis, DSS, GIS, integration information technologies, SDSS, CZM

1. Introduction

This paper is based on ICS-UNIDO experience in developing advanced information technologies for integrated coastal area management.

According to Fedra and Feoli (1998), the definition of coastal zone for the purpose of this paper is "the area, on both sides of the actual land-water interface, where the influences of land and water on each other are still a determining factor - climatically, physiographically, ecologically, or economically".

The characteristics of the coastal zone are therefore related both to the landside and the waterside.

In the first case the coastal zone shows a high population density with a large number of urban conglomerations, and as consequence, in most countries, a fast population growth.

In the water side there are fisheries and aquaculture, exploiting a generally highly productive ecosystem, that is not only productive in its own right, but plays an important role in the aquatic ecology as a breeding and nursing ground for many commercially important fish species. Very specific, valuable and vulnerable typical coastal ecosystem includes estuaries, salt marshes, mangroves, sea grass meadows and coral reefs.

The environmental problems of coastal zones are both anthropological and natural. Concerning the anthropological problems, Vallega (1993) offers a systematic classification of coastal zone uses. They include: seaports, shipping (carriers, routes, navigational aides), sea pipeline, cables, air transportation, biological resources, hydrocarbons, metallurgic resources, renewable energy, defence, recreation, waste disposal, research, archaeology, environmental preservation and protection. These are further subdivided into a total of 250 kinds of uses, all focussing more or less on the aquatic part of the coastal zone. A similar classification can be found for land-based activities, with urban development, industry (ranging from mining to manufacturing) and commerce, agriculture and recreation as the main elements in Ajjour and Drabih (1997). All these uses are causes of pollution and environmental changes.

The set of natural problems is related to the exposure of the terrestrial zone to marine impacts, and on other terrestrial impacts on the coastal waters. These include such phenomena as storms, tides, and surges, the occasional tsunami, and the prospect of sea level rise, coastal erosion, salt water intrusion in rivers and aquifers caused by (often man made) low-flow and excessive withdrawal.

Since all of the activities in coastal zone compete, for space, for resources, and for the waste absorption capacity of the coastal ecosystems, it follows that there is an obvious and often dominating spatial aspect, which makes coastal zone management a spatial problem.

2. Spatial Decision-Support Systems and Coastal Zone Management

2.1 Definitions

Definitions of decision-support systems range from: "Interactive computer based systems that decision-makers utilize data and models to solve unstructured problems" (Gorry and Morton, 1971) to "Any system that makes some contribution to decision-making" (Sprague and Watson, 1986).

A decision is a choice between alternatives. The *alternatives* may represent different courses of action or different hypotheses. Rational human behaviour involves

the evaluation of choice alternatives based on some criteria. A criterion is some basis for a decision that can be measured and evaluated. It is the evidence upon which a decision is based. Criteria can be of two kinds: 1) constraints, which serve to limit the alternatives under consideration; and 2) factors that enhance or detract from the suitability of specific alternative for the activity under consideration.

A decision rule is the procedure by which criteria are combined to arrive at a particular decision and by which evaluated data can be compared and acted upon. Decision rules are structured in the context of a specific objective, for example, to determine which area is suitable for a given activity. To meet a specific objective, it is frequently the case that several criteria will need to be evaluated (Multi-Criteria Evaluation).

The decision-support field is the "Development of approaches for applying information systems technology to increase the effectiveness of decision-makers in situations where the computer can support and enhance human judgement in the performance of tasks that have elements which cannot be specified in advance" (Sol, 1983).

2.2 Decision Process

According to Sol (1983), the decision-support field is a development of approaches for applying information systems technology to increase the effectiveness of decision-making, in particular in situations where PC-aided support can enhance human judgement in the performance of tasks that have elements, which cannot be specified in advance.

Decision-support systems must provide integration of information and feedback loops to support investigation in the quest for scientific discovery. The intangible factors in the decision-making process may be accounted for through information supplied and choices made by a decision-maker who operates the SDSS interactively or through an analyst. The above suggest that spatial decision-support systems may be developed as general-purpose tools for decision-making (Goodchild and Densham, 1990).

2.3 Spatial Decision-Support Systems (SDSS)

The SDSS have been extensively and adequately covered in literature (Craig and David, 1991, Densham, 1991, Goodchild and Densham, 1990, Moon, 1992, NCGIA, 1992). According to Densham (1991) and Geoffrion (1983), Decision-support Systems has six characteristics:

- Explicit design to solve problems;
- Powerful and easy-to-handle user interface;
- Ability to flexibly combine analytical models with data;
- Ability to explore the space analysis solution by building alternatives;

- **D** Capability of supporting a variety of decision-making styles; and
- **D** Allowing interactive and recursive problem solving.

The distinguishing capabilities and functions of SDSS are to provide mechanisms for the input of spatial data:

- Allow representation of the spatial relations and structures;
- Include the analytical techniques of spatial and geographical analysis;
- **D** Provide output in a variety of spatial forms, including maps.

Notwithstanding in the specialized literature (Holsapple and Whinston, 1991) DSS is mainly viewed as a mathematical technique or a set of techniques for decision-making by optimising something under some specific constraints, we consider SDSS in its broad meaning as an information system that can be used to support decisions at spatial level. By SDSS we mean the integration of all the methods and tools that can be useful to build up a decision-support system for spatially related problems. The system needs the following components: GIS, data analysis and image processing, modelling and expert systems, simulation and optimisation, multi-criteria decision analysis and a suitable user interface. These components are discussed mainly following Fedra and Feoli (1998). The relationships between them are explained in Figure 1.

The ultimate objective of a computer-based SDSS for coastal zone resources management is, or should be, to improve planning and decision-making processes by providing useful and scientifically sound information to the actors involved in these processes, including public officials, planners and scientists, and the general public.

3. The Components of SDSS

3.1 Geographic Information System (GIS)

A Geographic Information System (GIS) is designed as a computer tool to efficiently capture, store, update, manipulate, analyse, and display all forms of geographically referenced information (e.g., IDRISI, 1997). A GIS typically links data from different sets, using geo-referencing, i.e., spatial co-ordinates, as a common key between the data set. The power of a GIS stems from its ability to combine many data sets and display them in a common framework as thematic maps. To obtain thematic maps the GIS may have internal software tools for database and data analysis and image processing, or may be easily interfaced with external software. There are many GIS on the market, commercial and public domain GIS and they are all different each other. They may answer more or less easily and completely to the following typical questions depending on how they have been designed:

- Spatial identification: find locations where certain conditions are satisfied, including the logical (Boolean) or arithmetic combination of multiple layers (overlay analysis);
- Trends: find the differences within an area over time or along spatial dimensions;
- Patterns: find correspondences between distributions of different attributes; calculate similarity between patches, fractal index, indices of fragmentation, and so on;
- Description: Modelling. answer "What if" questions.

However, to answer more complex questions one need to combine (static) geographic data layers with dynamic and spatially distributed models. A progress toward this integration may be achieved by the object-oriented approach (Malczewski 1999). This can be considered as an alternative or a complement of the layer databases approach common in most of the GIS.

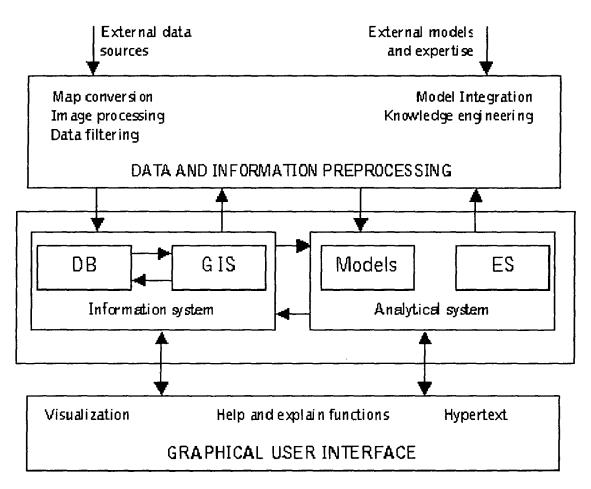


Figure 1: Graphical use interface

3.2 Data Analysis and Image Processing

The maps obtained by GIS may be seen not only as cartographic representations of a classification of the landscape at the end of an analytical process, but mainly as data sources for the landscape spatial pattern analysis through the many different indices of the landscape structure such as shape, fragmentation, fractal, diversity, etc. (Turner, 1989). Many GIS have internal data analysis and image processing systems that can calculate different pattern indices. Some GIS such as IDRISI, ILWIS, GRASS (see Malczewski, 1999 for a comparison between different GIS) have the possibility to treat remote sensing data (image processing) coming from LANDSAT, SPOT, NOAA, etc.; however there are not GIS including data analytical and statistical techniques that are able to classify specific Operational Geographic Units (OGU). The classification may be obtained by applying the clustering algorithms or other multivariate techniques. Feoli and Zuccarello (1996) treat this aspect. GIS can manage different OGUs to obtain maps. In GISs, raster or vector, the OGUs are stored in databasess generally structured according to the relational model. In the databases different characters may describe the OGUs. The probabilistic similarity functions proposed by Goodall (1964) that have been suggested and that have been included in the information system developed in ICS/UNIDO (Feoli 1993) would improve the interpretation by introducing the probabilistic aspect in the prediction. The functions can deal with 6 categories of characters: 1) binary; 2) qualitative, with more than two alternative values; 3) ordinal; 4) only small integer values; 5) quantitative grouped into classes; 6) quantitative not grouped into classes.

3.3 Modelling and Expert Systems

In GIS, the basic concept is one of location, of spatial distribution and relationship; basic elements are spatial objects. In environmental modelling, by contrast, the basic concept is one of state, expressed in terms of numbers, mass, or energy, of interaction and dynamics; the basic elements are "species", which may be biological, chemical and environmental media such as air, water or sediment. The overlap and relationship is apparent, and thus the integration of these two fields of research, technologies, or sets of methods, that is, their paradigms, is an obvious and promising idea (Fedra, 1995).

Simulation and *optimisation modelling* is a powerful tool of analysis, and forecasting. Modelling of marine systems has a considerable tradition in both physical and biological oceanography. For summary treatments, more recent model comparisons, and some classics in the field see, for example, Riley et al., (1949); Nihoul, (1975); Kremer and Nixon, (1978); Falconer et al., (1989); Fransz et al., (1991).

However, most of these approaches are designed to improve scientific understanding of physical and biological processes, with the notable exception of some of the fisheries models that have obvious management implications. Other management oriented and model-based studies of marine systems include the RAND Corporation classic on the Oosterschelde or the recent study of the North Sea Management Analysis North Sea (MANS), (Klomp, 1990). Other important recent models for coastal zone management are ECOPATH, ECOSIM and ECOSPACE (Pauly et al., 2000) that are offering a set of tools for evaluating ecosystem impact of fisheries based on modelling the mass balance fluxes in the fish food-webs both in homogeneous and heterogeneous space conditions.

Quantitative numerical models and rule-based, qualitative expert systems provide another example of possible integration. *Expert systems* can be used just as any other model to assign a value to an output variable given a set of input variables; they do this, however, by using rules and logical inference rather than numerical algorithms.

In the context of models, expert systems are often used to help configure models (implementing an experienced modellers know-how to support the less experienced user) and estimate parameters. A number of these "intelligent front end systems" or model advisors have been developed in the environmental domain (Fedra, 1992).

A rule-based approach can also be a substitute for a numerical model, in particular if the processes described are not only in the physical and chemical, but also in the biological and socio-economic domain. An example could be the environmental impact assessment based on a checklist of problems, which can be understood as a diagnostic or classification task. A qualitative label is assigned to potential problems, based on the available data on environment and planned action, and a set of generic rules assessing and grading the likely consequences. And finally, a model can be integrated into the inference chain of an expert system (Fedra, 1992). Recent examples of environmental expert systems are given in Hushon (1990) and Wright et al., (1993).

The flexibility to use, alternatively or conjunctively, both symbolic and numerical methods in one and the same application allows the system to be responsive to the information at hand, and the user's requirements and constraints. This combination and possible substitution of methods of analysis, and the integration of databases, geographical information systems, and hypertext, allows to efficiently exploit whatever information, data and expertise is available in a given problem situation.

The approach is based on a model of human problem-solving that recursively refines and redefines a problem as more information becomes available or certain alternatives are excluded at a screening level. Learning, i.e., adaptive response to the problem situation and the information available, and the ability to modify function and behaviour, as more information becomes available, is a characteristic of intelligent systems.

3.4 Multi-Criteria Evaluation

To meet a specific objective, it is frequently the case that several criteria will need to be evaluated. Such procedures are called Multi-Criteria Evaluations (MCE) (Eastman et al., 1995). Two of the most common procedures for multi-criteria evaluation are weighted linear combination and concordance-discordance analysis (Carver, 1991). Weighted linear combination is very straightforward in a raster GIS. Indeed, it is the derivation of the weights, within the context of the decision objective that provides the major challenge. The weights are developed by providing a series of PAIRWISE comparisons of the relative importance of factors to the suitability of pixels for the activity being evaluated (Eastman, 1997). Figure 2 presents a suitability map for industrial site planning and urban settlement in the Tunisian coastal areas.

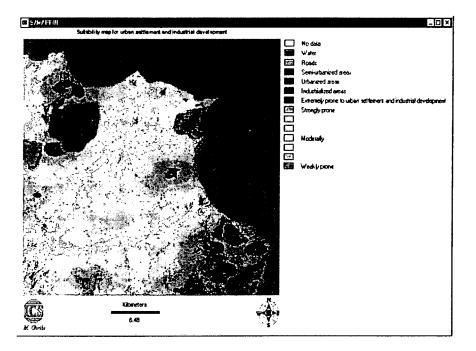


Figure 2: Suitability map

In a Multi-Criteria Evaluation, an attempt is made to combine a set of criteria to achieve a single composite basis for a decision according to a specific objective (Eastman et al., 1995). During the last two decades, further support has emerged for the view that a decision is a multidimensional concept (Nijkamp et al., 1990; Paruccini, 1994). Decisions about the allocation of land typically involve the evaluation of multiple criteria according to several often conflicting-objectives (Eastman et al., 1995). Making-decisions about the allocation of land is one of the most fundamental activities of resource development. With the development of GIS, we now have the opportunity for a more explicitly reasoned process of land use evaluation. However, despite the wide range of analytical tools these systems provide, they are typically weak in the provision of decision-support procedures (Honea et al., 1991).

The advantage of MCE is that it provides a flexible way of dealing with qualitative multi-dimensional environmental effects of decisions (Munda, 1995). Even in the absence of monetary information, the point of departure of any multi-criteria analysis is the generation of a discrete set of alternatives, the formulation of a set of criteria, and the evaluation of an impact of each criterion for every alternative. Janssen (1992) and Munda (1995) provide a comprehensive review of multi-criteria decision-support for environmental management.

Although a variety of techniques exist for the development of weights, one of the most promising would appear to be that of PAIRWISE comparisons, developed by Saaty (1980), in the context of a decision-making process known as the Analytical Hierarchy Process (AHP). In the PAIRWISE comparison method the decision-maker is asked to select the most important of each possible pair effects (Janssen and Van Herwijen, 1992). Subsequently, a comparison must be established in qualitative terms to what extent one effect is more important than the other one (Saaty, 1980) to express the differences of importance.

3.5 Multi-Objective Evaluation

While many decisions we make are prompted by a single objective, it also happens that we need to make decisions that satisfy several objectives. A multi-objective problem is encountered whenever we have two candidate sets (i.e., sets of entries) that share members. These objectives may be complementary or conflicting in nature (Carver, 1991:332).

a) Complementary Objectives

With complementary or non-conflicting objectives, land areas may satisfy more than one objective, i.e., an individual can belong to more than one decision set. Desirable areas will thus be those, which serve these objectives together in some specified manner. For example, we might wish to allocate a certain amount of land for combined recreation and wildlife preservation uses. Optimal areas would thus be those that satisfy both of these objectives to the maximum degree possible.

b) Conflicting Objectives

With conflicting objectives, the objectives compete for the available land since it can be used for one or the other, but not for both. For example, we may need to resolve the problem of allocating land for timber harvesting and wildlife preservation. Clearly the two cannot coexist. Exactly how they compete and on what basis one will win out over the other will depend upon the nature of the decision rule that is developed.

In case of complementary objectives, multi-objective decisions can often be solved through a hierarchical extension of the multi-criteria evaluation process. For example, we might assign a weight to each of the objectives and use these along with the suitability maps developed for each to combine them into a single suitability map indicating the degree to which areas meet all of the objectives considered (see Voogd, 1983). However, with conflicting objectives the procedure is more involved.

With conflicting objectives, it is sometimes possible to rank order to objectives and reach a *prioritised* solution (Rosenthal, 1985). In these cases, the needs of higher ranked objectives are satisfied before those of lower ranked objectives are dealt with. However, this is often not possible, and the most common solution to conflicting objectives is the development of a *compromise* solution. Undoubtedly, the most commonly employed techniques for resolving conflicting objectives are those involving optimisation of a choice function such as mathematical programming (Fiering, 1986) or goal programming. In both techniques, the concern is to develop an allocation of the land that maximises or minimises an objective function subject to a series of constraints.

3.6 The User Interface

Analysis in GIS is based on the map paradigm; maps in inherently static objects, and the basic level of analysis are (Boolean) map overlays. Complex relationships, and dynamic processes, may require another layer of tools, namely simulation models, to be integrated with a GIS.

The integration of GIS and environmental models can come in many forms. In the simplest case, two separate systems, the GIS and the model, just exchange files: the model obtains some of its input data from the GIS, and produces some of its output in a format that allows import and further processing and display with the GIS. This seems to be a rather common approach, since it requires little if any software modifications. Only the file formats and the corresponding input and output routines, usually of the model, have to be adapted. Depending on the implementation however, a solution based on files shared between two separate applications, usually with a different user interface, is cumbersome and possibly error prone if it involves a significant amount of manual tasks. Deeper integration provides a common interface and transparent file or information sharing and transfer between the respective components. One possible way is the use of higher level application language or application generators increasingly common as built-in features of commercial GIS packages as the basis of numerous integrated applications. Application generators and modelling capabilities with commercial GIS also offer the possibility of tight integration within the limits of the respective package options.

An alternative is to use an open GIS tool kit that uses a standardised interface, such as GRASS. Modules of the overall GIS system (which really is a set of tools with a standardised pipeline-type flexible coupling) can be included in modelling applications. The X Windows system and a number of interface-building toolkits make this a rather efficient integration strategy. Any integration at this level, however, requires a sufficiently open GIS architecture, that provides the interface and linkages necessary for tight coupling.

Using such predefined tools and components, while efficient for fast prototyping, can also be restrictive: investments in toolkits always carry the temptation to reformulate problems in terms of the available tools rather than the other way round. Another alternative is the use of do-it-yourself tool kits that provide both customised GIS functionality as well as interface components for simulation models. A relatively recent example of integration that draws together GIS, models, spreadsheet, and expert systems in a programmable system is RAISON (Lam and Swayne, 1991).

For having an efficient user interface we need:

- An interactive, menu-driven user interface that guides the user with prompts and explains messages through the application. No command language or special format of interaction is necessary, the computer assists the user in its proper use; help and explain functions can be based on hypertext and possibly include multi-media methods to add video and audio technology to provide tutorial and background information; in other words, they must be easy to use even for the non-specialist.
- Dynamic colour graphics for the model output and a symbolic representation of major problem components that allow easy and immediate understanding of basic patterns and relationships. Rather than emphasising the numerical results, symbolic representations and the visualisation of complex patterns support an intuitive understanding of complex systems behavior; the goal is to translate a model's state variables and outputs into the information requirements of the decision-making process; the coupling to one or several databasess, including geographical information systems, and distributed or remote sources of information in local or wide area networks, that provide necessary input information to the models and the user. The user's choice or definition of a specific scenario can be expressed in an aggregated and symbolic, problem-oriented manner without concern for the technical details of the computer implementation; data on chemical properties of the substances of concern, used in pollution fate and transport models, are one obvious example;
- Embedded Artificial Intelligence (AI) components, such as specific knowledge bases allow user specifications in allowable ranges to be checked and constrained, and ensure the consistency of an interactively defined scenario; as a special case this could also include QSAR methods that can be implemented in a combination of rule-based methods and algorithmic methods;

Interaction is a central feature of any effective man-machine system: a real-time dialogue, including explanation, allows the user to define and explore a problem incrementally in response to immediate answers from the system. Fast and powerful systems with modern processor technology can offer the possibility to simulate dynamic processes with animated output, and they can provide a high degree of responsiveness that is essential to maintain a successful dialogue and direct control over the software.

Visualisation provides the band-width necessary to communicate and understand large amounts of highly structured information, and permits the development of an intuitive understanding of processes and interdependencies, of spatial and temporal patterns, and complex systems in general. Also, many of the problem components in a real-world planning or management situation, such as risk or reliability, are rather abstract: a graphical representation of such concepts makes them tangible objects that can literally be manipulated and understood intuitively.

Intelligence requires software to be knowledgeable not only about its own possibilities and constraints, but also about the application domain and about the user, i.e., the context of its use. Defaults and predefined options in a menu system, sensitivity to context and history of use, built-in estimation methods, learning, or alternative ways of problem specification depending on the user can all be achieved by the integration of expert systems technology in the user interface and in the system itself.

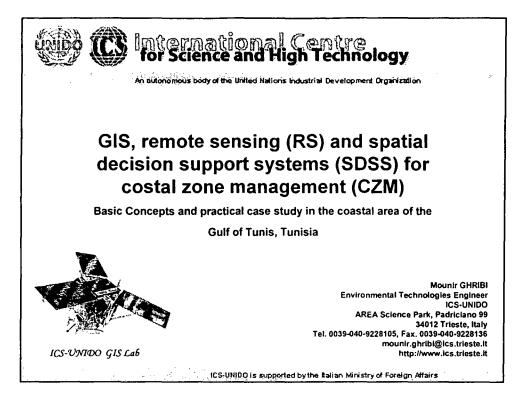
Customisation is based on the direct involvement of the end-user, and the consideration of institutional context and the specifics of the problem domain in systems design and development. It is the users view of the problem and their experience in many aspects of the management and decision-making process that the system is designed to support. This then must be central to a system's implementation to provide the basis for user acceptance and efficient use.

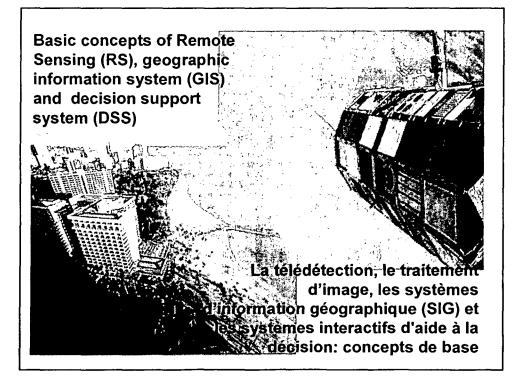
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Remote Sensing (Télédétection):

RS is the science (set of knowledge and techniques) of acquiring information about the earth's surface and its environment without actually being in direct contact with it.

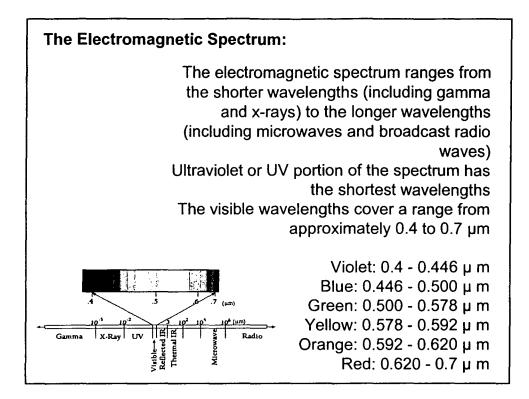
This refers to a process of gathering data by sensing and recording electromagnetic energy, usually by aircraft and satellite sensors.

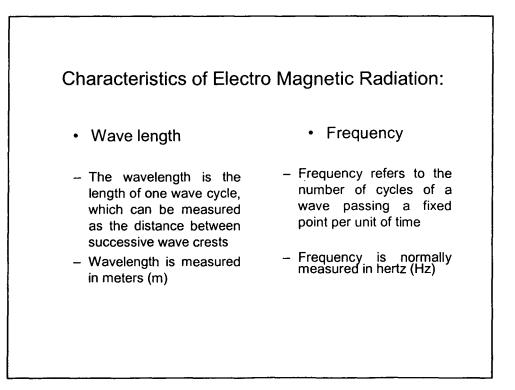
Examples range from aerial photography to multi-spectral scanning and radar.

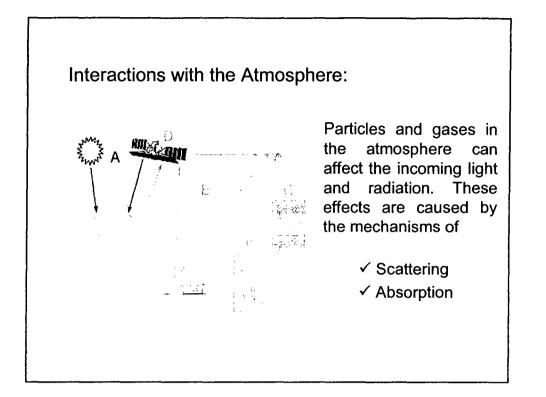
Télédétection

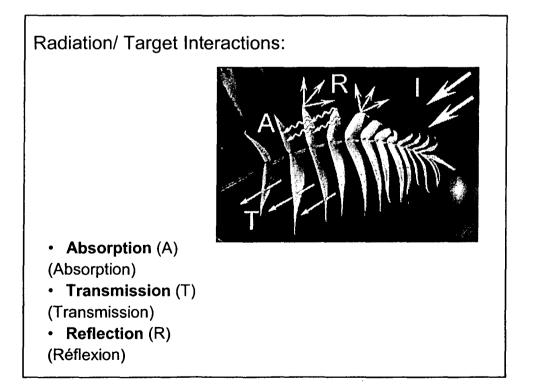
La télédétection est la détection d'un objet à distance. Ce terme correspond plus spécifiquement à une technique qui utilise les ondes électromagnétiques (rayonnement) pour obtenir des informations concernant la Terre et l'atmosphère.

