

KES: Knowledge Enabled Services for better EO Information Use

Andrea Colapicchioni Advanced Computer Systems Space Division a.colapicchioni@acsys.it



The problem

- During the last decades, the satellite image catalogues have stored huge quantity of data
- State of the art catalogues permit only to specify location, time of interest, metadata like platform, sensor, acquisition mode...



The interpretation task

- The interpretation of EO images requires
 - Fusion of data/information for better understanding of structures
 - Aggregation with existing knowledge specific to the application fields (at higher level)



A little bit of history..



- 1996-2000 IIM (Image Information Mining) (<u>http://isis.dlr.de/mining</u>)
- 2001: KIM (Knowledge Information Mining) (<u>http://www.acsys.it:8080/kim</u>)
- 2002: KES (Knowledge Enabled Services)
- 2003: KIMV (KIM Validation)
- 2004: KEO (Knowledge-centred Earth Observation)



KIM: Knowledge Driven Information Mining





More in detail





KIM Interactive learning

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From data to semantic





From KIM to KES (Knowledge Enabled Services)

- Image interpretation is not a simple task. Each user needs a set of accessory data, as for example GIS layers or texts obtained through Internet.
- Yet, the amount of available information makes searches a demanding and expensive task.
- An environment where images are at the focal point, and where each user can navigate through a taxonomically structured knowledge, could be of extreme value.





IGARSS 2004 - Image Information Mining



KES: New interface

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Semantic Grouping

- In KIM it is simple to define a feature by identifying a river through positive and negative examples.
- However, the "river" might become part of a wider concept (e.g.: water). This should be implemented without retraining the system for all possible water types.



Aggregated Features

- In KES a new kind of grouping, called "aggregated features", has been introduced.
- In a similar way as in defining positive and negative examples on the image, it will be possible to define the concept "water" as:
- Positive examples:
 - sea + river + lake + water reservoir
- Negative examples:
 - streets + houses + mountains





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Ontology

 Ontology is the specification of a conceptualisation. It can be related to a system (System Ontology, which can be domain independent and reused for different domains) or to a domain (Domain Ontology, specific for that domain).



Domain Ontology

- Domain ontology is the set of definitions and concepts pertaining and belonging to a specific domain (and shared by concerned people).
- Different domains have generally different ontologies. As an example, a climatology expert could have a different vision of (and terms to describe) water compared to that of an oceanography expert.







Data / Semantic / Ontology



Implicit and explicit knowledge transfer

- The KIM prototype permits to associate weighted combinations of primitive features to image features. While defining features, the user explicitly transfers knowledge to the system, which is stored and made available to the same or other users.
- In addition there is a further type of knowledge (implicit) that could be discovered: if the user domain of interest is known, by observing the user interactions with the system during search and browsing, it is possible to infer the data of likely user interest and link it with the pertinence domain.



Knowledge transfer





Knowledge Graph





Knowledge Discovery

Exploring this graph means discovering the user's knowledge



KIMV (KIM Validation)

A practical case study: Automatic cloud classification in MERIS images



Cloud Cover Characterization



MERIS Level 1 – Reduced Resolution





Cloud cover characterization



Map of Cloud object







MAP Object Extraction



Binarized, closed MAP of "cloud" label



Tiling





MASS

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Linux Cluster



- Cluster nodes: 10 Dell 1750, dual Xeon 3 GHz
- Database and Application Server: Dell 6600, 4 Xeon 2.8 GHz



System Context





KIM prototype

http://www.acsys.it:8080/kim

Andrea Colapicchioni
<u>a.colapicchioni@acsys.it</u>

Visit the ACS stand for a demo