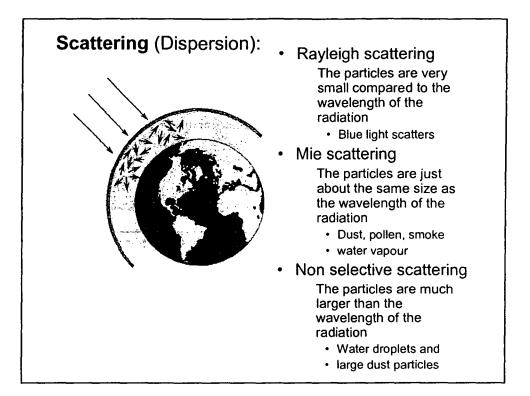
Le spectre électromagnétique comporte plusieurs parties distinctes, dont le rayonnement ultraviolet (UV), la lumière visible, le rayonnement infrarouge (IR) thermique, les rayons X, les micro-ondes, etc...

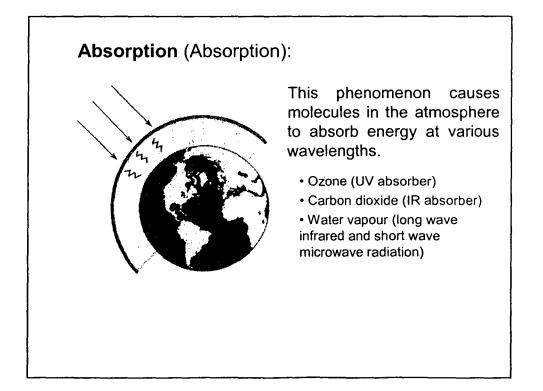


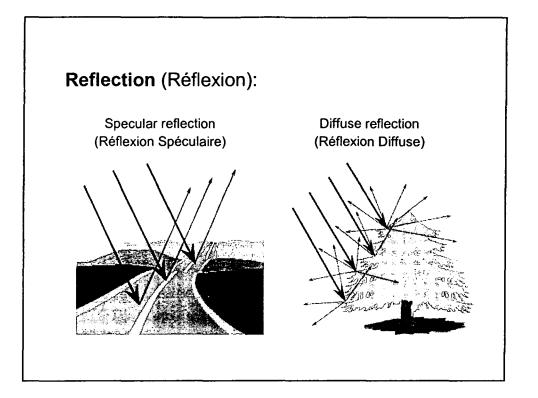


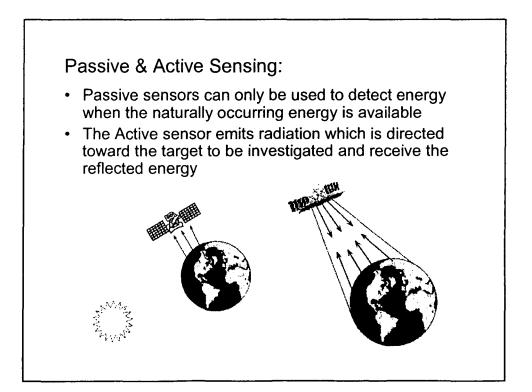


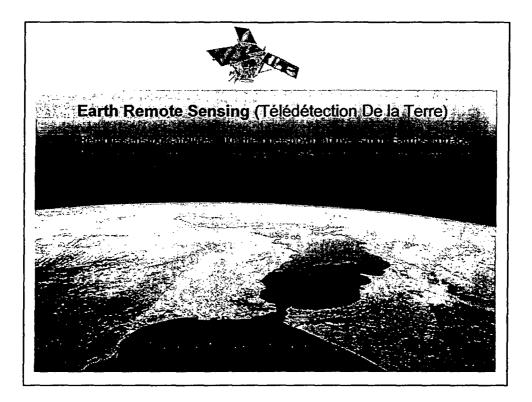




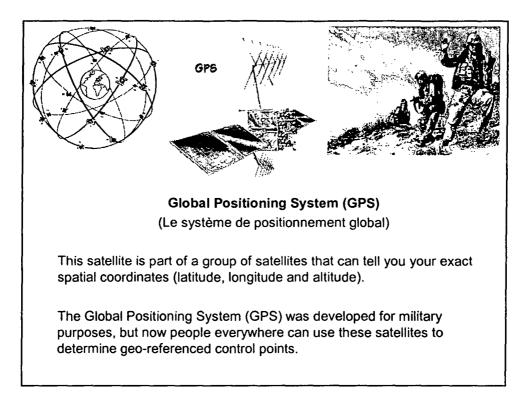


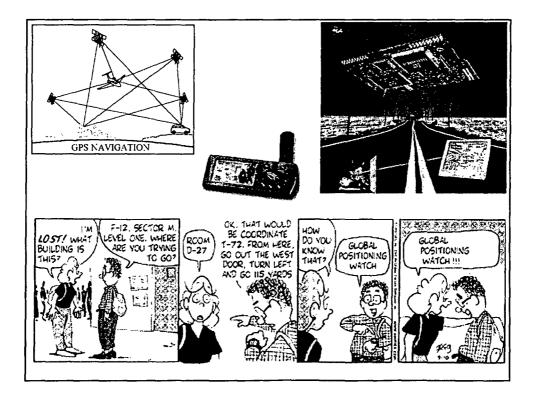


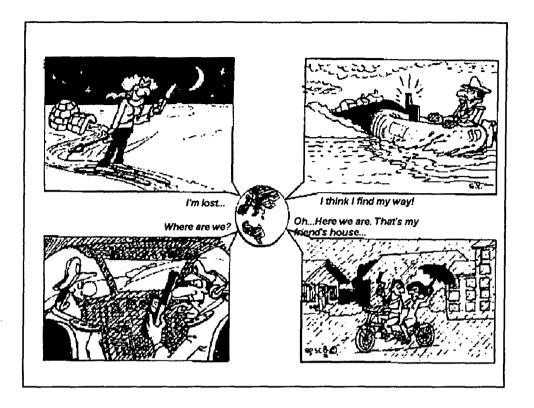




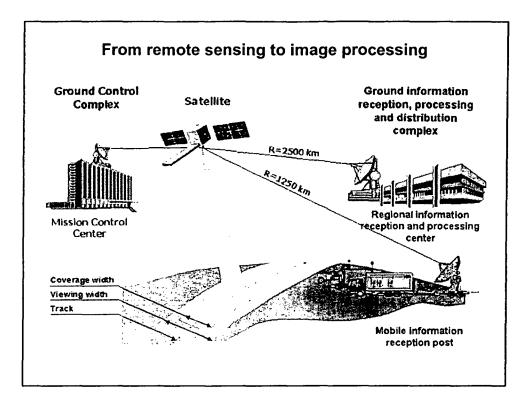
How we use Satellite Imaging? (Comment utilisons-nous l'information obtenue?) The instruments on remote-sensing satellites study earth's plant cover and surface water, among many other features. People who work in farming, fishing, mining, and many other activities find this information very useful. We can also use remote sensing satellites to study changes in the earth's surface that are caused by people.

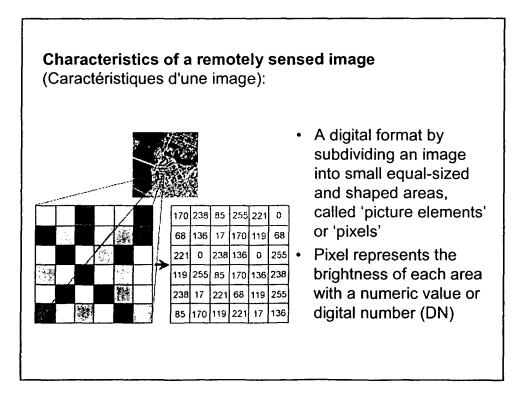


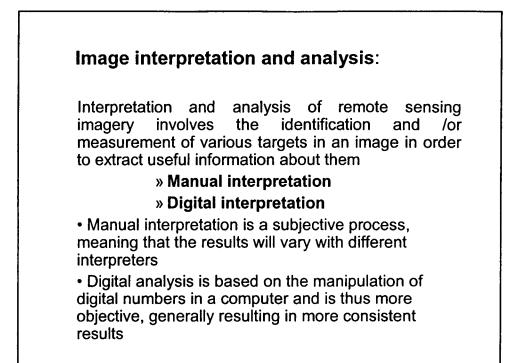


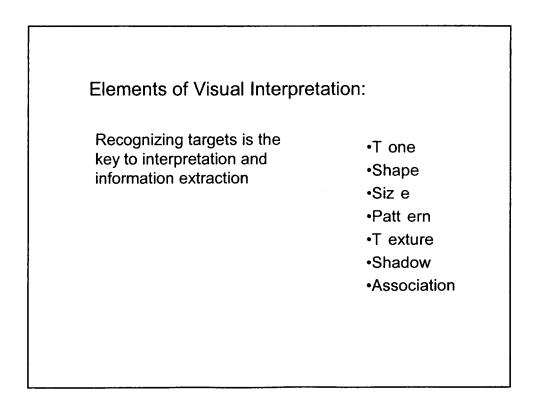


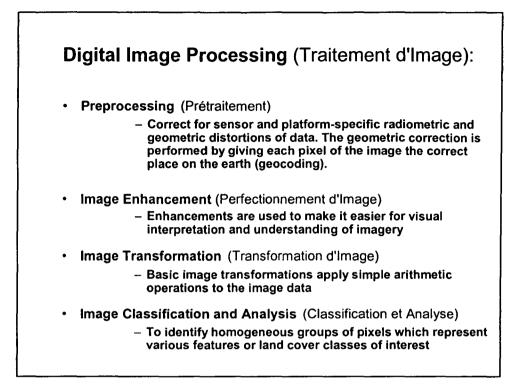
Links to GPS Sites on the WWW
There is plenty of additional information about the GPS on the WWW. In fact, you may easily lose yourself among the many sites which cater to a broad public at many different levels of complexity. Before you start exploring on your own hand, you may like to have a look at the sites below. Most of them have very useful onward links: <u>http://www.utexas.edu/depts/grg/gcraft/notes/gps/gps.html</u> - Global Positioning System Overview by Peter H. Dana (Department of Geography, University of Texas at Austin, USA) - Selected as the "Best GPS Site", one of GIS World's "Best of the Net" sites for 1996. Here you will find Map
Projection Overview, Geodetic Datum Overview, Coordinate Systems
Overview and Links to other related information.
http://galaxy.einet.net/editors/john-beadles/introgps.htm - Introduction to the NAVSTAR and GLONASS Global Positioning Systems (GPS), Global
Navigation Satellite Systems (GNSS) and their uses.
gopher://unbmvs1.csd.unb.ca:70/hPUB.CANSPACE.GPS.INTERNET.SERVIC
ES.HTML - Navstar GPS Internet Connections by Richard B. Langley
http://www.navsys.com/gps.htm - History of GPS - GPS Description- Links to
other GPS Web Sites
http://www.ion.org/redbookstoc.html - The "Red Books" about the Global
Positioning System: Papers Published in NAVIGATION

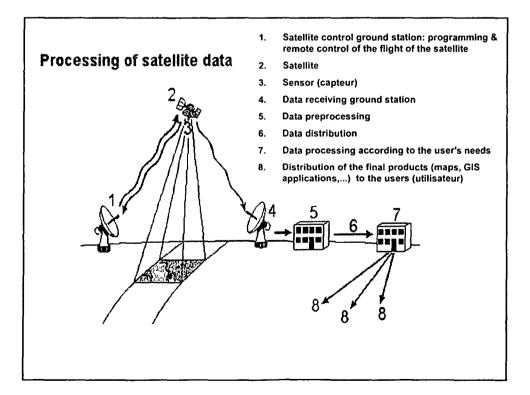


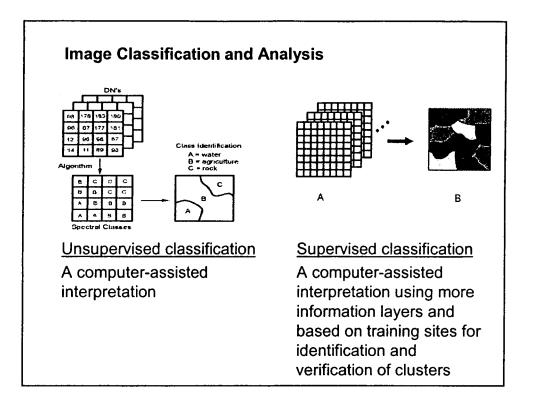


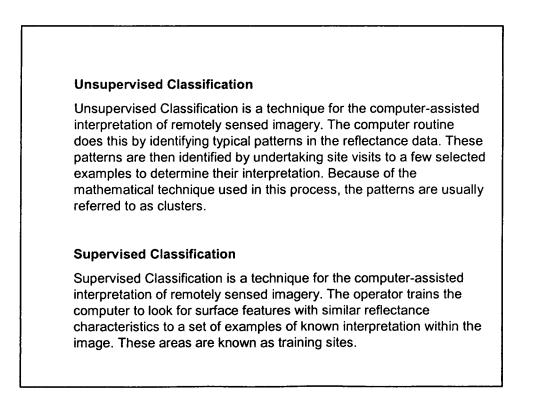


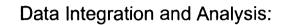






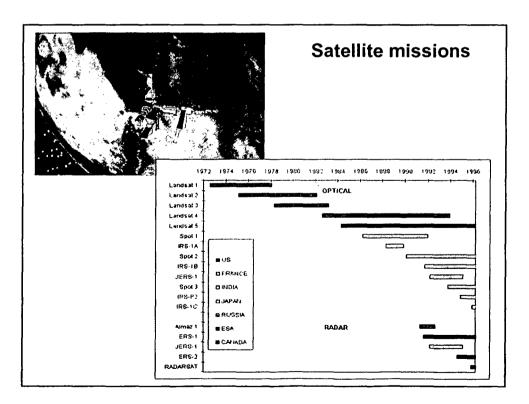


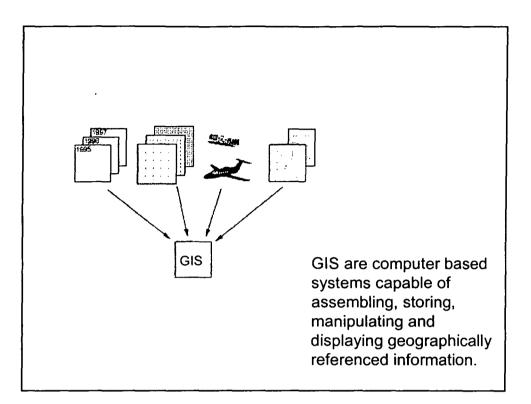


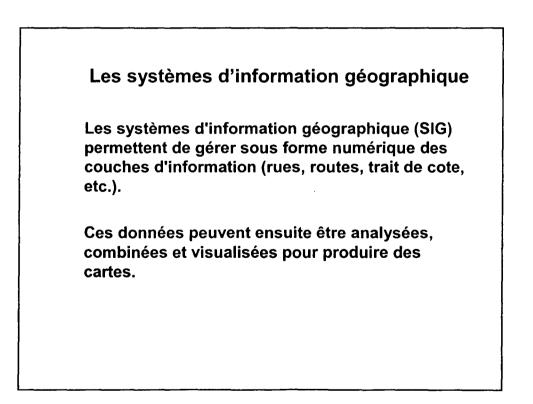


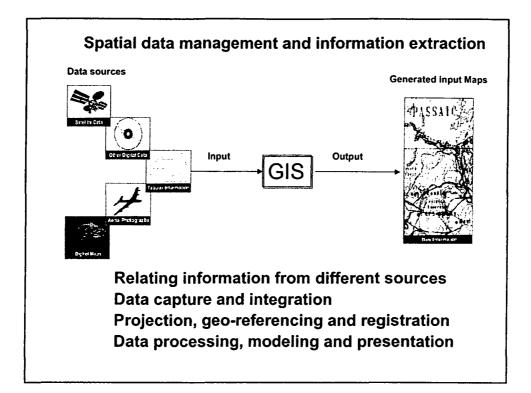
Data integration fundamentally involves the combining or merging of data from multiple sources in an effort to extract better and/or more information. This may include data that are:

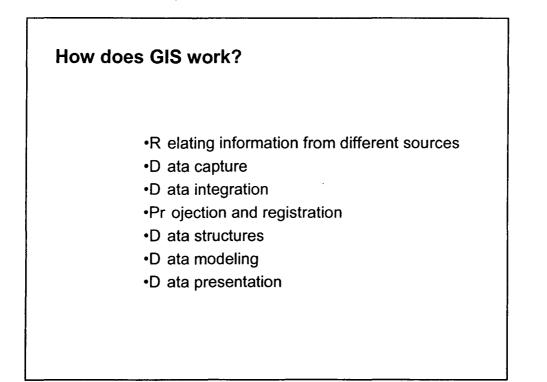
- -multi-temporal
- -multi-resolution
- -multi-sensor
- -multi-data type in nature

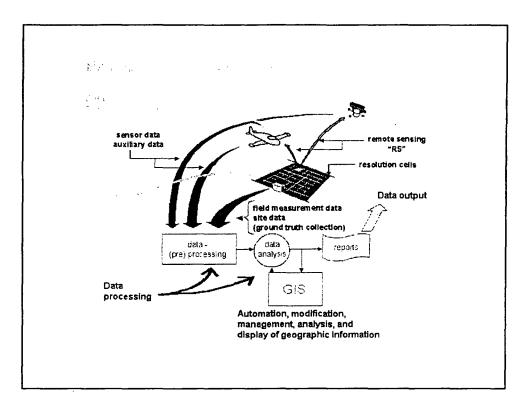


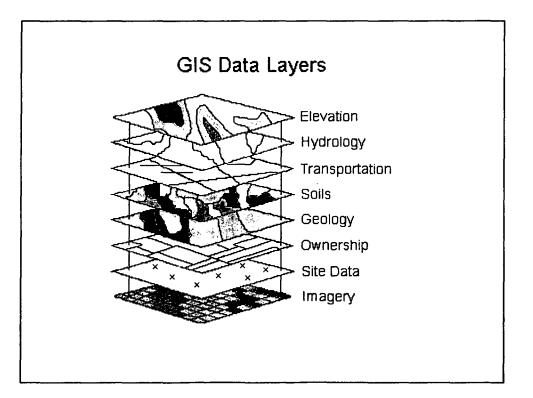












Map Projection



Projection refers to the representation of one surface onto another.

In the mapping sciences it is understood to refer to the representation of a **spherical** earth onto a **flat** medium such as paper or a computer screen.

Since it is physically impossible to flatten a globe without distortion, scale will vary across the projection surface with consequent distortions in **distance**, **area**, **and angular** relationships.

Geo-referencing

Geo-referencing refers to the location of an image or vector file in space as defined by a known coordinate referencing system.

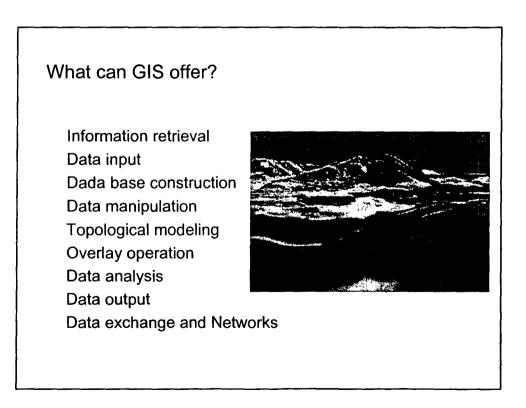
With raster images, a common form of geo-referencing is to indicate the reference system (e.g., latitude/longitude), the reference units (e.g., degrees), and the coordinate positions of the left, right, top and bottom edges of the image.

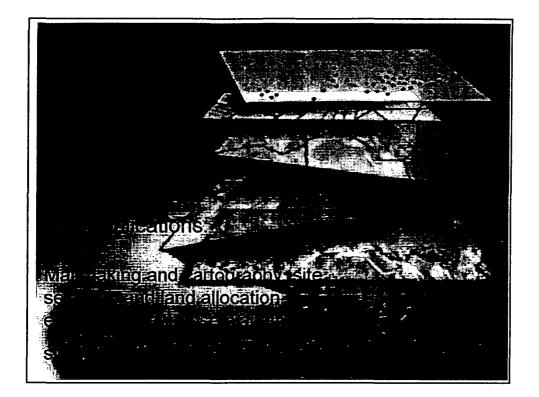
The same is true of vector data files, although the left, right, top and bottom edges now refer to what is commonly called the bounding rectangle of the coverage -- a rectangle which defines the limits of the study area.

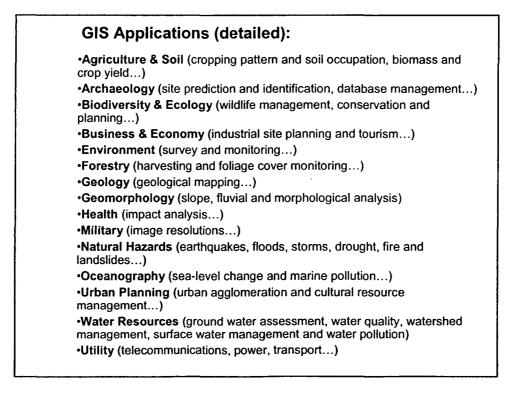
Some examples of reference system

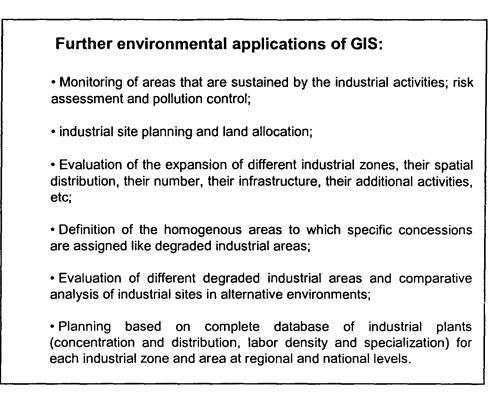
These include the geodetic coordinates **LATLONG** (latitude/longitude) using the decimal degrees as unit of measure, the **UTM** system (Universal Transverse Mercator) and SPC (State Plane Coordinate) systems based on the Lambert Conformal Conic and Transverse Mercator projections.

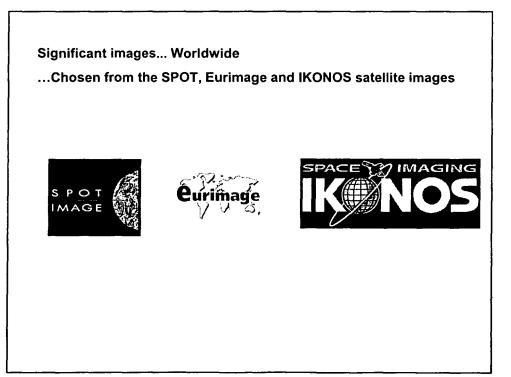
The UTM grid referencing system is frequently used for map projection. The unit of measure is meters or kilometres (m or km). It is usually necessary to know the UTM zone and datum used by the reference system.

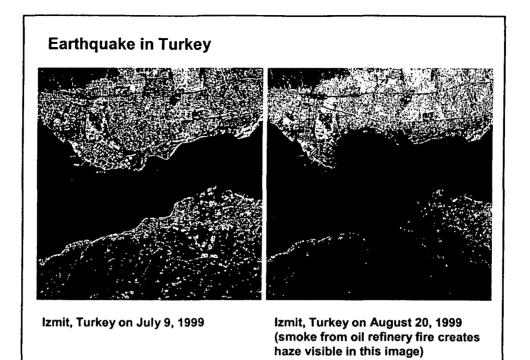


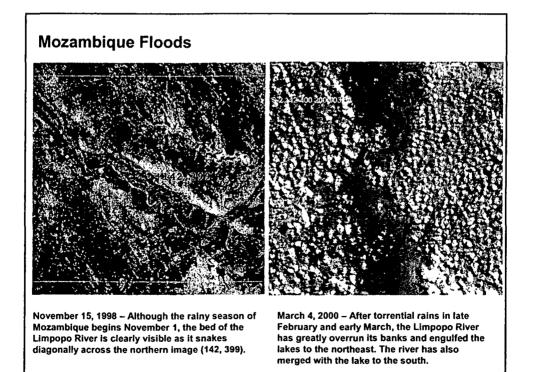


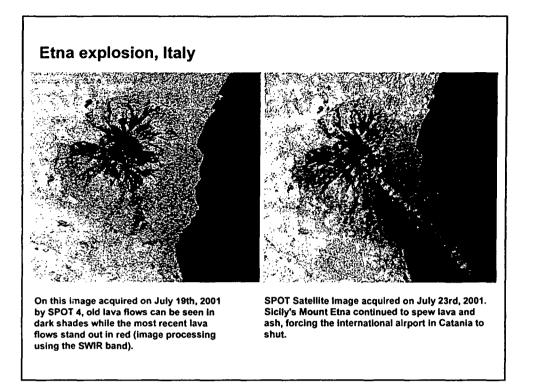






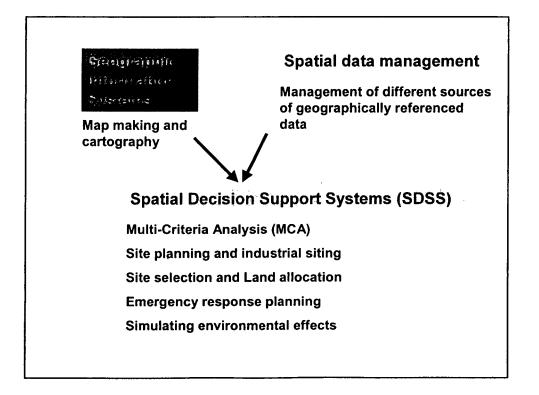












Decision Support Systems (DSS) (systèmes interactifs d'aide à la décision)

• A decision is a choice between alternatives.

• The alternatives may represent different courses of action or different hypotheses.

• Rational human behaviour involves the evaluation of choice alternatives based on some criteria.

• A criterion is some basis for a decision that can be measured and evaluated.

Criteria can be of two kinds:

• Factors that enhance or detract from the suitability of specific alternative for the activity under consideration

• Constraints which serve to limit the alternatives under consideration

What is DSS?

Definitions of decision support systems range

from:

"Interactive computer based systems that decision makers utilize data and models to solve unstructured problems" (Gorry & Morton, 1971)

to

"Any system that makes some contribution to decision making" (Sprague & Watson, 1986)

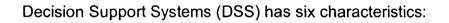
'The decision support field is the development of approaches for applying information systems technology to increase the effectiveness of decision makers in situations where the computer can support and enhance human judgment in the performance of tasks that have elements which cannot be specified in advance'

Multi-Criteria Evaluation:

• To meet a specific objective, it is frequently the case that several criteria will need to be evaluated. Such procedure is called Multi-Criteria Evaluation (MCE)

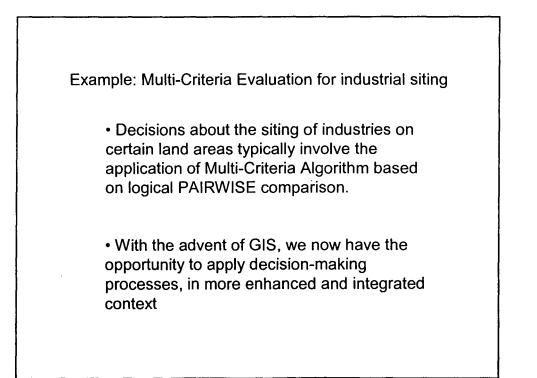
• In a MCE, an attempt is made to combine a set of criteria to achieve a single composite basis for a decision according to a specific objective

• The advantage of MCE is that it provides a flexible way of dealing with qualitative multi-dimensional environmental effects of decisions



- · Explicit design to solve problems;
- · powerful and easy-to-handle user interface;
- · ability to flexibly combine analytical models with data;
- ability to explore the space analysis solution by building alternatives;
- capability of supporting a variety of decision-making styles; and
- allowing interactive and recursive problem-solving

The distinguishing capabilities and functions of Spatial Decision Support Systems (SDSS) are to: • provide mechanisms for the input of spatial data ; • allow representation of the spatial relations and structures ; • include the analytical techniques of spatial and geographical analysis ; and • provide output in a variety of spatial forms, including maps



Mise en place d'un système d'information géographique (SIG)

Objectif

L'objectif est de donner aux ingénieurs, techniciens et cadres en charge, les bases nécessaires à:

- la saisie des données localisées ou géographiques (scannerisation, digitalisation, GPS...);

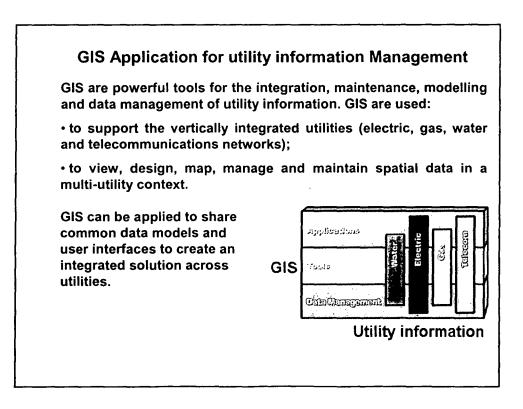
- la structuration de l'information (modèle et équation) ;

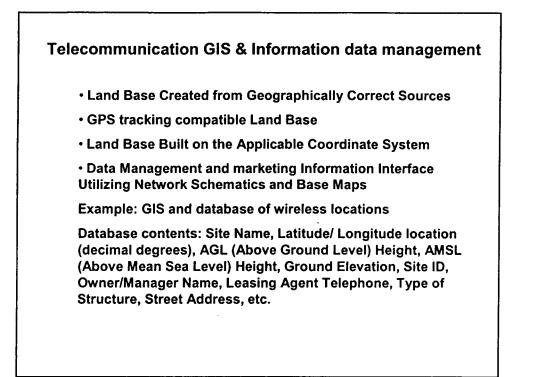
- la mise en œuvre des applications informatiques de gestion de l'information géographique;

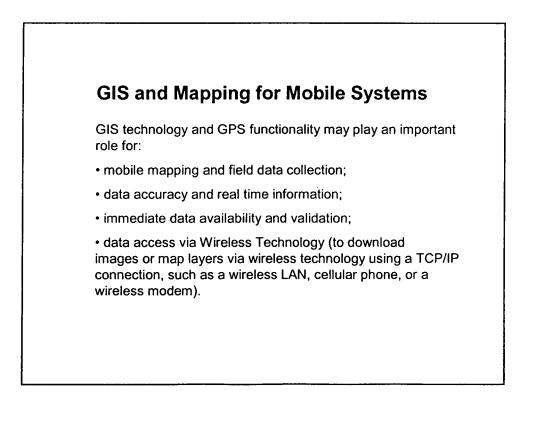
le traitement des données (requêtes, analyse, traitement d'image...);
l'édition des données (cartes, tableaux...)

Procédure

La mise en œuvre d'un système d'information géographique ne saurait se limiter au simple apprentissage de l'outil logiciel qui n'en constitue que l'interface d'accès et de manipulation des données. Il faut replacer le sujet dans une vision globale du système d'information alliant la représentation cartographique à une véritable banque de données.







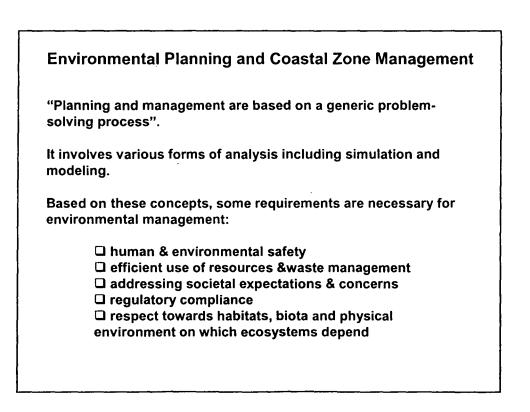
Coastal Zone Management (CZM)

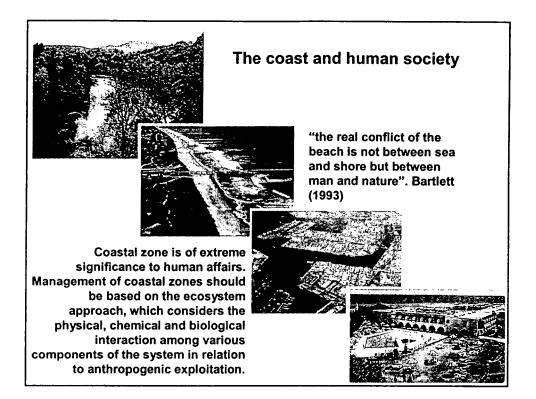


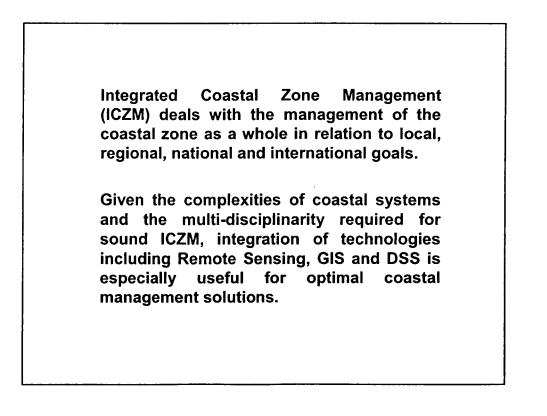
Coastal zone management is the management of coastal resources.

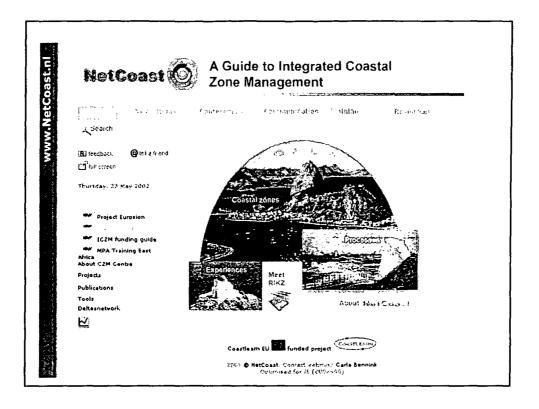
Resources are common to other areas (residential settlement, agriculture, forestry, military base, tourism and recreation, quarrying, dumping and waste disposal, etc.)

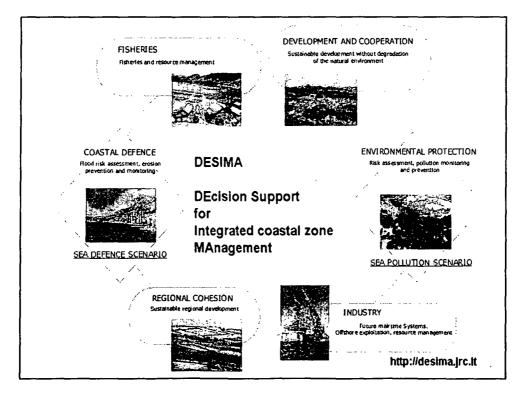
Resources are also specific to coastal area (fishing, navigation, port and harbor, desalination and salt production, etc.)

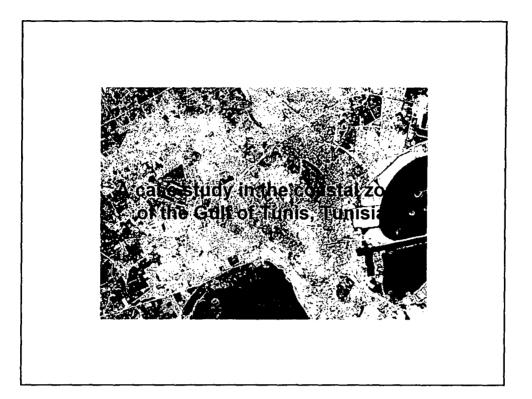












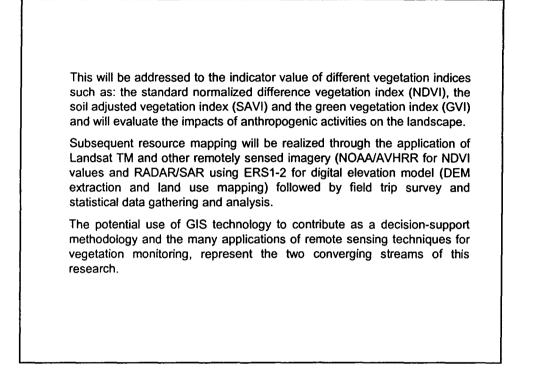
ABSTRACT

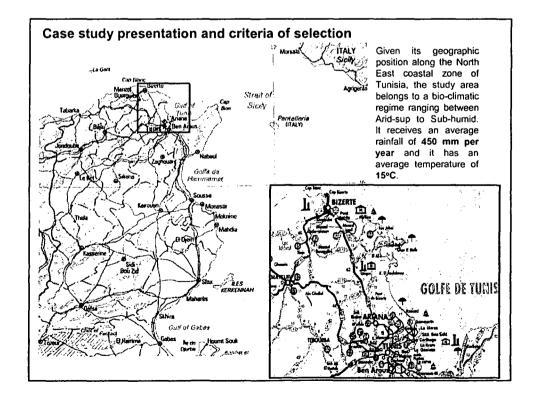
The Tunisian coastal area, as for many other coasted countries of the Mediterranean Basin, is subject to an increasing pressure from a wide variety of activities such as recreational and tourism projects, industrial development and urban settlement.

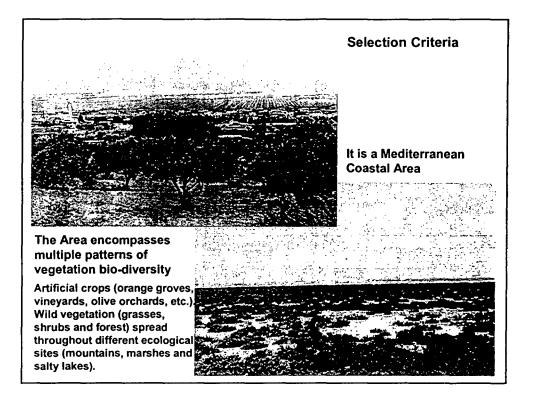
Environmental managers are faced with the challenge of balancing this pressure using an integrated coastal area management (ICAM) approach.

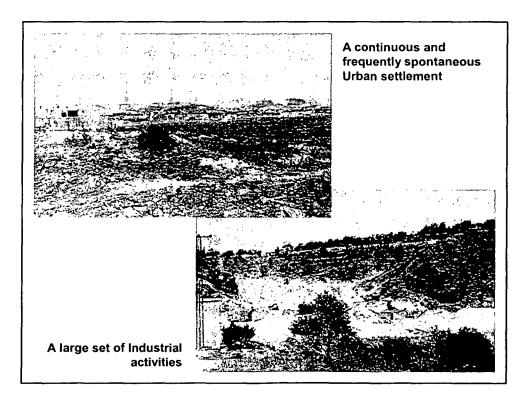
An integrated system adopts the integration of information technologies to incorporate data from a variety of sources and should be built on: the collection and the analysis of socio-economic and climatic data, the use of remote sensing data, the development of models and the use of advanced techniques and tools.

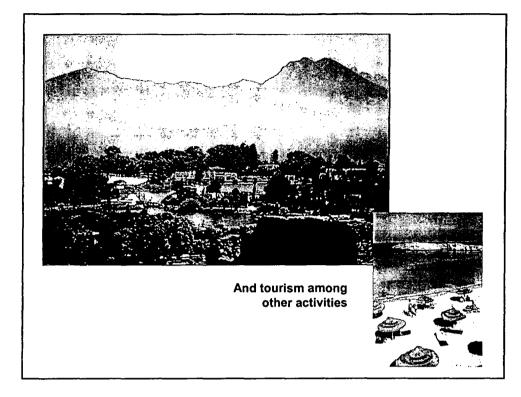
The present research will focus on the importance of those tools including geographic information systems (GIS), remote sensing (RS) and decisionsupport systems (DSS) and will illustrate the many advantages of spatial decision support systems (SDSS) through the application of Multi-Criteria Evaluation (MCE). From conventional approaches to urban mapping and industrial siting taking in considerations the indicator value of the vegetation for monitoring and analysis.











NDVI Data Extraction And Analysis

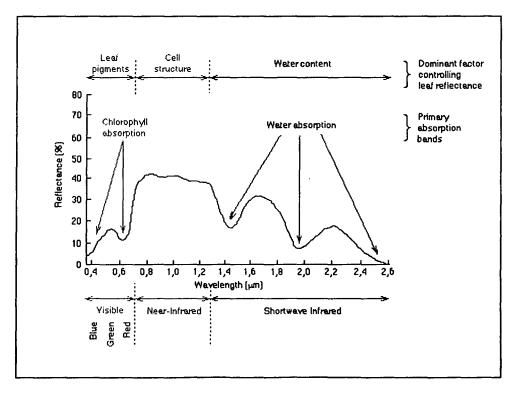
Access to land processes data, including satellite and aircraft acquired data stored in the Earth Resources Observation Systems (EROS), plays an important role to promote the interdisciplinary study and understanding of the integrated Earth system (USGS, 1998). Since 1983, Advanced Very High Resolution Radiometer (AVHRR) of the National Oceanic and Atmospheric Administration (NOAA) satellites has given a continuous spatial cover on regular time scale of the photosynthetic potential activity. This can be expressed by vegetation indices such as the Normalized Difference Vegetation Index (NDVI) (Richard et al., 1998).

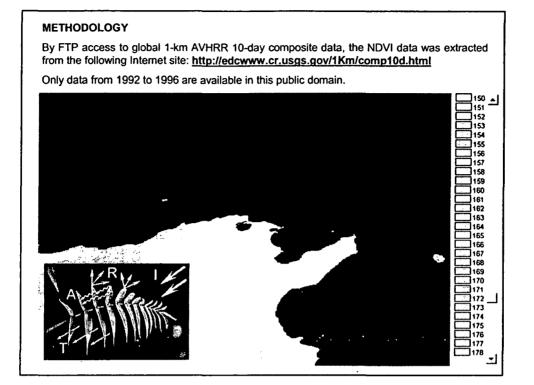
The NDVI, referred to the greenness index, is used as a measure of vegetation health and density. A healthy vegetated area will have a high NDVI value, while a sparsely vegetated or unhealthy area will have a lower value. The values of NDVI will range from -1 (the lowest possible value indicating areas void of vegetation, ice, water, barren ground etc.), through +1 (the highest value indicating highly vegetated, healthy environment). In the real world, useful values fall between 0 and +1. The NDVI values are often rearranged from 0 to 255 in order to make use of the full 8-bit nature of the graphics.

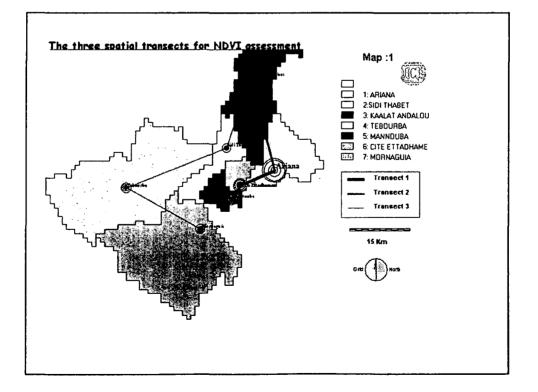
The NDVI is derived from the reflected solar radiation in the near-infrared (NIR) and red (RED) wavelength bands via the algorithm:

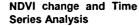
NDVI=(NIR-R)/(NIR+R), Example NDVI=(band4-band3)/(band4+band3) for Landsat TM

The NDVI is a nonlinear function which is undefined when RED and NIR are zero. The NDVI values vary with absorption of red light by plant chlorophyll and the reflection of infrared radiation by water-filled leaf cells.



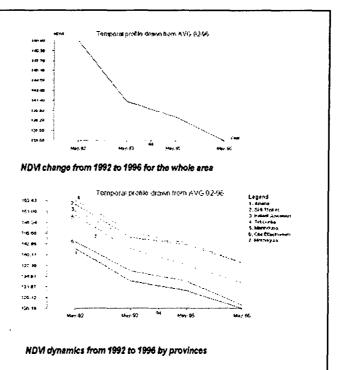






Time series analysis concerns the examination of change over a sequence of ten-days images obtained for the 3 decades of May for 1992, 1993, 1995 and 1996 (data for 1994 was not available).

To measure the global changes over profiles of time, the vegetation dynamics of the whole study area, by each province (délégation) and by main residential sites was measured taking the normalized difference vegetation index (NDVI) as the numerical indicator.



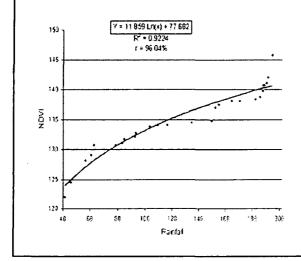
INTERPRETATION

As mentioned earlier, NDVI is a measure of the greenness or vegetation health. High values indicate presence of healthy vegetation while lower values indicate less vegetation or stressed vegetation. The graphs clearly indicate the vegetation change over time characterized by a global decline of NDVI for the whole area and by each province.

There are some justifications to the change of the NDVI values corresponding reasonably well with the season and the rainfall period but also the man effect is a crucial factor affecting the NDVI dynamics. For instance, Ariana and Mannouba represented the lowest NDVI average values (this can be explained by their proximity to the capital), while Tebourba and Sidi Thabet are covered by more green areas thanks to the dominant agriculture vocation.

NDVI sensitivity to rainfall, urbanization, industrialization and elevation

Using the "Idrisi" module "Analysis/ Statistics/ Regress", some simple linear regression models between the NDVI average as a dependent variable and the rainfall, the industrial growth ratio, the urban growth ratio and the altitude as independent variables were carried out successively. To enhance their statistical relationship, some mathematical transformations of the variables were done.



The spatial distribution of the vegetation cover is strongly related climatic to severe conditions (Saugier, 1996). As a vegetation the NDVI can be indicator, considered as a climatic recorder, and mainly a rainfall recorder. In fact, it has been shown that NDVI was highly sensitive to an extended rainfall anomaly, like the 1984 Ethiopian drought (Henricksen, 1986).

The evolution of the amount of rainfall is followed by an increase of NDVI and this is explained by a positive correlation between the two variables (positive slope and a high correlation expressed by a relatively high coefficient (96.04 %).

Correlation NDVI/ Urban growth ratio

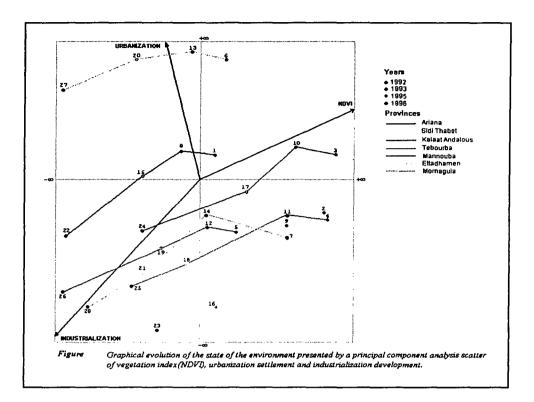
The expansion of urban settlement is followed by a decrease of NDVI and this is explained by a higher negative correlation between the two variables (negative slope -0.021935 and relatively high coefficient of determination r^2 =55.61 %). According to statistical table forms, the model is significant at 1% level.

Correlation NDVI/ Industrial growth ratio

The industrial development is followed by a **decline of NDVI** and this is explained by a higher negative correlation between the two variables (negative slope – 0.02195 and relatively high coefficient of determination r²=55.98 %). The model is significant at 1% level.

Correlation NDVI/ DEM

For both dates, there is a positive correlation between the altitude and the NDVI explained by a moderate coefficient of determination (70.08 % in May-92 and 71.76 % in May-96). An increase of elevation is followed by **an increase of NDVI**.



Multiple linear regression

Using the "Idrisi" module "Analysis/ Statistics/ MULTIREG", some multiple linear regression models between the NDVI average as a dependent variable and the yearly industrial growth ratio, the yearly urban growth ratio, the altitude and the amount of rainfall as independent variables were carried out.

Interpretation

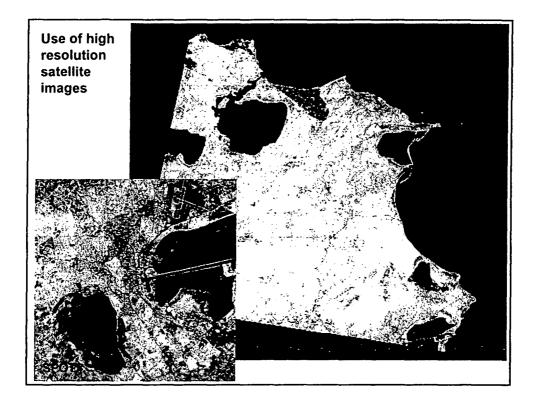
The NDVI is positively correlated with rainfall and DEM and negatively correlated with the yearly urban growth ratio and the industrial growth ratio.

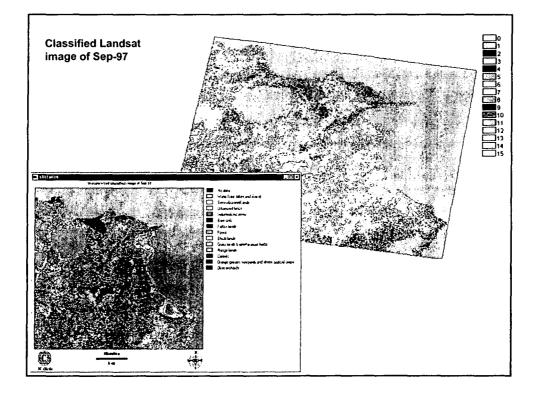
According to F values (F Statistical at 5% = 2.84, F Calculated = 40.881172 and F Statistical <<< F Calculated), the multiple regression's model is significant at 5%.

Concerning the partial correlation and according to t values: t at 5% confidence level =1.684, the NDVI is significantly negatively correlated with the urban growth ratio since -3.397961 belongs to the interval]- ∞ , -1.684] and significantly correlated with the industrial growth ratio since -3.371853 belongs to the interval]- ∞ , -1.684], significantly positively correlated with the altitude since 3.827727 belongs to [1.684, + ∞].

The analysis of the state of vegetation through the Normalized Difference Vegetation Index correlation with the industrial growth, the urban growth, the altitude and the quantity of rain, shows that the vegetation is more affected negatively by the human factors (urban and industrial development) than the natural parameters (rainfall and DEM).

While the changes in the vegetation are noticeable in time and over space, there is a gap of information obtained from the NOAA images of 1km resolution. A more detailed analysis involving LANDSAT images of 30 m resolution acquired for Sep-89, Sep-93 and Sep-97 would give a more meaningful representation of the actual vegetation patterns in this region of Tunisia.





Multi-criteria evaluation for industrial siting

Concept

Decisions about the siting of industries typically involve the application of Multi-Criteria algorithm based on logical PAIRWISE comparison. GIS offers the opportunity to apply decision-making processes in a more enhanced and integrated context.

Aim

To design suitability map of industrial sensitivity of the coastal zone of the study area.

Methodology

Investigation of interactive effects by contributing factors and constraints that may enhance or decrease industrial susceptibility. Factors contributing to the change of industrial susceptibility include:

✓Proximity to main industrial zones

- ✓ Proximity to main residential sites
- Proximity to main roads
- Proximity to water (sea, lakes and rivers)
- Proximity to protected forest & national parks
- ✓ Proximity to bares lands

The output will consist of a number of categories, each reflecting its susceptibility to the installation of new industries.

Weighted Linear Combination (WLC) is the most often used technique in multi-criteria decision making.

Criteria here may include weighted factors and constraints.

Calculating the product of weight and factor multiplied with all constraints at any location, and then summing up all products yields a total overall score.

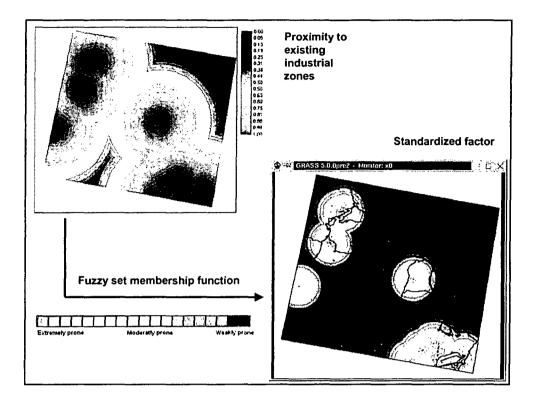
The score for each alternative A is:

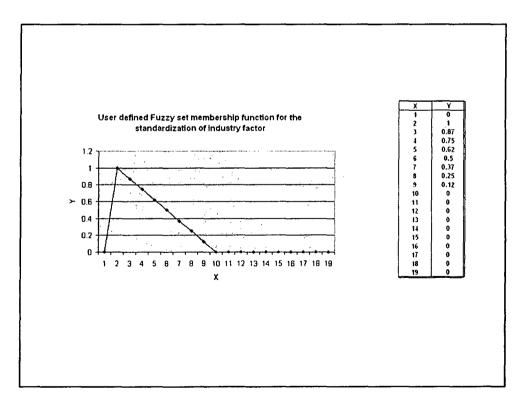
 $A = \sum wi xi$ or $A = \sum wi xi \prod cj$ if a constraint is part of the decision Where:

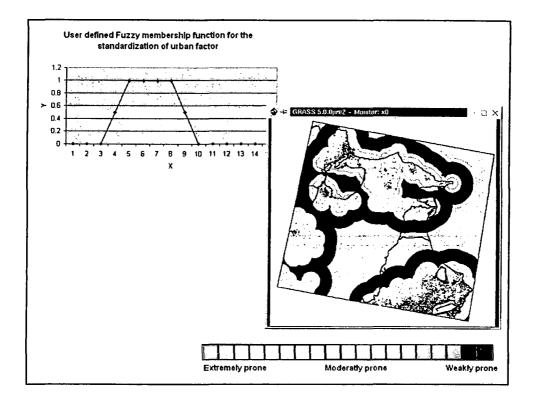
xi = criterion score of factor I;

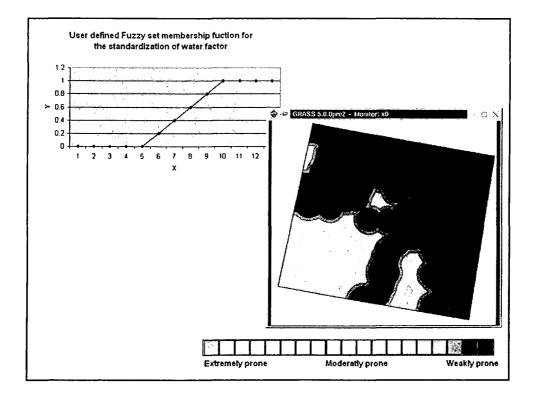
wi = weight of factor I;

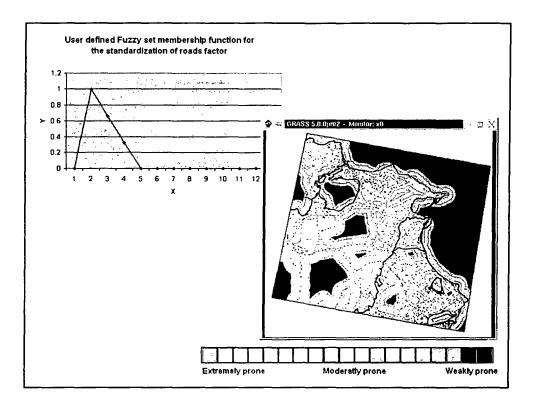
And cj = criterion score of constraint j.

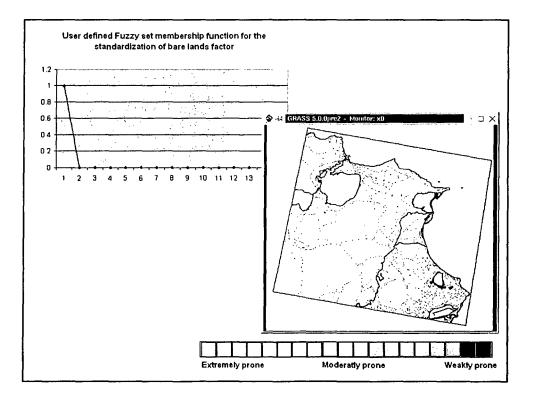


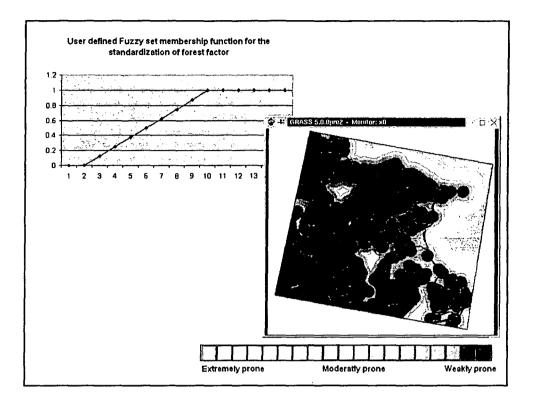




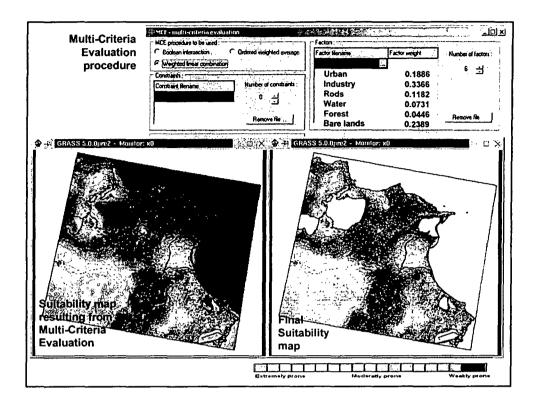








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Roads	1/2	1/3	1				
Water	1/3	174	1/2	1			
Forest	1/5	1/5	ר/3	1/2	1		
Bare lands]2	1/2	2	3	5	1	
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						Industry	: 0.33
						Rods	: 0.118
						Water	: 0.07:
						Forest	: 0.044
						Bare lands	. 0.23



Interpretation

The suitability map resulting from a Multi-Criteria Evaluation shows different classes for which the degree of susceptibility to accept new industrial plants vary from extremely prone areas to weakly prone.

The strongly prone areas are concentrated around the main existing industrial sites ("Echarguia", "Ariana", "Ettadhamen" and "Mannouba") and in the sub-urban (rural-urban interface) areas of the main cities.

This put in evidence the proximity to main roads and its good connection with the market (access to transportation is an important consideration) and also its natural location (coast, rivers, lakes, slope, forest, etc...) which can represent an environmental limitation.

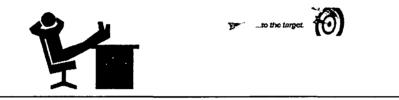
Conclusion

GIS has the capacity to integrate information form a variety of sources into a spatial context in particular from remotely sensed (RS) data.

GIS is well suited to support decision making procedures and can act as a tool in helping the decision-makers in evaluating certain alternatives.

In the end, it is the decision maker, who determines the criteria, the factors, the constraints, the individual weighting and the decision rules.

...But also who takes decision.





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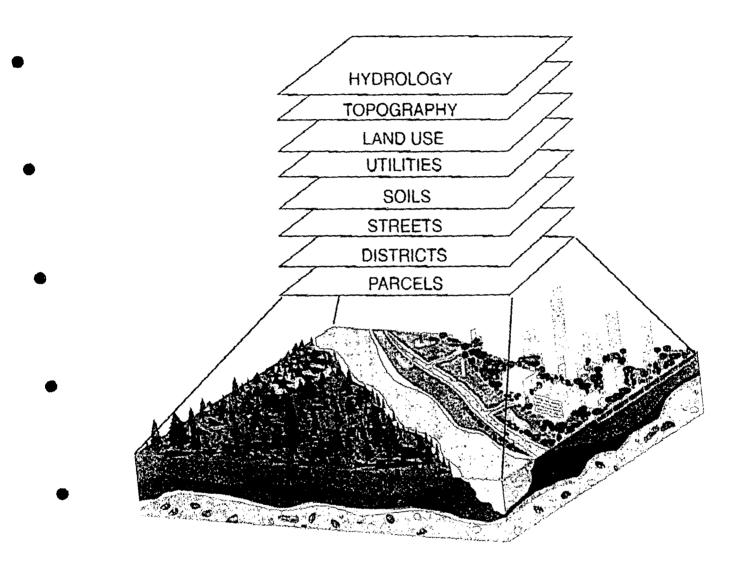
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Practical Exercises on Image Processing and Multi-Criteria Evaluation (MCE) Using IDRISI Software Package by M. Ghribi





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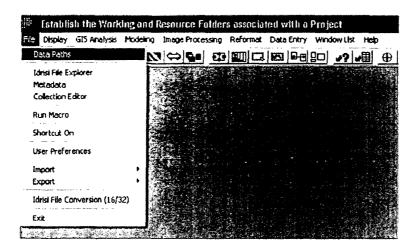
Exercises

1. Getting started with IDRISI

IDRISI for Windows consists of a main interface program (with a menu and toolbar system) and a collection of over 150 program modules that provide facilities for the input, display and analysis of geographical and remotely sensed data. These data are described in the form of map layers.

1.1. Start IDRISI

From the Desktop double-click on the IDRISI GIS Software Version 32, release2, shortcut to start the program. After the start up process, you should establish a working directory. From the menu bar choose **File**, and then **Data Paths**.



The Operating Environment dialogue box will appear. Directories are shown on the right. Move to the **c:\ics\data** directory by double-clicking each folder successively.

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To explore the exiting data, go to File/ Idrisi file explorer

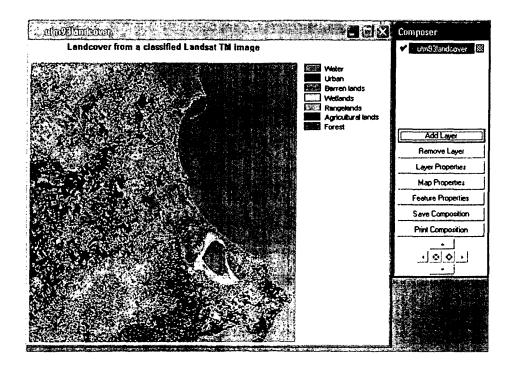
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To open an existing project, choose a raster file name from the list, for example **utm93landcover** and then click **Display Layer** or click on **Display** in the menu bar and choose **DISPLAY Launcher**.

🕮 DISPLAY Launcher						
File type to be displayed	Palette file					
🙃 Raster Layer	C Quantitative (Standard IDRISI Palette)					
C Vector Layer	C Qualitative					
	Color Composite (8 bit)					
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utm93landcover	C NDVI (Green Vegetation Index Palette)					
TAutoscale 🔽 Title 🗗 Legend	User-defined :					
OK Cancel Help	utmlandcover					

1.2. Open an existing raster file and overly vector files

In the Display Map Composition Launcher dialogue box, double-click on the input image file box and choose **utm93landcover**. Select the palette file **utmlandcover** and click the map components boxes to add the Title, Legend, Logo, etc. To display the raster map layer click OK in the icon field.

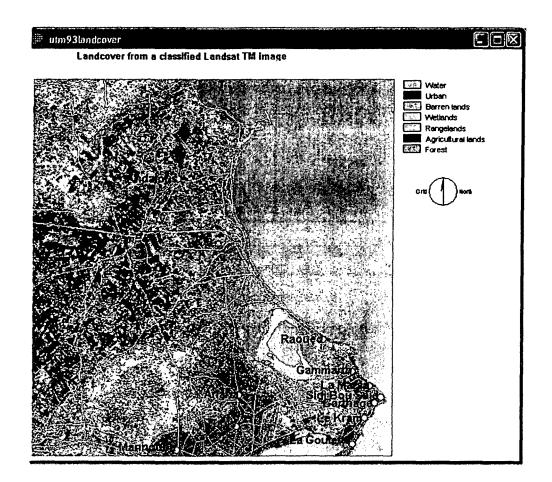


Once the raster map layer is displayed, go to the Composer dialogue box and press Add Layer icon to add vector map layers. Double-click on the input file window and choose **utm93roads** with **uniwhite** as symbol file, the **utm93cities** with **utm93cities** as symbol file and **utm93citiestext** with **utm93text** as symbol file. This will provide respectively: the main roads, the main cities and the names.

To add other components of your map, click on Add Layer in the Composer dialogue box and choose some components: **scale bar**,**n orth arrow**, etc.

To **Save composition**, in the composer go to **Save composition** then save composition as MAP File and give a file name to your final map composition for example **utm93mapcomposition**.

🕮 Save Composition								
Output Options © Save composition to MAP file								
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2. Markovian Transition Estimator, cellular automata and Markov change land cover prediction

Since we have 3 sets of data (3 Landsat TM satellite images of the same period for 1989, 1993 and 1997), we will use the **Markovian Transition Estimator** to compare the first 2 land cover images of 89 and 93 (4 years time period) and we will project the results for 1997. And since we have already the 97 land cover, we will be comparing the predicted and the classified land cover images.

Go to GIS Analysis/Change-Time Series/Markov

🎬 Markov - Markovian transition estim	noter noter						
First (earlier) land cover image:	utmR9landcover						
Second (later) land cover image:	utm93landcover						
Prefix for output conditional probability images: utmmarkov							
Number of time periods between the first and sec Number of time periods to project forward from the Background cell option (C Assign equal probabilities (C Assign relative frequency	- 1						
Proportional error:	0.0						
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The MARKOV module analyzes a pair of land cover images **utm89landcover** and **utm93landcover**. The output is a transition probability matrix, a transition areas matrix, and a set of conditional probability images (having **utmmarkov** as prefix).

The transition probability matrix is a text file that records the probability that each land cover category will change to every other category.

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						a.				//,

The transition areas matrix is a text file that records the number of pixels that are expected to change from each land cover type to each other land cover type over the specified number of time units.

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Class	s 5	:	2934	58578	2068	69570	119138	11387	
Clas:	s 6	:	1046	12027	1018	39598	11493	5957	

In both of these files, the rows represent the older land cover (1989) categories and the columns represent the newer categories (1993).

CA_MARKOV

CA_MARKOV is a combined cellular automata / Markov change land cover prediction procedure that adds an element of spatial contiguity as well as knowledge of the likely spatial distribution of transitions to Markov change analysis. For our case we will predict the change for **4 years** period. The basic land cover image is **utm93landcover**, the Markov transition areas file is **utmmarkovtransition_areas** and a transition suitability image collection **utmmarkovsuitabilities_group**. The cellular Automata filter is the standard 5x5 contiguity filter. The output is a projected land cover image of 1997 (**utm97landcover_markov**).

Go to GIS Analysis/Change-Time Series/CA_MARKOV

Basis land cover image:	utm93landcover					
Markov transition areas file:	utmmarkovtransition_areas					
Transition suitability image collection:	Sutmmarkovsuitabilities_group					
Dutput land cover projection:	utm97landcover_markov					
Number of Cellular Automata iterations:	4					
Cellular Automata filter type:	 Standard 5 x 5 contiguity filter User-defined filter 					
Dutput doc	umentation					



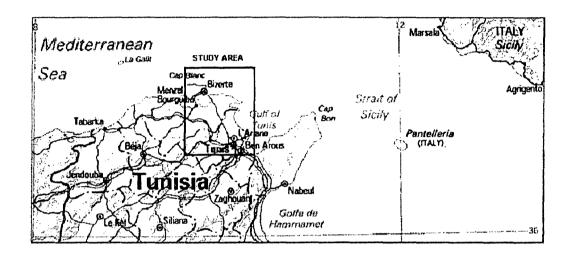
We should first create the utmmarkovsuitabilities_group file.

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3. Multi-criteria evaluation for urban settlement

3.1. Description of the study area

Human settlements in the Mediterranean basin have existed since ancient times and have had a profound impact on the flora and fauna of the region. Consequently the Mediterranean system offers a very interesting field for investigations on human impact. The area has been selected for study because it is typical of the Mediterranean environment, since is located on the coast of the Gulf of Tunis in the north- eastern Tunisia.



3.1. Purpose

In the present exercise, we will try to design a suitability map depicting the sensitivity of the coastal areas of the study area to the expansion of urban settlement. It consists of landscape evaluation for urban planning and includes a GIS technology developed for the IDRISI geographic information system software.

3.2. Procedural steps

A GIS-based methodology is integrated by using **Weight** IDRISI module. The concept consists of inserting interactive effects of several contributing factors and constraints that may contribute in enhancing or decreasing the susceptibility.

🕮 C:\DATA\utm93urban pairwise.pcf	. 145. P	 5	
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cursor position: 1,18	 		

The factors are raster images containing the target features from which distance is measured thanks to Distance IDRISI module. The constraint, on the other hand, is to exclude certain areas from consideration, such as water and reserved lands (urban and roads).

Go to **GIS Analysis/Distance Operators/Distance** to create distance files to all selected factors that contribute to the final suitability map.

Table: Factors contributing to the change of urbanization susceptibility

Factors	File name	Description
1	Utm93urbandisfuzzy	Proximity to main urbanized areas
2	Utm93industrydisfuzzy	Proximity to main industrialized sites
3	Utm93roadsdisfuzzy	Proximity to main roads
4	Utm93waterdisfuzzy	Proximity to water (sea, lakes and rivers)
5	Utm93forestdisfuzzy	Proximity to protected forests and national parks
6	Utm93barrendisfuzzy	Proximity to bares soils
7	utm93slopedisfuzzy	Proximity to hills and high slope

To standardize factors, we will use a fuzzy set membership for each category. Go to **GIS Analysis/Decision Support/Fuzzy**

FUZZY - Fuzzy Set membersh	ip function			
Membership Function Type :	<u> </u>	**************************************		
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The weights are developed by providing a series of PAIRWISE comparisons of the relative importance of factors to the suitability of pixels for the objective being evaluated. Ratings are provided on a continuous nine-point scale from extremely less important to extremely more important.

Select GIS Analysis/Decision Support/MCE.

MCE is a decision support tool for Multi-criteria evaluation. A decision is a choice between alternatives (such as alternative actions, land allocations, etc.). In a MCE, an attempt is made to combine a set of criteria to achieve a single composite basis for a decision according to a specific objective.

MCE procedure to be used :		Factors :			
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.tm93urban01	L	utmS3industrydistuzzy	0.2031		
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utput image : utmS3urben	suitability	Retrieve p	varameters S	ave pa	rameters

The result is a set of factor weights used by a weighted-linear combination in the multicriteria evaluation.

3.4. Suitability map and Interpretation



The suitability map produced from a multi-criteria evaluation shows different degrees of susceptibility to new industrial plants. As shown in figure 2.8, the extremely prone areas are concentrated around the main existing industrial sites (Echarguia, Ariana, Ettadhamen, Mannouba, Mateur, Menzel Bourguiba and Bizerte), in the urbanized areas and along the coast. Highly prone areas to industrial development are characterized by:

- Proximity to an existing industrial zone in which there is a high probability that infrastructure and organized industrial system exist;
- Proximity to residential and commercial sites fulfilling the requirement for social activities. The availability of high number of workers will satisfy labour requirements for industrial activities;
- Proximity to main roads and connection with markets;
- Natural allocation (coast, rivers, lakes, slope, etc.), which can represent an environmental constraint.

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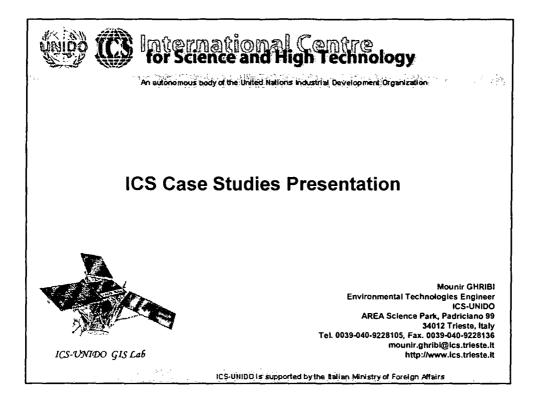
Sol, H. G. (1983) Processes and tools for decision support: Inferences for future developments. In Processes and Tools for Decision Support, (ed. Sol, H. G.), North Holland, Amsterdam, the Netherlands, pp 1-6

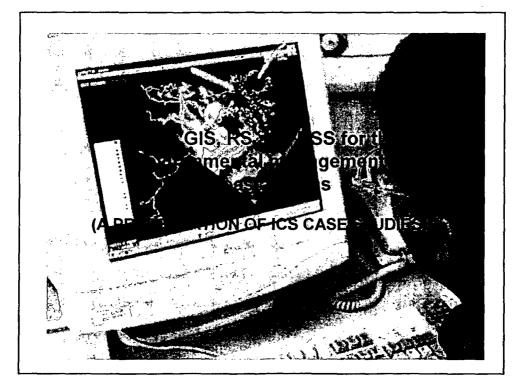
Sprague, R.H. and Watson, H.J (1986) Decision Support Systems: Putting Theory into Practice. Pretice Hall, New Jersey

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ICS Case Studies Presentation by M. Ghribi





ICS GIS lab software packages

GRASS (GIS) Geographic Resources Analysis Support System, is a public domain (free) Geographic Information System used for data management, image processing, spatial modeling, etc.

MultiSpec, is a processing system for interactively analyzing Earth observational multi-spectral image data such as that produced by the Landsat series of Earth satellites and data from current and future airborne and space borne systems such as AVIRIS (free).

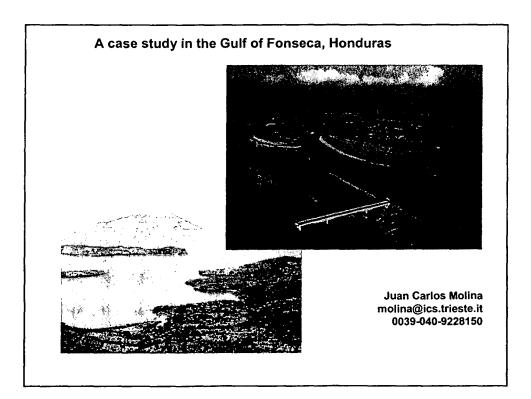
IDRISI (GIS) is a pc-based, (mostly) raster GIS from Clark labs. It couples the extensive analytical capabilities of the GIS and Image Processing System with a highly interactive graphical user interface. It provides raster analytical functionality covering the full spectrum of GIS and Remote Sensing needs from database query, to spatial modeling, to image enhancement and classification.

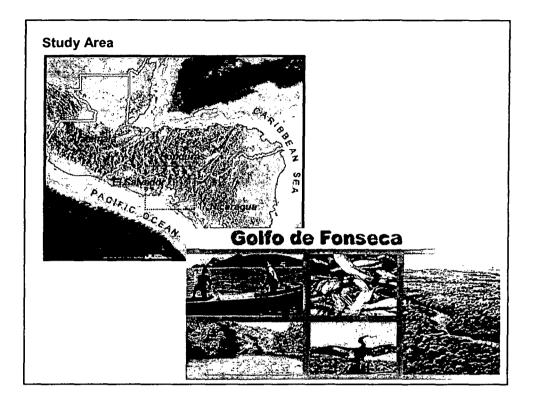
ArcGIS (GIS) ESRI products containing ArcView and ArcInfo. It is a desktop mapping and a complete GIS software. Has hundreds of mapping and spatial analysis capabilities.

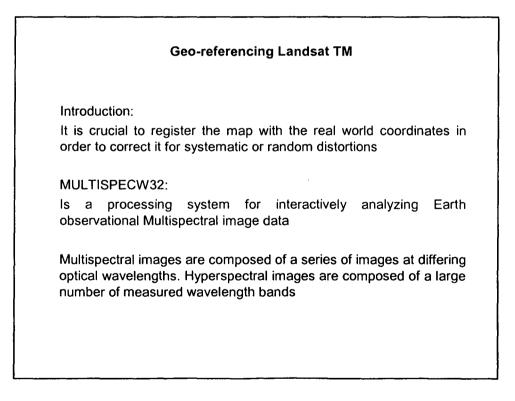
ERDAS IMAGE (Image Engineering) offers a flexible and scalable solution for mage input, management, display and enhancement, spectral analysis, digital photogrammetry, GIS data integration and map composition. Imagine provides a robust set of tools for geo-correction, image analysis, visualization, and map output.

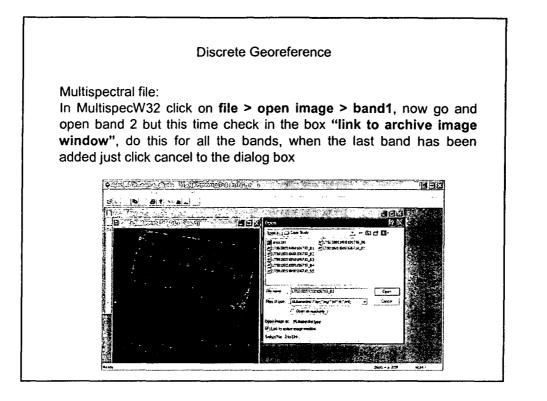
Delf-3D (Environmental Simulation) is a framework of software modules for 2 and 3-dimensional flow, water quality, ecology, short wave propagation and morphology and their interactions. The software has been developed over the passed 15 years and still being extended and improved in functionality's and application.

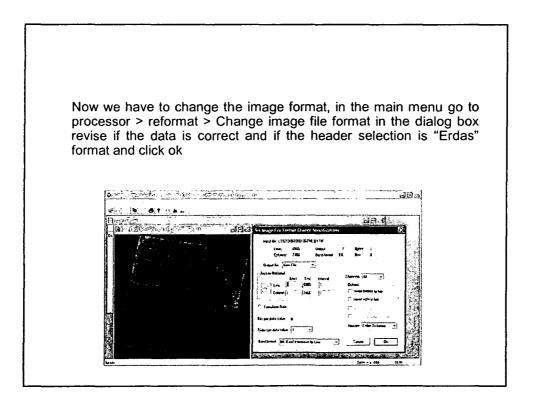
SPSS & STATISTICA (statistics) generate decision-making information using powerful statistics, tabular and graphical output, and offer results to other packages using a variety of reporting methods, including secure Web publishing.











IDRISI:

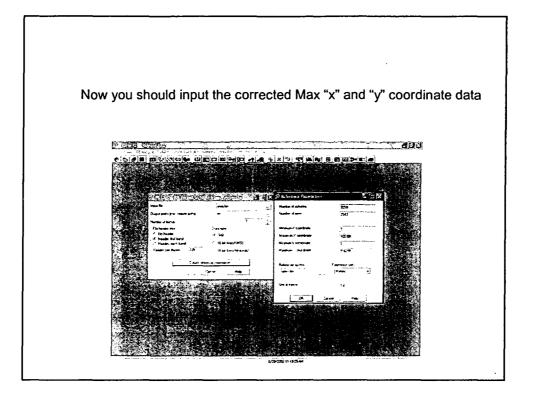
consists of a main interface program (the one with the menu and toolbar system) and a collection of over 150 program modules that provide facilities for the input, display and analysis of geographic data.

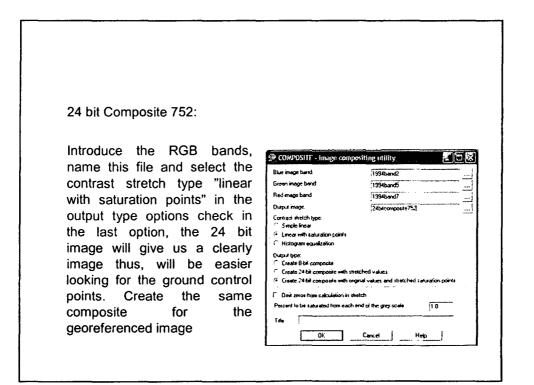
Importing the Multispectral file: In IDRISI's main menu go to click on File > Import > General

Conversion Tools > BILIDRIS (BIL with/without header)

When introducing the Maximum and Minimum "x" and "y" coordinates you have to be careful, first check if resolution is 30 if not apply the following formulas to find the real coordinate values

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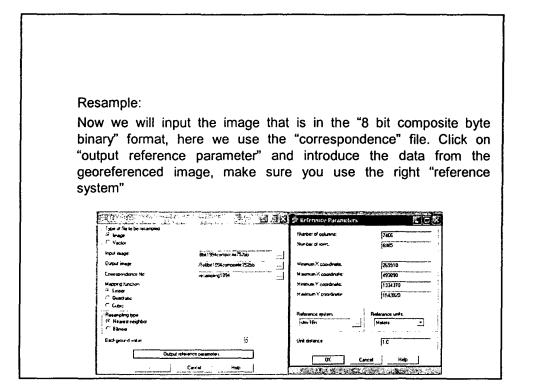


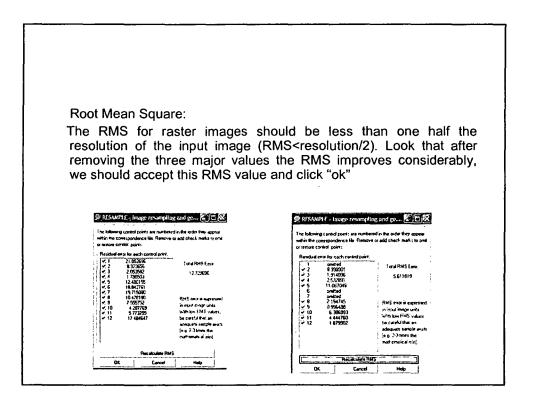
Correspondence file:

This file contain the coordinates of specific points termed Ground Control Points in two different reference systems, the first pair of coordinates correspond to the locations in an arbitrary plane reference system, the second corresponds to the same locations in the desired georeferencing system

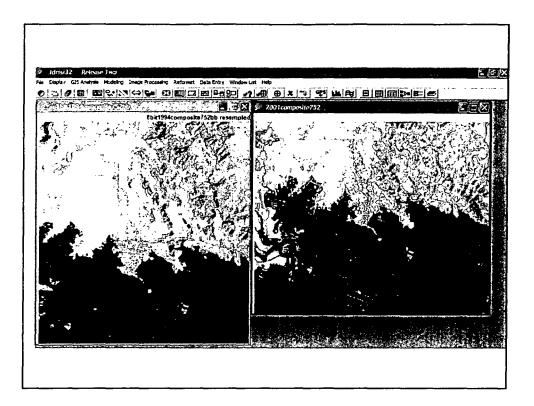
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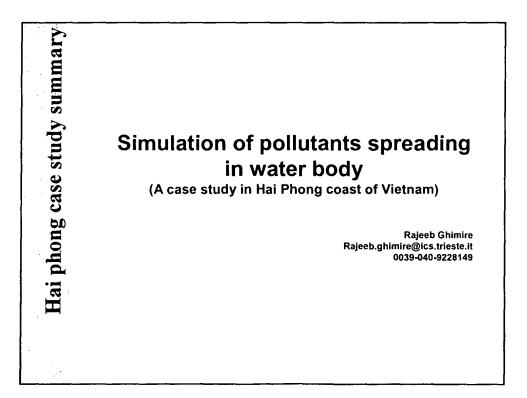
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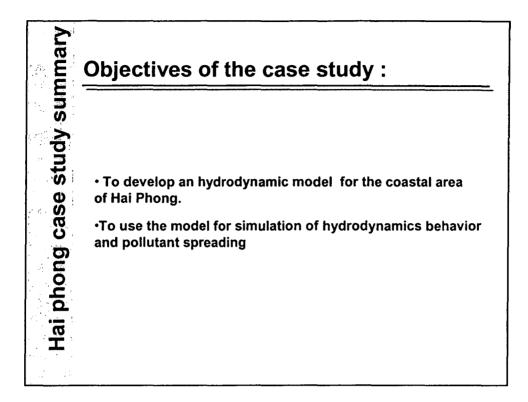


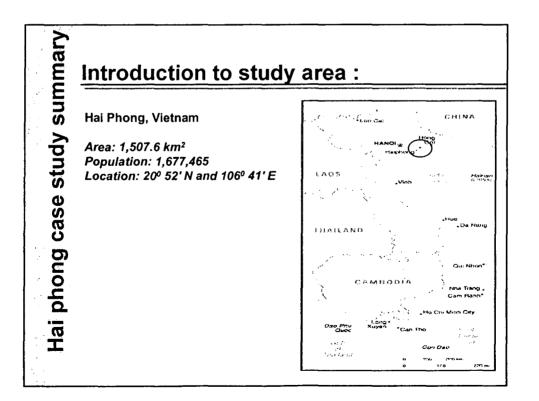


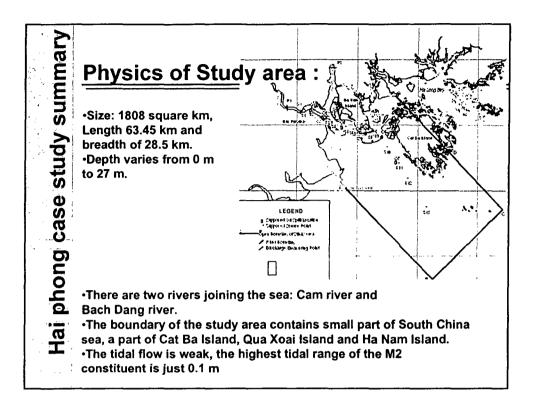
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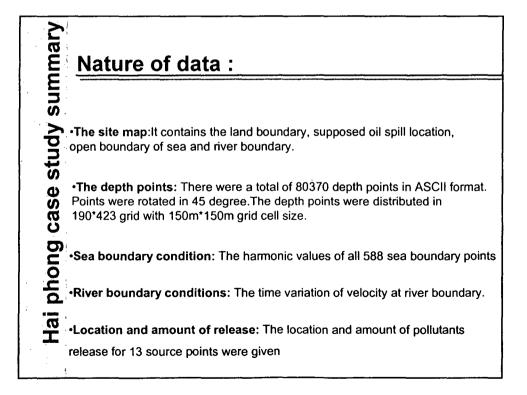


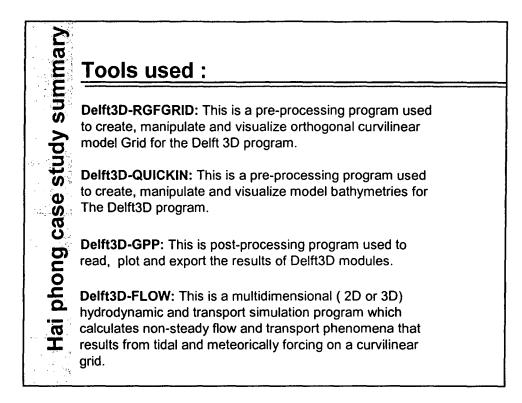


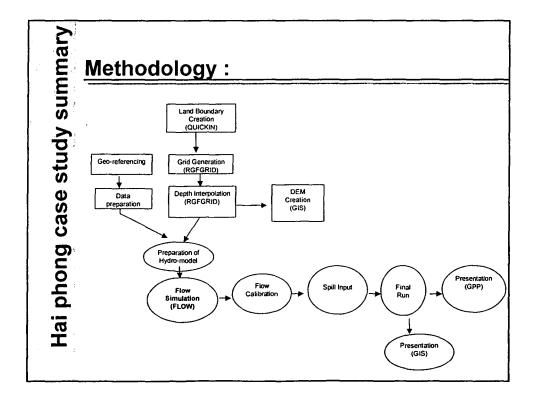


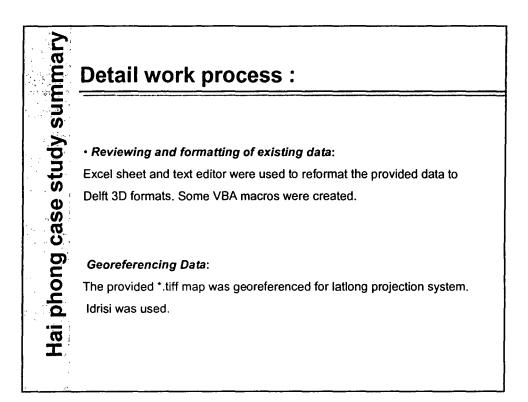


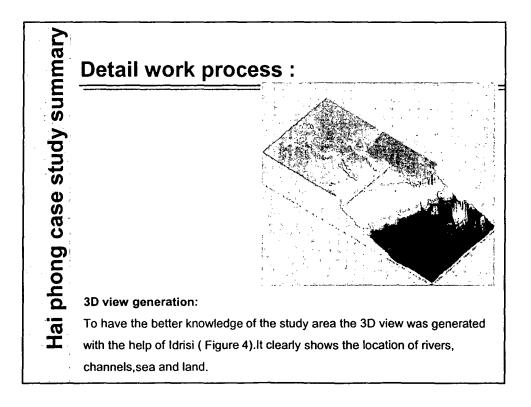


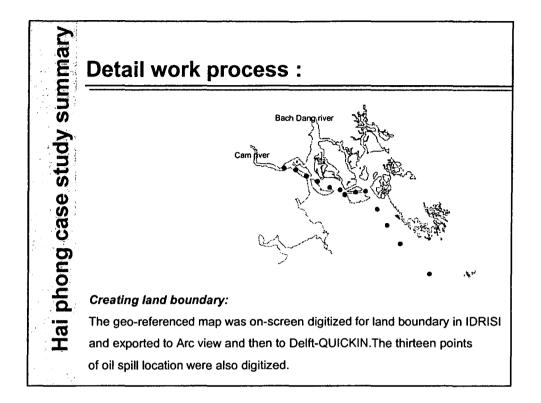


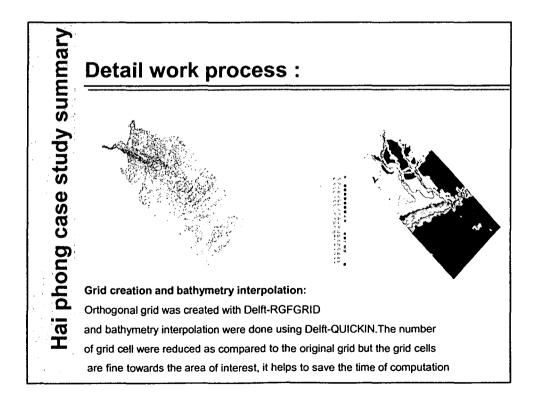


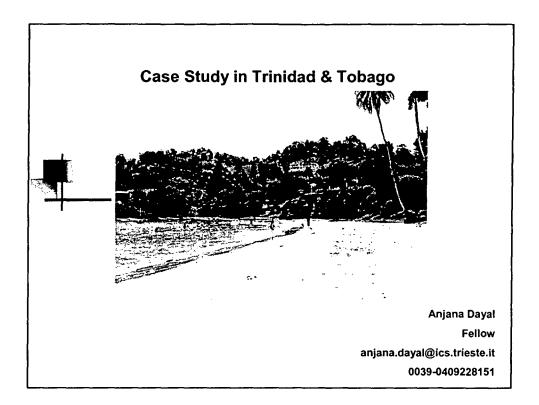


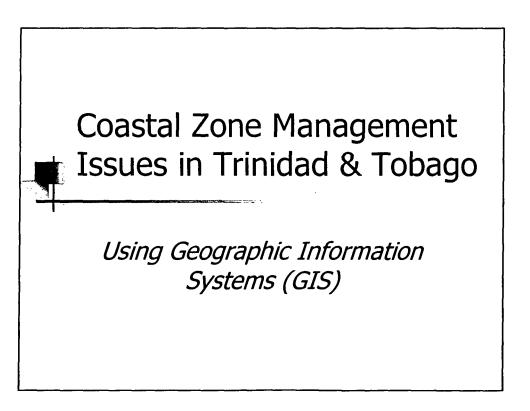


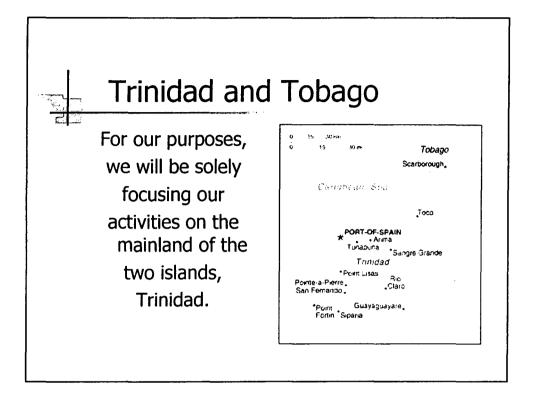


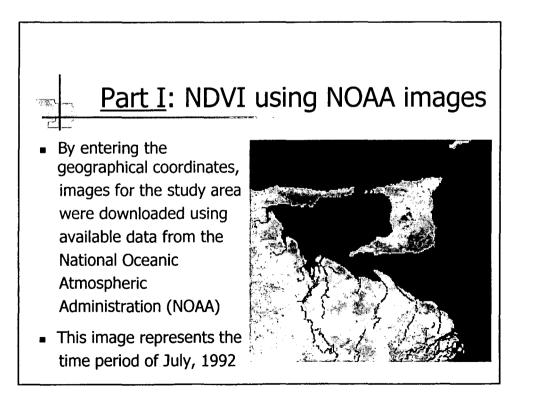






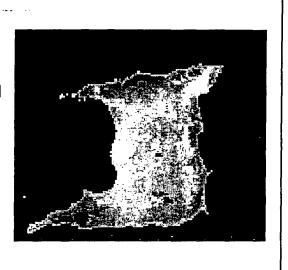


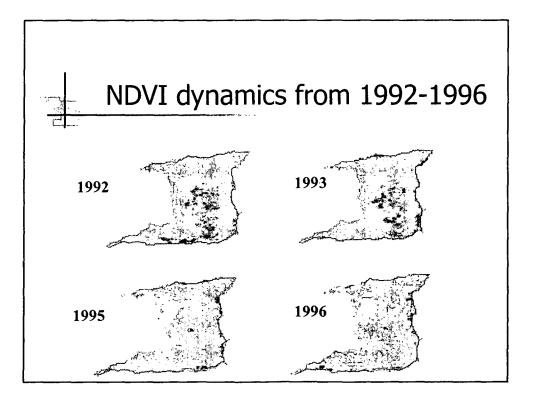


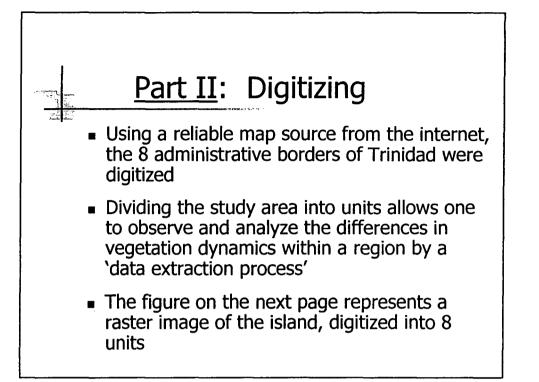


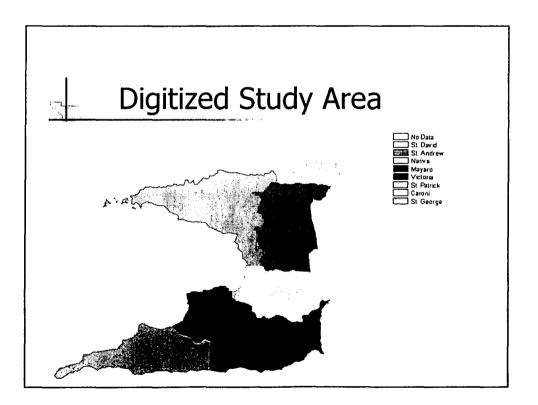
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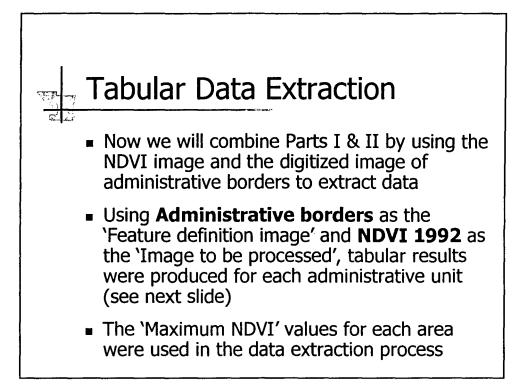
- The following image is a reformatted window taken from the original study area in the previous slide
- The next slide illustrates the changes in NDVI over a time period of 4 years

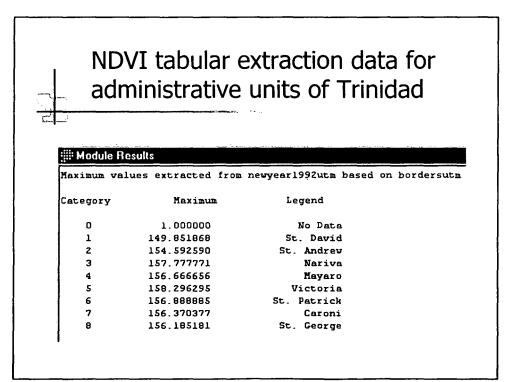


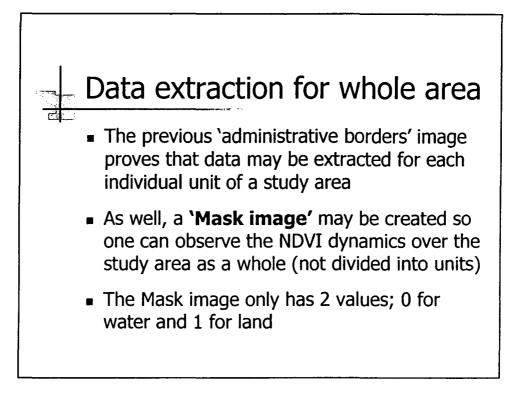


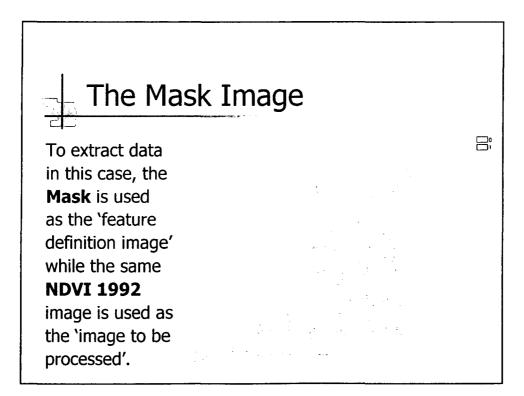


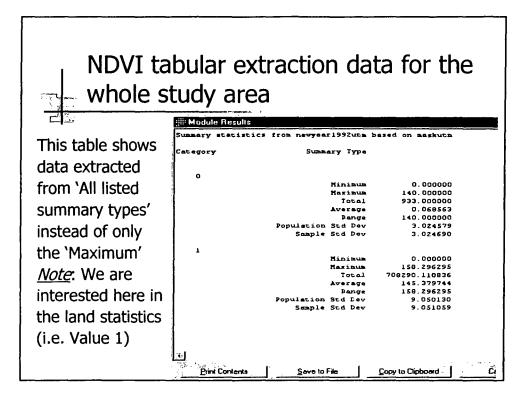


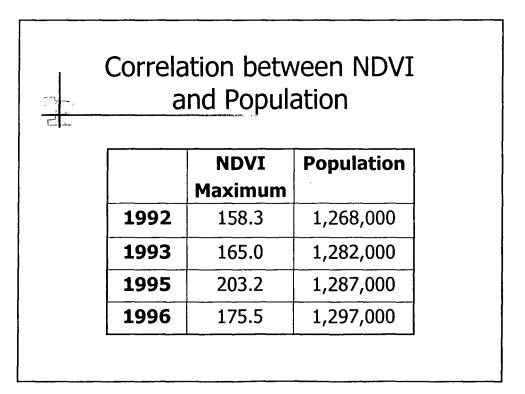


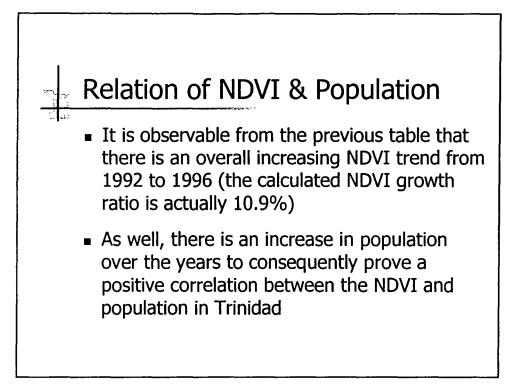












Case Study for the municipality of Koper, Slovenia

M. Russi

INTRODUCTION

•Slovenia is mainly a continental state. The total length of the coastline, all facing the Bay of Trieste, is only 46 km. This small coastal area is of strategic economic and political importance for the whole country.

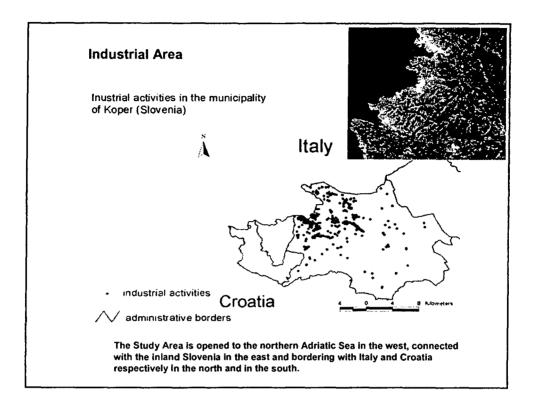
•Along this coastal area Three towns have developed there: Koper (administrative center of the area, <u>industry</u> and harbour activities), Izola (tradition in the fishing industry (cannery)), Piran and neighbouring Portorož (highly developed tourist center) and smaller settlements and tourist facilities between them, so the entire coast is one large urban area.

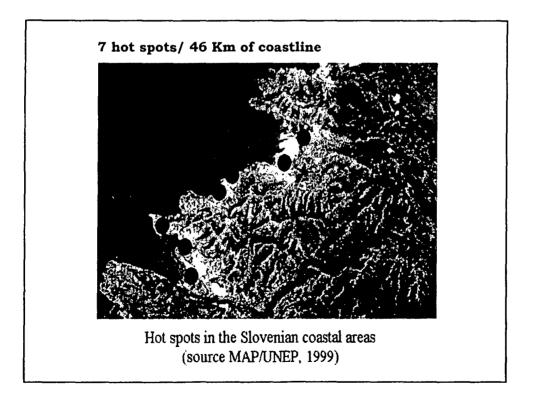
•For this reasons competition for the land use among housing, marine activities, tourism and industry is strenuous. However, close to this large urban area, environmentally valuable - unique - locations are located (ex. the Valley of Dragonja).

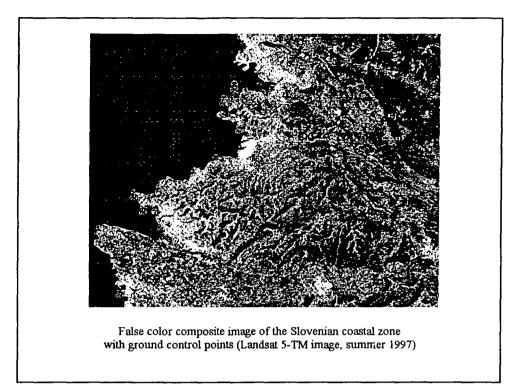
•As a result of these conflicts **seven hot spots** have been identified in the Slovenian coastal zone in the frame of the Mediterranean Action Plan (MED POL programme) survey. They are located in all the three municipalities.

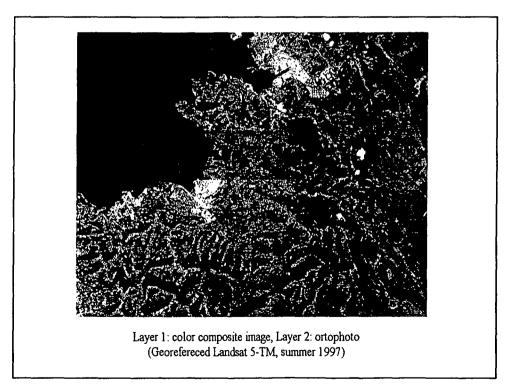
•Surface flows, numerous urban wastewater outlets and one submarine outfall are the main sources of pollution of the coastal sea. Untreated industrial (three hot spots) and urban wastewater in the rivers are among the biggest problems.

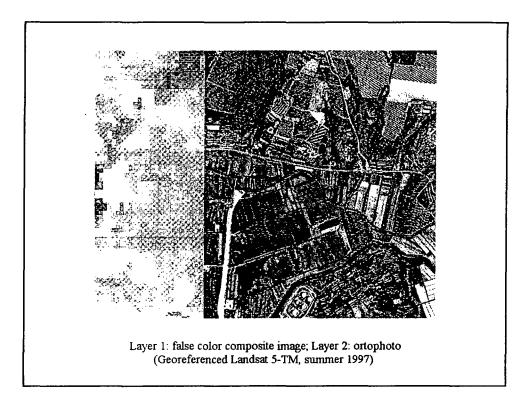
•The contribution of land-based sources of pollution situated on the Slovenian coastal zone is much smaller than the overall value for the entire Bay of Trieste. The Italian side contribution must be -also- considered.

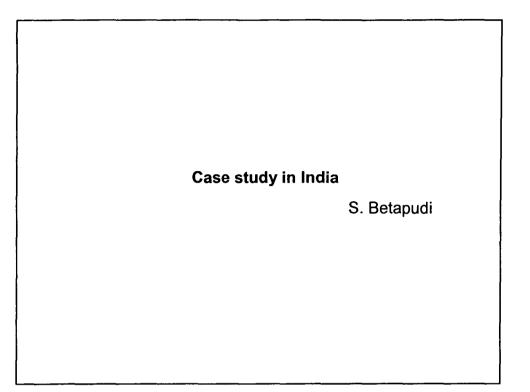


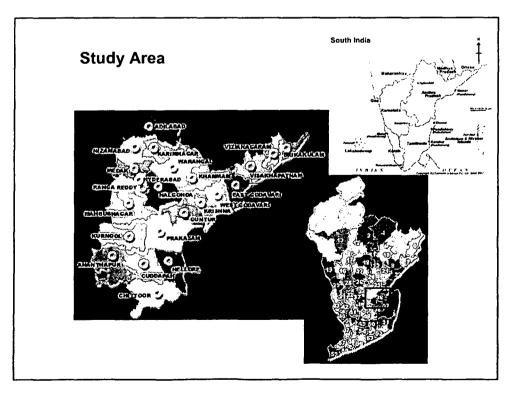


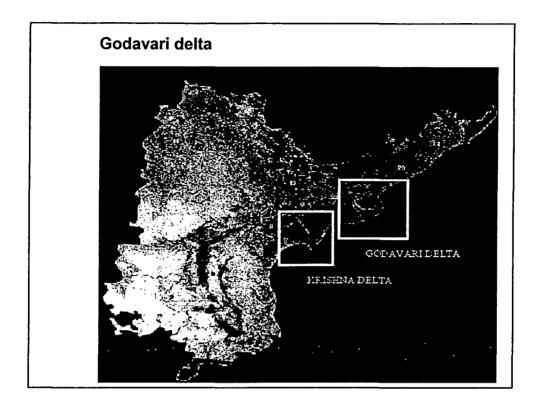


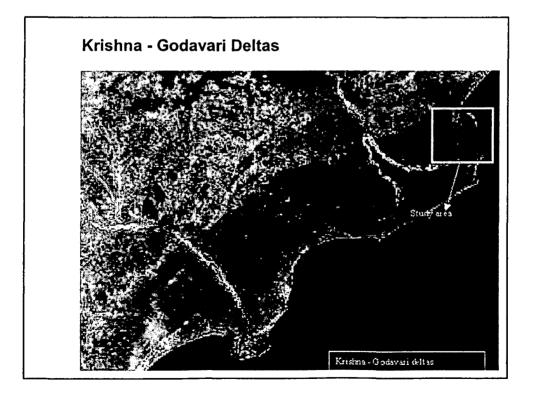


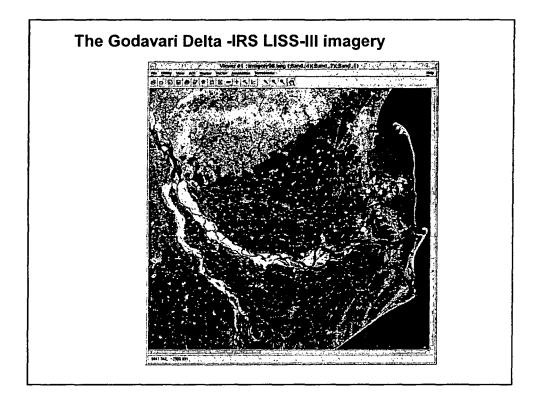


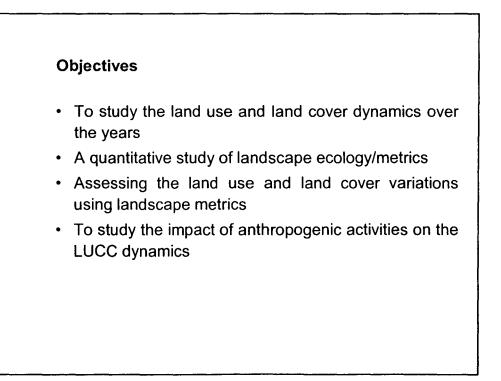


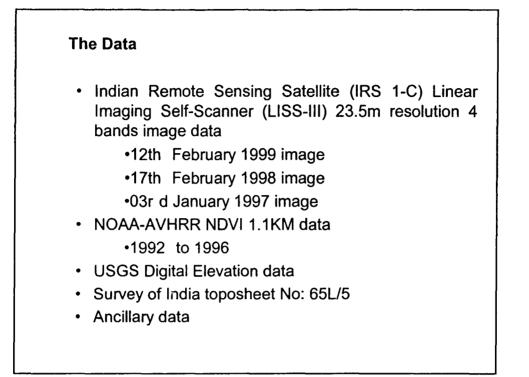


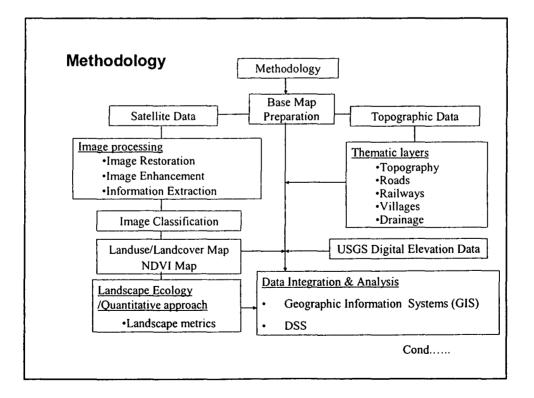


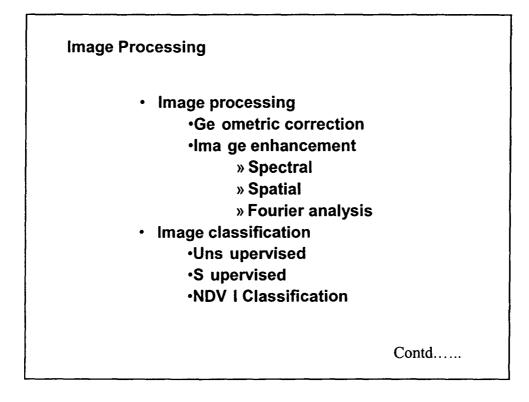


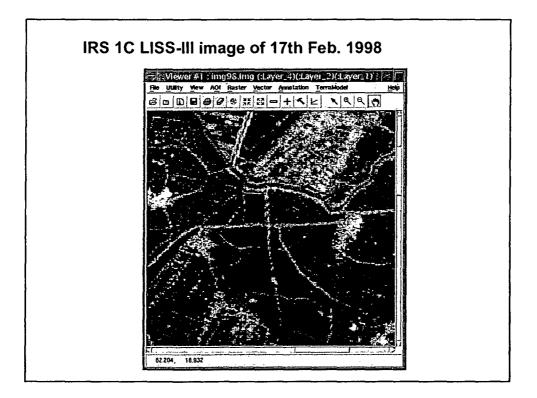


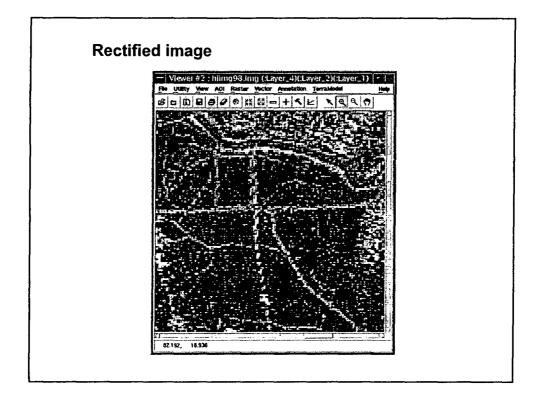


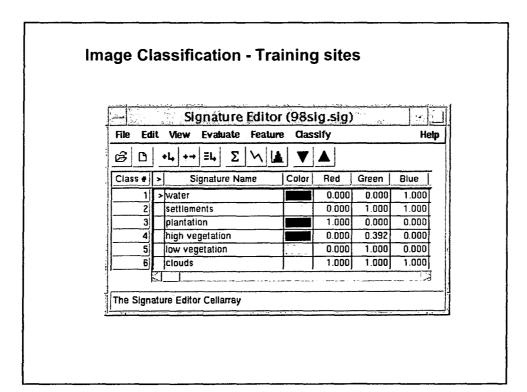


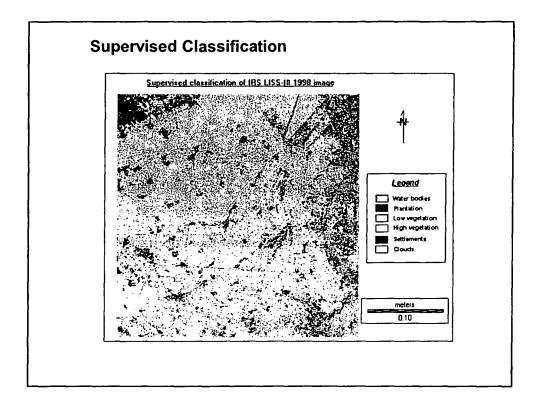


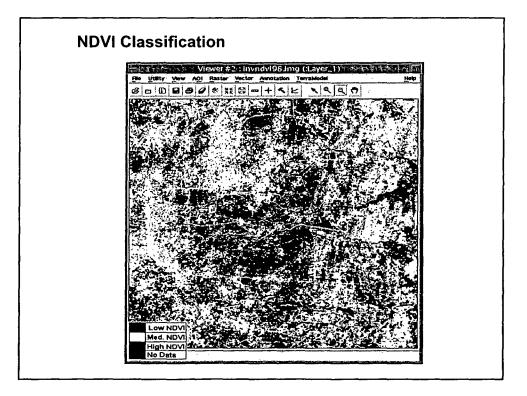


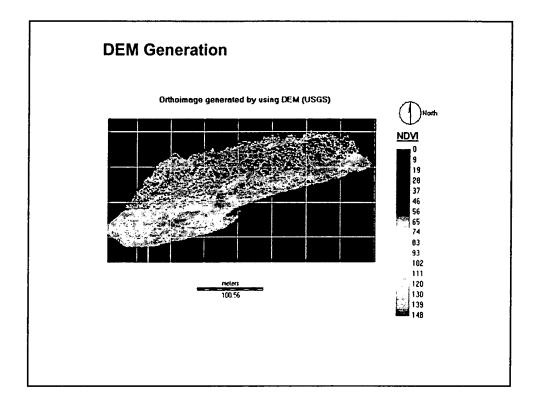


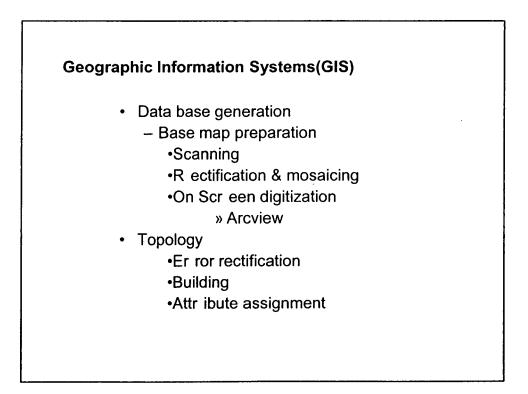


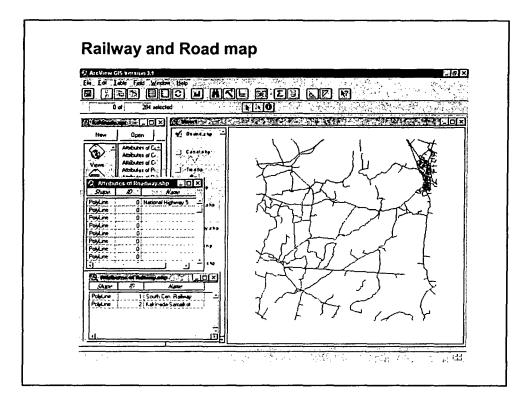


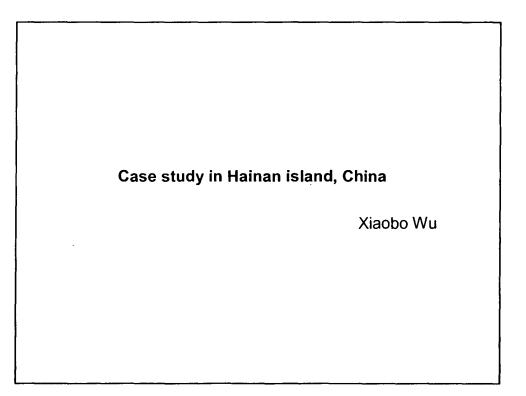


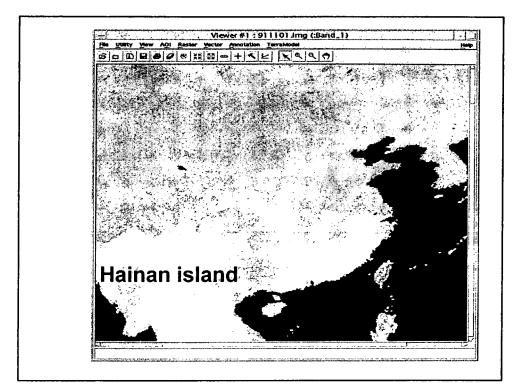


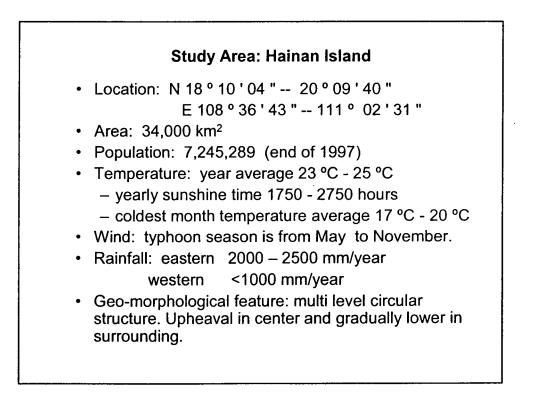




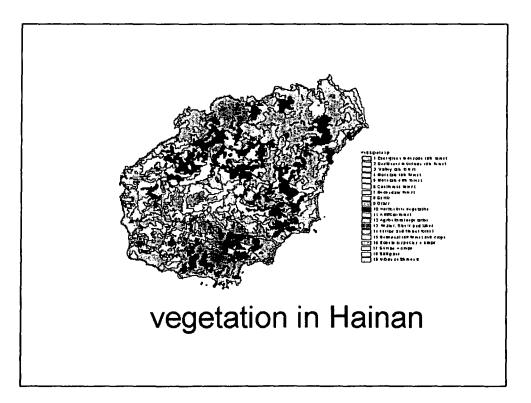


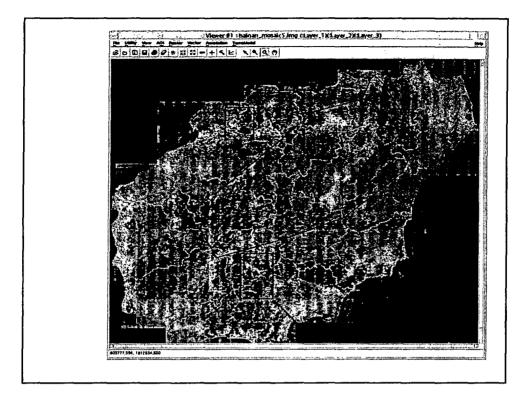


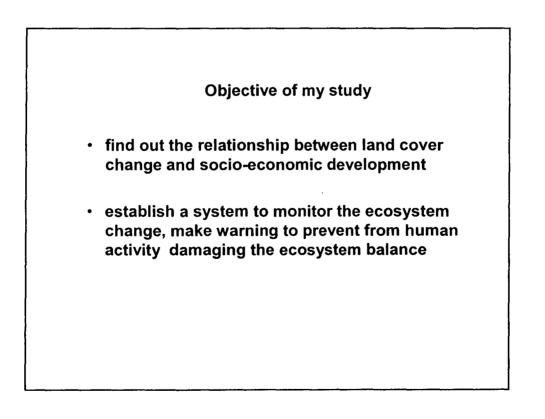


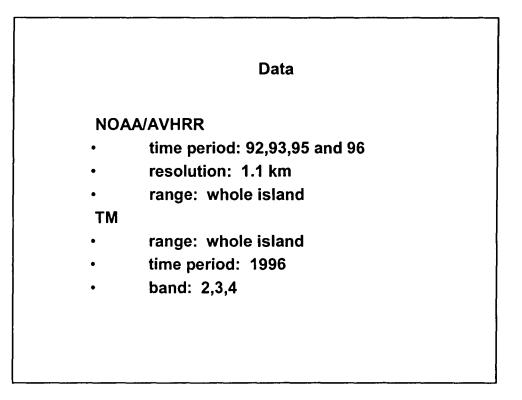


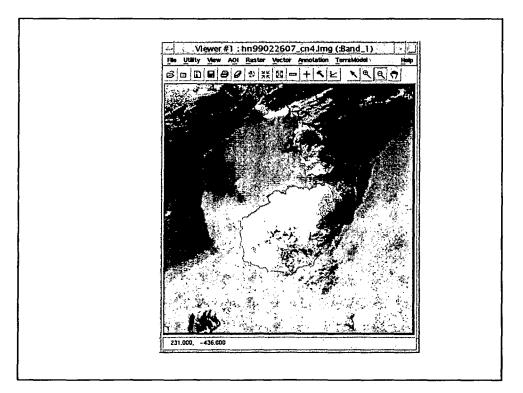
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 - · defoliate monsoon rain forest
 - valley rain forest
 - mountain rain forest
 - mountain evergreen broadleaf forest
 - tropical coniferous forest
 - coastal mangrove
 - · tropical tree-scattered shrubs
 - tree-scattered steppe
 - damp steppe
 - low hill and tableland steppe
 - hill and mountain steppe

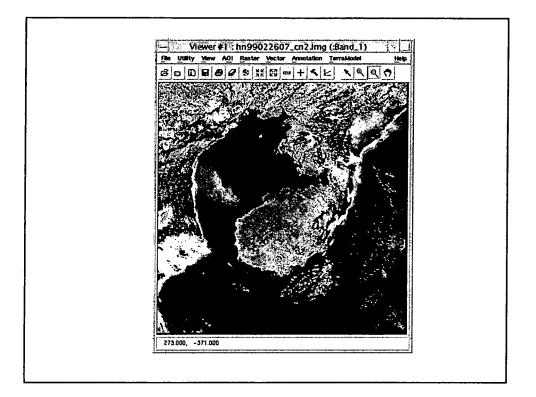


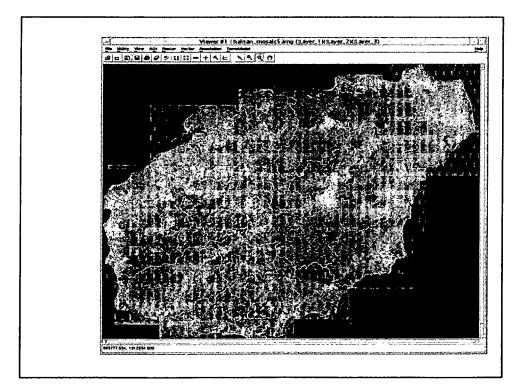






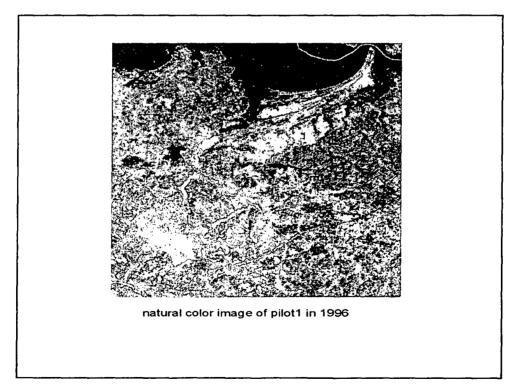


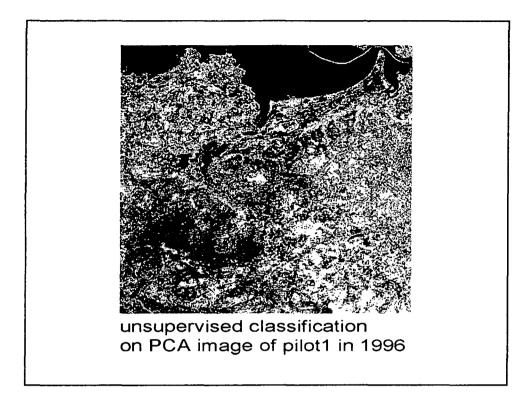


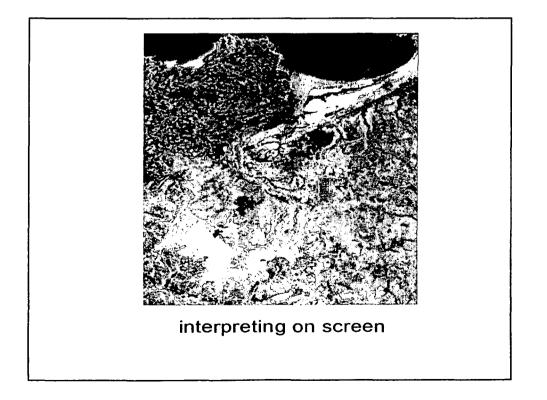


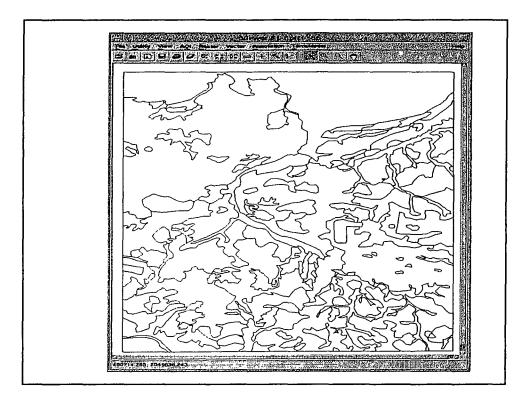
ancillary data (Chinese)

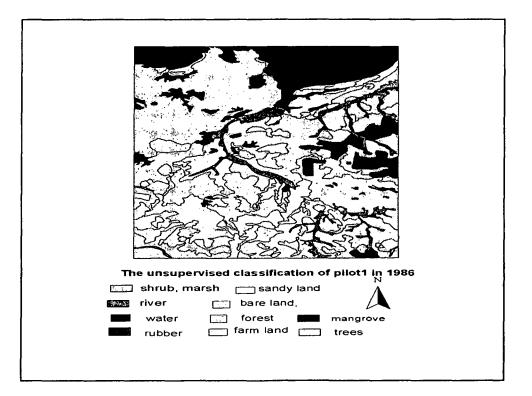
- some published papers and books on soil, landscape, geological structure, geomorphologic feature, climate etc in Hainan Island.
- forest distribution map of Dan Zhou andTong Zha (scale: 1: 50,000 1:100,000)
- DTM of Nan Du River Basin
- forest distribution of Nan Du River Basin in digital format
- socio-economic data on county level from 1987 -1999

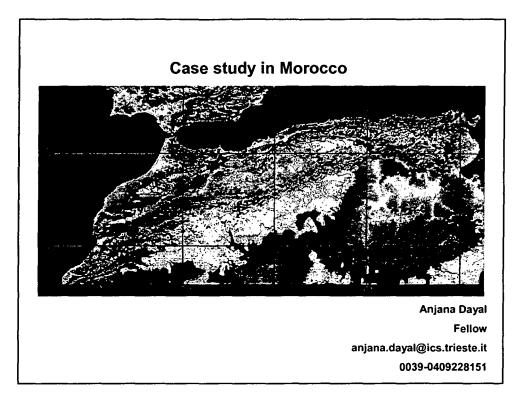


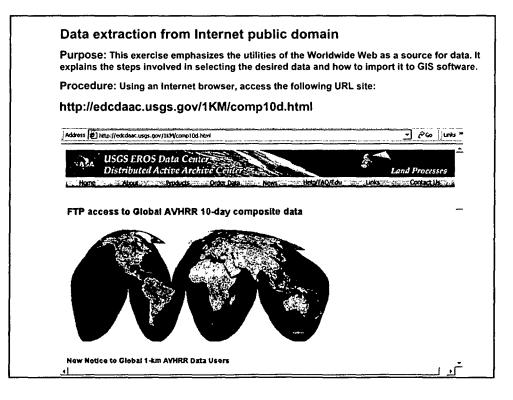


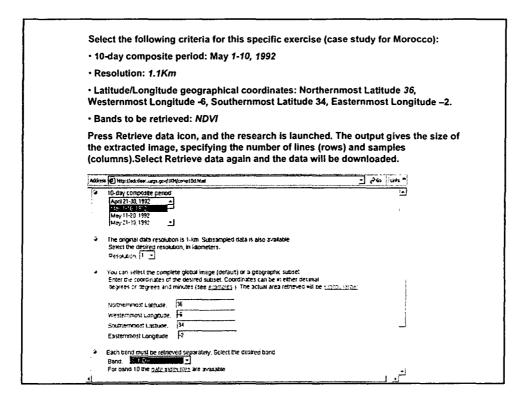








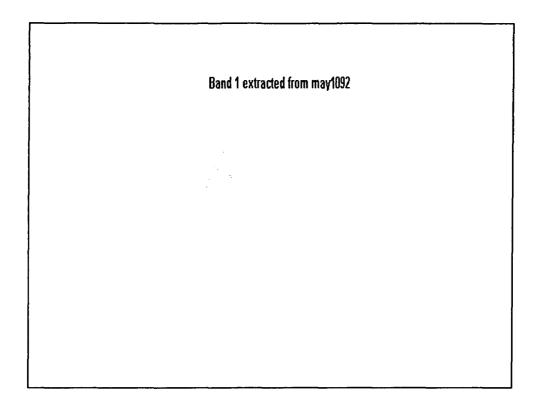




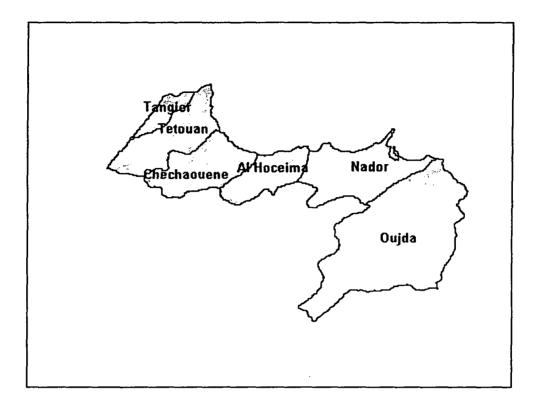
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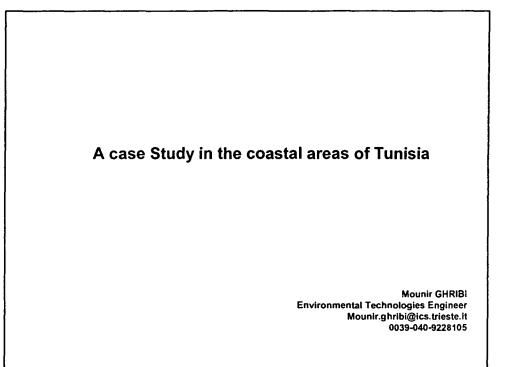
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•	BIL, Band Interleaved by Line
•	BSQ, Band SeQuential
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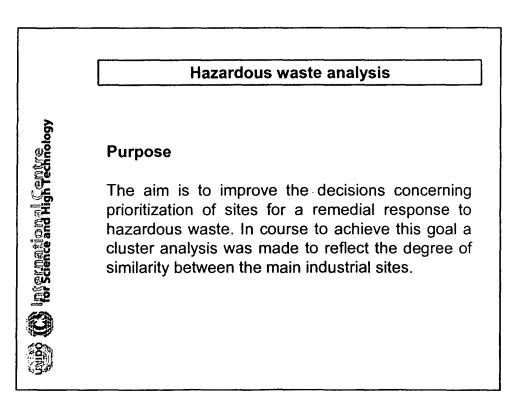
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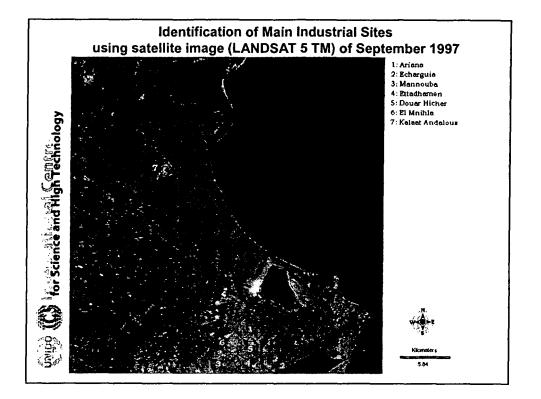


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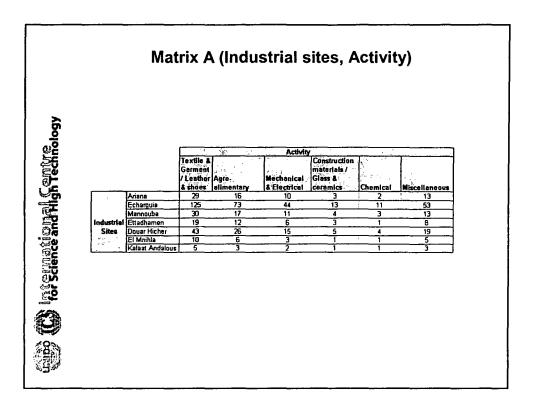


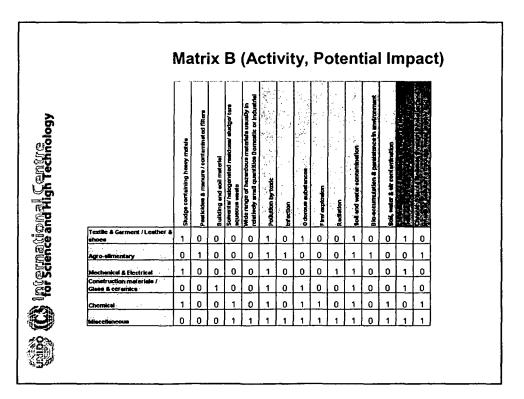


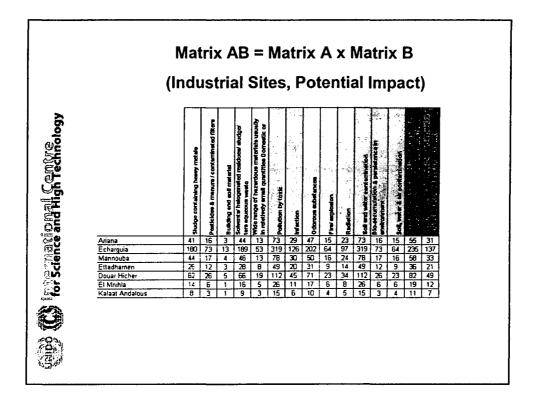


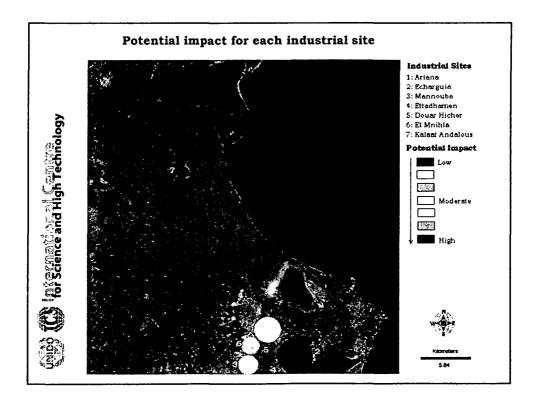


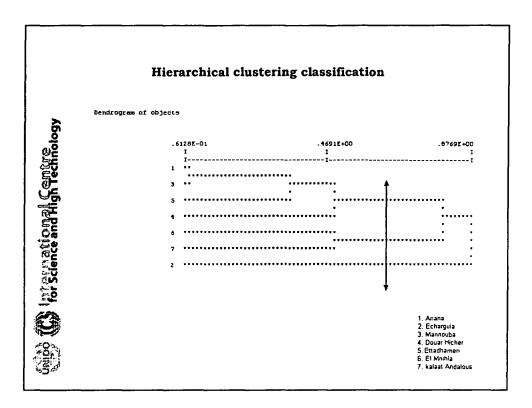
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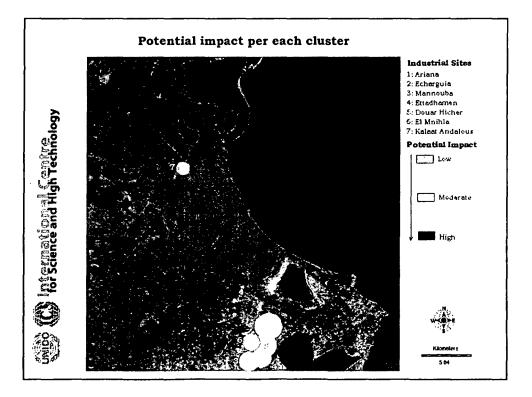


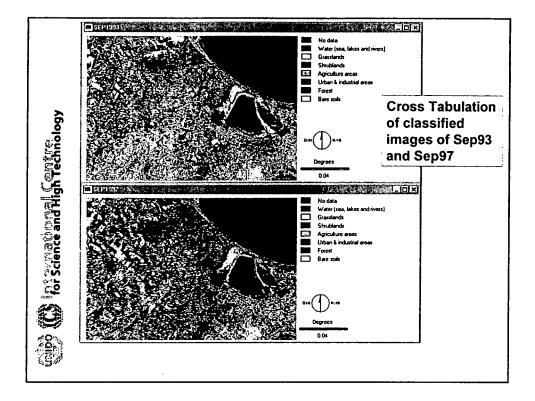




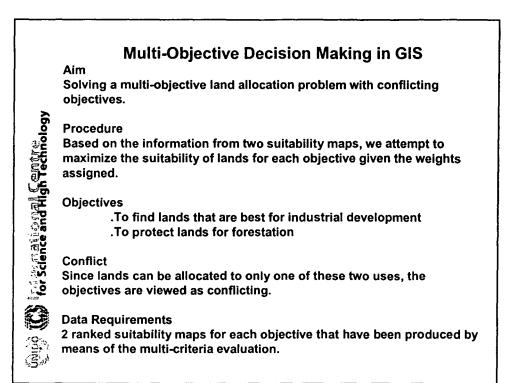


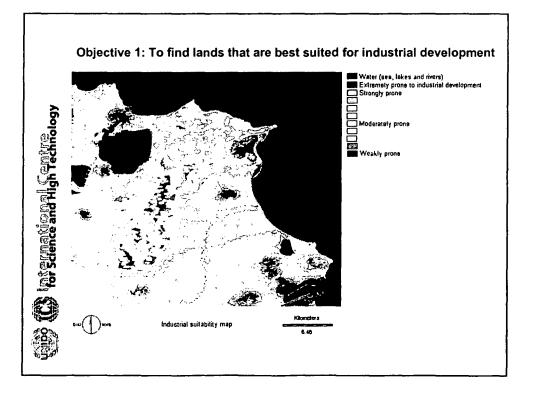


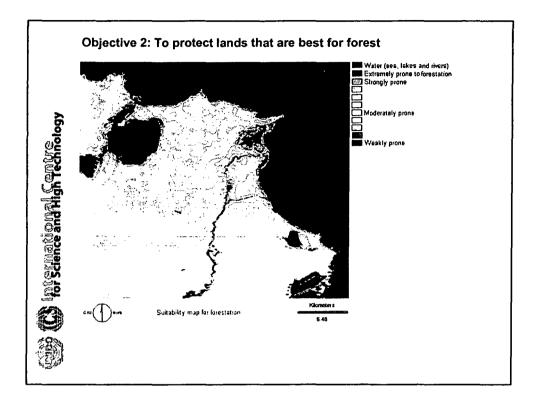


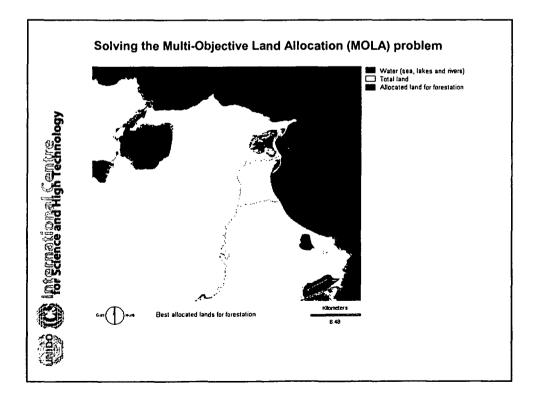


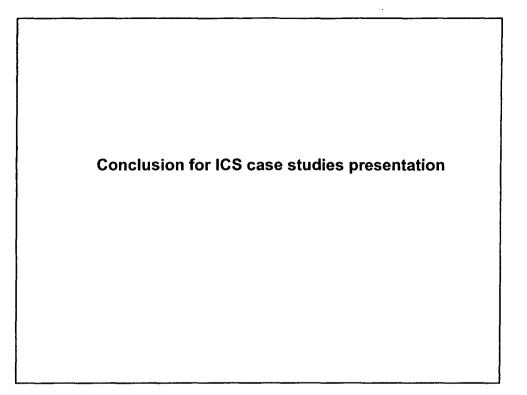
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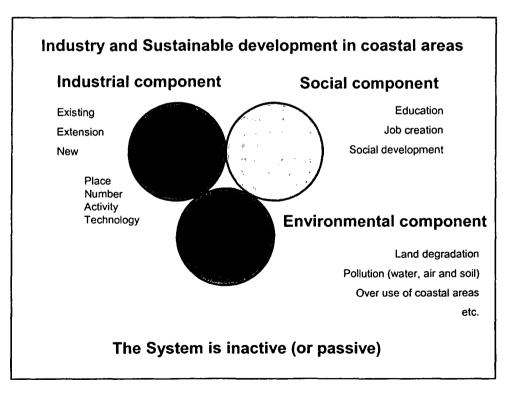


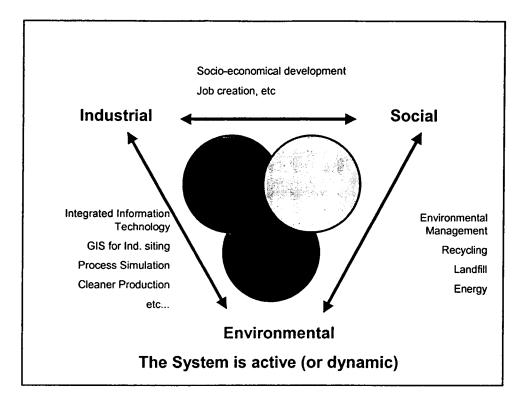


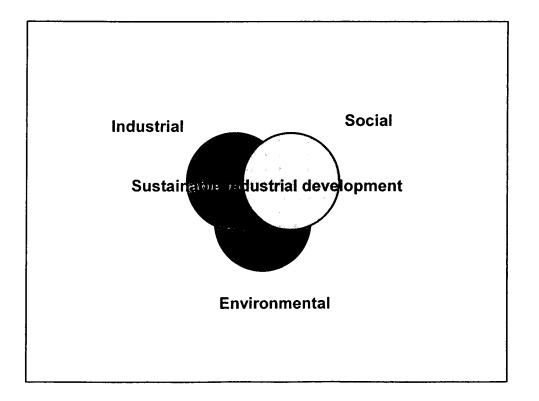


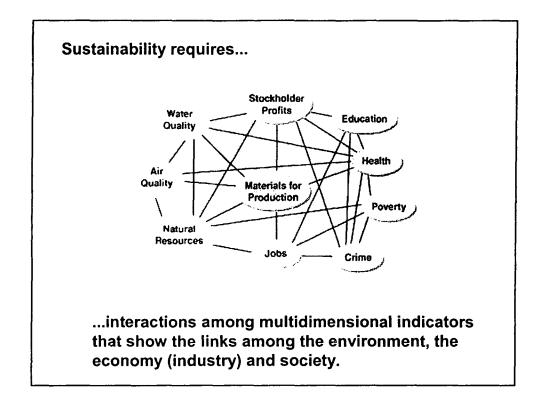


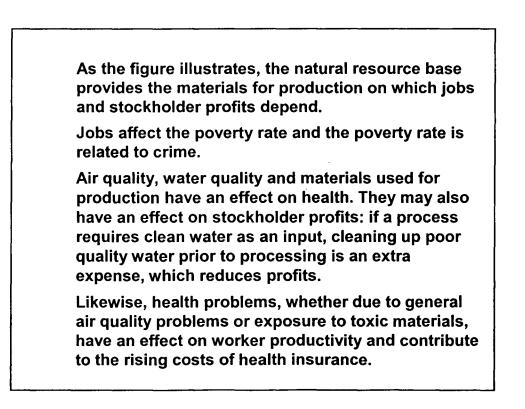












Application of Classification and Ordination Methods in Landscape Assessment by E. Feoli

Application of Classification and Ordination Methods in Landscape Assessment

Draft for ICS-UNIDO Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas" Iglesias-Cagliari, Italy, 4-8 June 2002

by

Enrico Feoli ICS Senior Scientific Consultant Earth, Environmental and Marine Sciences and Technologies ICS-UNIDO AREA Science Park, Bldg. L2 Padriciano 99, Trieste, Italy Tel.: +39-040-9228109 Fax: +39-040-9228136

Abstract

The data describing the landscape system in terms of pattern and/or landscape elements are giving rise to multidimensional spaces. For understanding pattern and processes in the landscape and to model them it is necessary to represent in efficient way such multidimensional spaces by few dimensions. Exploring multidimensional spaces on the basis of spectral and singular value decomposition theorems (SVD) and eigenanalysis (EA) is an exercise that can be useful to get information on landscape patterns and can be used for modeling complex systems in iterative way without feeling too much restricted to precision problems. The idea of modeling clusters of points, rather than lines is suggested. Fuzzy set theory can be used as a useful data interpolation method and data transformation before EA.

1. Introduction

Some definitions are necessary before to start to discuss on the application of singular value and spectral decomposition theorems (SVD) and eigenanalysis (EA) that are used and could be used to explore the multidimensional spaces generated by matrices describing the landscape system:

1) *Landscape* is a complex system with states determined by all the interactions between man, plants, other living organisms and the chemical-physical environmental factors.

2) Landscape ecology is the discipline aimed to understand all the aspects of landscape system concerning its composition (diversity), structure (spatial pattern and spatial relationships), functions and dynamics (the ways and the the causes of landscape changes in space and time). It uses the concept of

multidimensional space (*ecological space*) defined by all the ecological factors influencing the human life and the life of the other organisms.

3) Landscape is sampled and described by Operational Geographic Units (OGU). They are sampling units, which are supposed to represent in a satisfactory way a landscape at a given time in a given place. These sampling units may be defined by the user in a GIS procedure, or defined automatically by remote sensing techniques (pixels). The variables that are sampled and described in each OGU may be of different nature such as spectral bands, environmental chemical-physical factors and/or human and other biological factors that are supposed to influence the landscape states. OGUs can be sampled following some sample design or by preferential sampling as it may happen in vegetation science (Orloci 1991a,c; Orloci and De Patta Pillar 1991). It must be clear that an OGU is an operational concept, which is useful to integrate in a GIS the remotely sensed data with other multisource data.

The concept of Operational Geographic Unit (OGU) was defined as the basic cartographic element for understanding biogeographical patterns by classification and ordination techniques (Crovello 1981). It is analogous to the concept of Operational Taxonomic Unit in numerical taxonomy (Sneath and Sokal 1973), with the only difference that OGU is a georeferenced unit. The concept of OGU has been already introduced in GIS context by Feoli and Zuccarello (1996), to describe ecological processes in space and time. OGUs may be used to store information in data bases following what Peuquet (1984) calls geographical data models, however we prefer to use the concept to integrate and extract georeferenced data already stored in GIS data bases. The data integration is achieved when each OGU is described by remotely sensed data (average values, class frequencies, min-max values, etc. of bands) and by other lanscape multisource data not directly attributable to the single pixels (e.g. population density, urbanisation density, yeld production, biodiversity, etc.), usually called ancillary data. OGUs can be obtained by a GIS in a very flexible way. They, as adjacent or non adjacent polygons (or points), may be easily defined by administrative boundaries or by superimposing to the maps regular or irregular tessellated structures (grids, triangulated networks, quadtree patterns, hexagonal mesh, thiessen polygons, etc., see Peuquet, 1984; Skole et Al. 1993).

4) OGUs that are considered similar enough are grouped into the same land cover type or class, which is used for cartographic purposes. A GIS can easily shows the results of OGUs classifications by maps showing the spatial distribution of the classes of OGUs. Since these classes may be characterized by the character states, used for the OGUs classification, the maps shows the spatial pattern of those character states in the landscape. In this sense OGUs are useful tools of the Dobson's "macroscope".

5) In a given landscape a land cover type is spread within the geographic ranges of the combination of characters that are chararacterising it (Characteristic variables) as a different type from other land cover types present in the landscape. The extension and the shape of the area occupied by a land cover type depend on the extension of the environment favourable to the combination of its characteristic variables in the geographical space. The environment hosting a land cover type defines, in analogy with the community niche concept (Feoli, Ganis and Zerihun 1988), the land cover type niche. However the concept of niche is not only related to environmental variables, if socio-economic variables are only used the multidimensional space will represent a socio-economic multidimensional space and the niche of a land cover type will be a socio-economic niche. The niches may be wide or narrow in the multidimensional space irrespective its extension in the geographic space, the niche width depends on the heterogeneity of the land cover type.

6) Once defined the land cover types on the basis of a set of characters (variables), landscape can be studied on the basis of other sets of variables. The classification of landscape based on a set of land cover types can be used as a reference system on which to overimpose other kind of classifications or other information.

7) Land cover type niches are partially or completely represented in a matrix X given by a set of OGUs (almost always disposed as column vectors). Matrix X represents partially the multidimensional ecological space (it is just a sample of it). In this each variable has a number of co-ordinates equal to the number of OGUs and each OGU has a number of co-ordinates equal to the variables that are describing it. Variables and OGUs may be seen as vectors with a common origin, namely the origin of the space. Their length and direction are determined by their co-ordinates in X. The dimensionality of the space represented by X is, however, equal or lower than the minimum number among the number of OGUs and the number of variables.

The dimensionality (rank of the matrix) is, in fact, the minimum number of orthogonal axes that are able to get the projections of all the OGUs and all the variables. Only the variable or OGU with all the co-ordinates equal to the co-ordinates of the origin of the space do not have axes where to be projected. These orthogonal axes are the reference system for variables and OGUs. They give the minimum number of co-ordinates necessary to correctly positioning variable and OGUs in the multidimensional space. These axes are also used to represent the space by a reduced number of dimensions. Their efficiency in doing this is measurable. X may be divided into sub-matrices on the basis of the kind of variables used to describe the OGUs: e.g. X(1) matrix of biological variables, X(2) matrix of environmental variables, X(3) matrix of socio-economic variables, X(4) matrix of human impacts, etc.

Each matrix X is the matrix representing the states of landscape system according to some set of variables. To study the correlation between different matrices is possible by many mathematical methods.

2. Spectral and Singular value decomposition theorems

The spectral and singular value decomposition theorems of matrix algebra are generally mentioned in the Appendix of multivariate data analysis textbooks addressed to applications (Jongman, Ter Braak and Van Tongeren 1995). However it would be more appropriate to present them at the beginning of these books since they are essential for understanding why multidimensional space of a given matrix X may be represented by few orthogonal dimensions. The theorem of spectral decomposition says that a symmetric matrix S of rank p can be decomposed by the following three matrices:

 $S = T \Lambda T'$

where T is an orthogonal matrix (mxp), i.e. $T_i T_h = 0$, Λ is a diagonal matrix

(pxp) and T' is the transposed of T. Λ is the matrix of the p positive eigenvalues of S that are disposed in decreasing order. T is the matrix of standardized eigenvectors of S.

 Λ may be calculated by the characteristic equation of eigenvalues:

$$det (S-\lambda I) = 0$$

where *det* means the determinant of the matrix (S- λ I), λ is a scalar and I is the identity matrix (i.e. a matrix with ones along the diagonal and zeros outside). Out of this equation only the positive p solutions are considered in the diagonal matrix Λ . Once calculated the p positive eigenvalues (λ i, i=1,...,p), T may be obtained by solving p times the system of equations:

 $(S-\lambda i I p) T_i = 0$ ⁽²⁾

The singular value decomposition theorem says that given a rectangular matrix X of m rows and n columns (mxn) of rank p it can be decomposed in the following product of three matrices:

 $X = U\Lambda V'$

(3)

(1)

where U is an orthonormal matrix (mxp) (U'U = Ip, identity matrix), V is an orthonormal matrix (nxp) (V'V=VV'= Ip) and Λ is a diagonal matrix (pxp) with positive elements.

If we consider that:

 $X'X = (U\Lambda V')'U\Lambda V'$

it follows that

 $X'X = V\Lambda' (U'U) \Lambda V'$

and since U'U = Ip it follows that

 $X'X = V\Lambda^2 V'$

on the other hand

 $XX' = U\Lambda^2 U'$

and since both X'X and XX' are symmetric matrices (S) the theorem of spectral decomposition holds true. From this it follows that the eigenvalues of the two symmetric matrices are the same.

The relevant implications of these two theorems for exploring multidimensional spaces of X (or X(1), X(2), X(3), etc.) are the following:

1) Each numeric matrix X, i.e. a matrix for which it is possible to apply the product matrix (X'X) or (XX'), may be represented by two sets of orthogonal axes that may be used to order column vectors (objects) and the row vectors (variables) simultaneosly by biplot graphs (Gabriel 1971; ter Braak 1995). The efficiency of the representation is given by the ratio between the sum of the eigenvalues corresponding to the eigenvectors used for ordination and the total sum of eigenvalues;

2) For any numeric symmetric matrix S obtained in some way from X it is possible to compute its eigenvalues and eigenvectors by equation 2). T_i may be adjusted in several ways, e.g. $T_i'T_i = 1$ or $T_i'T_i = \lambda i$ or $T_i'T_i = \lambda i^{1/2}$;

3) If S is the symmetric matrix for the rows, T_i is an axis of ordination for the elements in rows. If S is a symmetric matrix for the columns, T_i is an axis of ordination for the elements in columns. For better clarity we can indicate with R the symmetric matrix for rows and with Q the symmetric matrix for columns (in analogy with the well known terminology R-mode and Q-mode algorithms, ter Braak 1995), and respectively we can indicate by α_i the eigenvectors of R and by β_i the eigenvectors of Q. Only in some circumstances the matrix of the β_i can be computed by multiplying the matrix of the α_i by X. This happens when R is obtained by XX'. When Q is obtained by X'X, α_i can be obtained by multiplying X by β_i (Orloci 1967, 1978). Not always the R and Q matrices can be obtained by scalar products between row or column vectors. If the matrix X is ordinal or alfanumeric, other measures of similarity or correlation (association) rather scalar products should be applied, e.g. Gower's and/or Goodall's and Burnaby' s indices (Goodall 1964; Dale 1991; Carranza et Al. 1998). In this case only the SVD of the symmetric matrix makes sense.

3. Ordination methods and gradient analysis

In landscape analysis, as in ecology, the methods able of positioning objects (OGUs, or variables) in a reference system of orthogonal or obliques axes are called ordination methods (Orloci 1966, 1967, 1978, ter Braak, 1995). Ordination methods are used to display the relationships between variables and objects with the aim to discover clusters, trends of variation (gradients) and/or trajectories. More in general they are used to study patterns (spatial, temporal,

structural) of landscape and gradients, i.e. the way in which landscape is changing in function of environmental factors. Gradient analysis is playing an important role in landscape ecology. Its primarily aim is to reconstruct the sequence of response curves of variables along single environmental variables (primary or secondary ecological factors), or combination (linear or non-linear) of them. Two methods of gradient analysis have been so far distinguished: direct gradient analysis and indirect gradient analysis (Whittaker 1967, Whittaker 1978, Prentice and ter Braak 1988).

- Direct gradient analysis

In direct gradient analysis two matrices of data are considered: the matrix of independent variables (causal or explanatory variables, e.g. chemical-physical variables in ecology) X(2), and the matrix of dependent variables of our interest that have been used to describe landscape features X(1). Direct gradient analysis may be done by two approaches: one considers only X(2) and uses it to order X(1) the other considers both the matrices simultaneosly.

Let us consider the first approach. The following procedures are used:

1) X(1) is rearranged (ordered) according to environmental axes given by the single environmental variables or according to some combination (linear or non-linear) of all of them or only some of them. X(2) is subjected to SVD after having standardized the variables. The T axis are used to rearrange X(1). If according to the scattergram of T axes, a curvilinear pattern of points is evident, than the curve interpolating the points, or an angular seriation, can give the non-linear combination of the variables in X(2) (Feoli and Orloci 1991). The new axis so obtained is used to plot the response values of the variable.

2) Graphs are produced by simply plotting the variable scores against the environmental axis, and/or by curves fitting the scores by regression analysis (ter Braak 1995).

3) A model of response curves is described and used for predicting landscape composition all along the gradient.

In the second approach X(2) and X(1) are used simultaneously following two main procedures. In the first one the two matrices may be fused in one with m+s rows and n columns, where m (number of environmental variables) are the rows of X(2) and s (number of variable) the rows of X(1). This matrix (X) after standardization of row vectors may be decomposed according to SVD. This gives the simultaneous ordination of variable, environmental variables and OGUs. This procedures includes the methods known as Principal Component Analysis (PCA) and Correspondence analysis (CA). They differ only because in PCA the variables are centered and/or standardized ($z_{ij} = (x_{ij}-\xi_i)/\sigma_i$), where ξ_i is the mean of i-th variable and σ_i is its standard deviation, while in CA the variables are transformed according to the deviation of the scores from the expected scores calculated as the product of the marginals (row totals and column totals) divided by the grand total of the matrix X.

In the second procedure the two matrices are kept together by multiplying X(2)by X'(1) in order to obtain a matrix of scalar products C(mxs). When the s variable and the m variables are standardized by the z variable this matrix is a non-symmetric correlation matrix. C(mxs) can be decomposed by SVD. The resulting matrix U' of p eigenvectors coresponding to the p positive eigenvalues may be multiplied by X(2) (standardized by rows) in order to obtain $\beta(2)$, namely the p ordination axes for the n OGUs. This gives ordinations based on the m variables under the constraint of their correlation with the s variable. The multiplication of matrix of V' by X(1) (standardized by rows) gives $\beta(1)$, namely the p ordination axes of the n OGUs based on the s variable under constraint of their correlation with the m variables. Scattergrams can be obtained by using $\beta(2)$ or $\beta(1)$ and also by using $\beta(2)$ against $\beta(1)$, i.e. j-th axis of $\beta(2)$ against j-th axis of $\beta(1)$. This type of scattergrams show the pattern of correlation between X(2) and X(1). The methods of multivariate data analysis known as canonical correlation analysis, canonical correspondence analysis and redundancy analysis are based on this procedure. They differ only by the way in which the variables are transformed or combined into the two matrices before the multiplication in order to get C. In terms of equation $(S-\lambda i I p)T_i = 0$ they differ only by the way in which S is computed based on C. In canonical correlation analysis:

 $\mathbf{S} = \mathbf{R}_{11}^{-1} \mathbf{R}_{12} \mathbf{R}_{22}^{-1} \mathbf{R}_{21}$

for obtaining $\beta(1)$, $\beta_1(1) = T_1 X(1)$

or

 $\mathbf{S} = \mathbf{R}_{22}^{-1} \mathbf{R}_{21} \mathbf{R}_{11}^{-1} \mathbf{R}_{12}$

for obtaining $\beta(2)$, $\beta_1(2) = T_1'X(2)$

where \mathbf{R}_{11} is the correlation matrix between the variable in X (1), \mathbf{R}_{22} between the variables in X(2) and \mathbf{R}_{12} or \mathbf{R}_{21} is the C matrix. \mathbf{R}_{11}^{-1} and \mathbf{R}_{22}^{-1} are respectively the inverse matrices of \mathbf{R}_{11} and \mathbf{R}_{22} .

In redundancy analysis \mathbf{R}_{11} is the identity matrix. In Canonical correspondence analysis the data are transformed as in correspondence analysis and centered, then the equation for getting $\beta(2)$ is equal to that of canonical correlation analysis (ter Braak 1995).

If the variables cannot be standardized, because measured in different scales (e.g. variable in ordinal scale and environmental variables in ratio scale), the two matrices have to be both transformed in simmetric matrices (resemblance matrices), namely Q(1) and Q(2), by suitable resemblance functions (Dale 1991, Podani 1991a). Both the matrices can be decomposed by SVD. $\beta(1)$ and $\beta(2)$ are

respectively the matrices of eigenvectors of Q(1) and Q(2). X(1) can be ordered and response curves of variables detected by using $\beta(2)$ as ordination axes (first procedure above). In order to get the second procedure, i.e. constrained ordination, $\beta(1)$ is multiplyied by $\beta(2)$ and the resulting matrix decomposed by SVD. In this case the constrained ordinations axes are obtained respectively by multiplying U' by Q(1) and V' by Q(2).

The proper direct ordination is always the one based on X(2). To constrain the ordination by matrix X(1) it means to keep into consideration all the factors that have determined the matrix X(1) included those in X(2). The costrained ordination axes of X(2) obtained based on matrix C(mxs) are always axes that in some way maximize their linear correlation with the axes obtained from X(1), irrespective the type of relationship between the variables. Scattergrams obtained by plotting $\beta(1)$ axes versus $\beta(2)$ axes may show linear correlation between variable, between environmental variables and between variable and environmental variables may be non-linear.

In direct gradient analysis constrained ordination axes can be obtained by Procrustes analysis (see Podani 1991b and Feoli, Podani and Sun 1993, for applications and for references). By it, given k axes of $\beta(2)$, in a matrix called for convenience X(kxn), and k axes of $\beta(1)$, in a matrix Y(kxn), the k axes of $\beta(2)$ are transformed in such a way to fit as much as possible the configuration of OGUs in the space defined by Y. This means that the new axes of X are adjusted in such a way to "absorb" the effects that other environmental factors have and/or have had in defining the ordination of n OGUs in $\beta(1)$.

The required rotation of X is obtained by its multiplication with the nxn (n=number of OGUs) orthogonal transformation matrix:

H (nxn)=V(nxp)U'(pxn)

where **U** and **V** result from SVD of **Y'X** (nxn)=**U**(nxp) Λ (pxp) **V**'(pxn)

Rescaling parameters should be necessary for measures the goodness of fit of the rotation (Podani 1991b). This would be useful to measure what variables in X(2) are better related to X(1) in a step-wise process followed according to some suitable criterion.

- Indirect gradient analysis

Indirect gradient analysis is based on the assumption that the variables describing the landscape are good indicators of the landscape states. It is assumed that the landscape pattern is determined by all the environmental and historical factors that can be and that cannot be measured and that the pattern is indicating both the environmental situation and the environmental impact of human activities.

Indirect gradient analysis uses only the matrix X(1) to get ordination axes. The explanatory matrix X(2) is only used to try to explain the pattern of X(1). Indirect gradient analysis does not require that the explanatory matrices have the same number of colums of X(1). Their number of columns can be less that the number of columns in X(1). The ordination axes obtained by SVD of X(1), or of Q matrices obtained by different resemblance functions applied to X(1), are used for detecting gradients and for what ter Braak (1995) calls calibration, i.e. the prediction of scores of explanatory variables. In this case X(2) is rearranged according the ordination axes obtained by X(1). Along these axes, or along nonlinear combination of them (projections of OGUs to curves or angular seriations) the scores of variable and/or those of environmental variables are plotted (Feoli and Feoli Chiapella 1980).

4. Ordination based on classification and fuzzy set theory

The ordination methods based on SVD and eigenanalys of a matrix X (or of its submatrices obtained by separating groups of variables) are generally maximizing dispersion or correlation. When the OGUs are classified into groups by explanatory variables, a direct ordination can be obtained by Discriminant Analysis. In terms of equation 2), Discriminant Analysis (DA) is:

$(W^{-1}B - \lambda i I p)T_i = 0$

where W^{-1} is the inverse of the matrix of covariance of the variables within the sets of OGUs and **B** is the matrix of covariance of the same variables beween the sets of OGUs.

 $\beta(2)$ ordination axes are obtained by $\beta(2) = T'X(2)$. These axes maximize the separation between the sets of OGUs. Discriminant analysis allows to find what are the variables that better discriminate the OGUs sets and also gives the probability of each OGU to belong to the sets under the assumption of normal distribution, linear correlation and equality of variance-covariance of the variables within the clusters. All these limitations are rarely met in data matrices, however they can be tested and the tests can be useful to understand landscape patterns rather than to prove hypothesis (Feoli and Lagonegro 1991). Discriminant analysis (DA) can be also used as an indirect ordination method when W⁻¹ and **B** are obtained by the dependent variables. To avoid singularity problems, the number of variable should be always less the number of OGUs in the sets. This is rarely met. A way to overcome this problem is to group variable in sets before DA. In this way OGUs are described by variable groups (average scores of the variable in each group).

Another method of ordination based on classification is the one proposed by Feoli and Zuccarello (1986, 1988). This is based on fuzzy set theory since the ordination axes are given by the degree of belonging of OGUs to sets of OGUs. According to fuzzy set theory a fuzzy set is given by a set of elements with their degrees of belonging to the set (see Feoli and Orloci 1991 for references). In this method the similarity matrix \mathbf{Q} is considered a fuzzy matrix, where each OGU

represents a set. The similarity measures (ranging between 0 and 1, with 0 no similarity, 1 identity) in the i-th row or j-th column (i=j, being the matrix symmetric) are the degree of belonging of all the OGUs to the set represented by i-th OGUs. On the other way, they are also the degrees of belonging of the ith OGU to all the sets represented by all the other OGUs. Fuzzy axes can be obtained by averaging the values of similarity in a matrix of similarity after having rearranged its rows and columns according to some classification. The **Q** matrix becomes a rearranged Op(nxn) that is transformed in a Op(nxc) matrix where c is the number of clusters (sets). The c clusters can be obtained by any classification method, provided, and this is essential for fuzzy set theory, they are well characterized, i.e. "conceptualized". Each set has to be described by variables that have been used for classification or that have been added later. For doing this it is essential to rearrange the original data matrix X according to $\mathbf{Q}_{p}(\mathbf{n}\mathbf{x}\mathbf{n})$ and discovering what are the discriminating variables. Once discriminat variables have been defined the sets are characterized according to them. In this way each j-th vector of the c vectors in $\mathbf{Q}p(\mathbf{nxc})$ is associated to a variable or to a set of variables higly correlated and represent the degree of belonging of all the OGUs to that set, each j is a fuzzy set. If the values in $Q\rho(nxc)$ are adjusted to the row totals in order to get $\Sigma_i \xi_{ii} = 1$ (with $\xi_{ii} = degree$ of belonging of i-th OGU to j-th fuzzy set) $Q\rho(nxc)$ represent a fuzzy partition, let call it $F_{\rho}(nxc)$. SVD of $F_{\rho}(nxc)$ gives the representation of the fuzzy system space where the n OGUs and the c fuzzy sets find their mutual position. This method of ordination can be applied to any kind of matrices and it is not confined, as DA, to variables measured in ratio scale. When the data are mixed (alfanumeric, ordinal, etc.) Q matrices can be obtained by suitable similarity measures that are not based on scalar product, such as Gower, Burnaby or Goodall's similarity indices. Direct gradient analysis can be done using the matrix $\Phi(nxm)$ obtained, as was suggested by Feoli and Zuccarello (1988) by multiplication $F_{\rho}(nxc)E'(cxm)$. $\Phi(nxm)$ is a matrix that describes the OGUs by the variables in E(mxc). There are many ways to put values in E(mxc): average values, maximum values, the most probable value, etc. The result in $\Phi(nxm)$ depend on the way chosen. $\Phi(nxm)$ represents a new description of all the n OGUs, (notwithstanding data were not collected for all the OGUs in E) in this sense fuzzy set ordination can be considered an interpolation method in the fuzzy system space. If the values in $\Phi(nxm)$ are adjusted in such a way that the row totals are equal to 1, than $\Phi(nxm)$ is a fuzzy partition with m environmental fuzzy sets. Each column represents the degree of belonging of OGUs to the sets defined by the j-th environmental variable. The SVD of $\Phi(nxm)$ represents the fuzzy system space where OGUs and environmental variables find mutual position. Once defined a fuzzy partition, let say Fp(nxc), many other fuzzy partitions can be obtained by matrix multiplication. Keeping the c fuzzy sets as axes all the objiects that may belong to the c sets (environmental variable, morphological characters, chorological, etc.) can be overimposed into a single ordination, examples are given by Feoli and Zuccarello (1988, 1992, 1994) and Banykwa, Feoli and Zuccarello (1990).

5. Conclusions

Mapping multidimensional spaces described by X matrices is a good exercise to study niches of land cover types and to study relationships between variables or sets of variables and between them and land cover types. Data analysis is often criticised for not being enaugh explanatory: elements of deductive approach are considered missing and the methods are considered not predictive. This criticism is not realistic. It is true that the mechanistic element in landscape studies is weak, but it is also true that cannot be stronger owing to the complex nature of landscape. Landscape is a self -organizing system, objects or entities that can be described by several variables (characters) are in a certain area for historical and ecological reasons that cannot be modelled in a mechanistic way. Shapes of features are the results of many interactions. Based on SVD we are able to model clusters in multidimensional space by creating new data by interpolations or extrapolations in the space defined by V in $X = U\Lambda V'$. Iterative calibrations according to ter Braak (1995) can be done accordingly. The space defined by V based both on real or fuzzy data, can be used to find the position and width of the ecological niches of variable and communities in the ecological space (Zerihun et Al. 1989) and can be used for detect or simulating gradients by direct or indirect methods.

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Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas" Iglesias-Cagliari, Italy

glesias-Cagliari, Ital 4-8 June 2002

Enrico Feoli

Multicriteria Analysis:

Establishing priorities

(notes from the book of T. Saaty 1999 Decision making for leaders: The analytic hierarchy process for decisions in a complex world. RWS Publications University of Pittsburg, 322 Mervis Hall)

Calculating the priorities using an approximation method

Let have a Saaty's matrix of pairwise comparison between three criteria:

		С	Т	L
	С	1	1/2	1/4
	Т	2	1	1/2
	L	4	2	1
Column total		7	3.5	1.75

A normalized matrix is computed by dividing each element in the matrix by the corresponding column total

	С	Т	L
С	1/7	1/7	1/7
Т	2/7	2/7	2/7
L	4/7	4/7	4/7

The priority of each criterion is calculated by averaging the row scores of the normalized matrix

In the present example it is:

Priority C = (1/7+1/7+1/7)/3=1/7=0.14Priority T =(2/7+2/7+2/7)/3=2/7=0.29Priority L = (4/7+4/7+4/7)/3=4/7=0.57

The concept of consistency:

In the example we can see that C=1/2 T and that C=1/4 L so that 1/2T = 1/4 L which brings T=1/2 L that is what we have in the Saaty matrix. This proves that the matrix is consistent. Another proof: of consistency is given by the fact that L is 2 T, and C is 1/2 T so L should be 4 C, that is actually what happen in the Saaty matrix.

Calculating the priorities using the exact method

This is based on the idea of consistency. If the matrix is consistent all its powers (multiplication of the matrix by itself 1 time = power 2, 2 times = power 3, etc.) give the same priority or dominance pattern.

When the matrix is consistent, the normalized sum of each row tells us how much each element dominates the others in relative terms. The sum of the entries in each column tells us how much each element is dominated by other elements.

To derive the priorities from the matrix we add the numbers in each row and divide each of the results by their total sum to obtain the normalized scores. The operation is repeated till results are stabilized. (In the approximation method each value in the row is divided by the sum of the corresponding column!). The resulting vector is approximating the first eigenvector of the matrix, i.e. the eigenvector corresponding to the first eigenvalue of the matrix.

If the matrix is consistent the product of the vector of the sum of the elements in columns by the vector of the normalized scores is equal to n, where n is the number of elements (criteria). If the product is higher then n it means that the matrix is not consistent.

The inconsistency is assumed to be generated when the scores in the pairwise comparison matrix the values are generated by random.

The ratio CI= (eigenvalue - n)/n-1 is the consistency index, the consistency ratio is given by the ratio between CI and CIR where CIR is the consistency index corresponding to a completely random judgement (there are tables for this!!!). If the consistency ratio CR=CI/CIR is less than 0.05 the matrix of Saaty can be considered good!

There are computer programs for calculating priorities, CI and CR. IDRISI is calculating them automatically within a GIS environment. EXPERT CHOICE is doing out of a GIS environment.

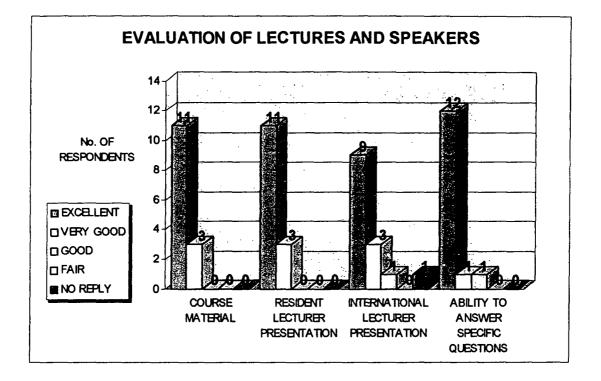
Focus on the Impacts of Industrial Development on Marine Environment and Landscape by N. Candeloro

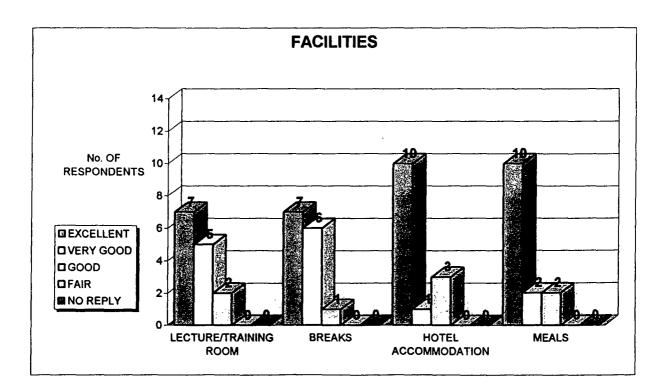
Focus on the Impact of Industrial Development on Marine Environment and Landscape by Nicola Candeloro

At the end of second world war, more than 50000 of people worked into the lead and zinc mines of the areas of Sulcis-Iglesiente. When the Italy State has opened its market to abroad trade, the lead and zinc production became not economic convenient. The companies started to reduce the employees and, in the same time, the unemployment was going to increase in the area. The Unions and the local authorities required an alternative to the mine production to reduce the level of unemployment. The government decided to transform the land of Portoscuso, a village of farmers and fishermen, in industrial zone. The state companies havebeen required supporting the government decisions. They designed and set up the Alumina refinery, aluminium smelter, lead and zinc refinery and the facilities to support the activities of plants. The chimneys of plants replaced the vineyards. More of 5000 people were employed during the phase of the construction and the same number were involved to run the plants. Fishermen and farmers have been transformed in workers. Portoscuso transformed itself and in few years the residents increased from 500 to 7000. At the beginning all the people were satisfied of the situation. The problems started when the company reduced the number of employs and the environment issues became actual. During the first phase of production the waste material of alumina and lead and zinc refineries has been put in the open pond. When the resident of Portoscuso became to protest against the companies, the authorities imposed the construction the new ponds and reduced the limits to the emissions and the area has been declared "Area ad alto rischio ambientale". This declaration imposed to the companies the recovery of the polluted areas. Actually the companies are involving to burn the low sulphur oil, cover the stock pile to reduce the dust, increase the efficiency of process to reduce the off sets and design the recovery of polluted areas. In the same time it is nearly impossible to construction new ponds. The future of plants of areas is uncertain and depends of the public opinion that prefers now the tourism to the industry. The history of Portoscuso shows it is very important to consider the interest of the stakeholders before those of shareholders in marine environment.

Annex IV STATISTICS Training Course on "Image Processing and Remote Sensing for Monitoring Industrial Development in Coastal Areas" Iglesias, Cagliari, Italy 4-8 June 2002

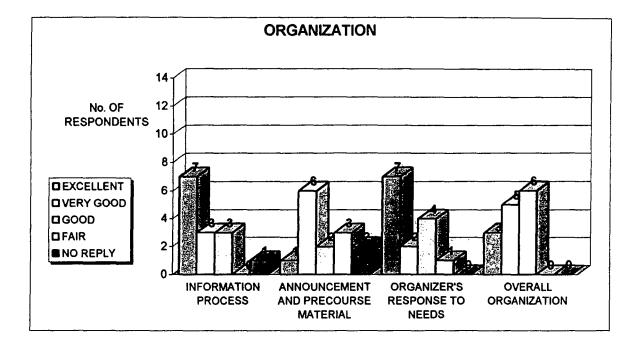
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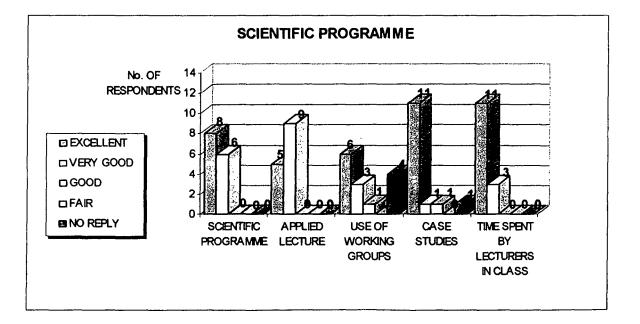




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EVALUATION





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EVALUATION

