

Selection of Reliable Machine Learning Algorithms for Geophysical Applications

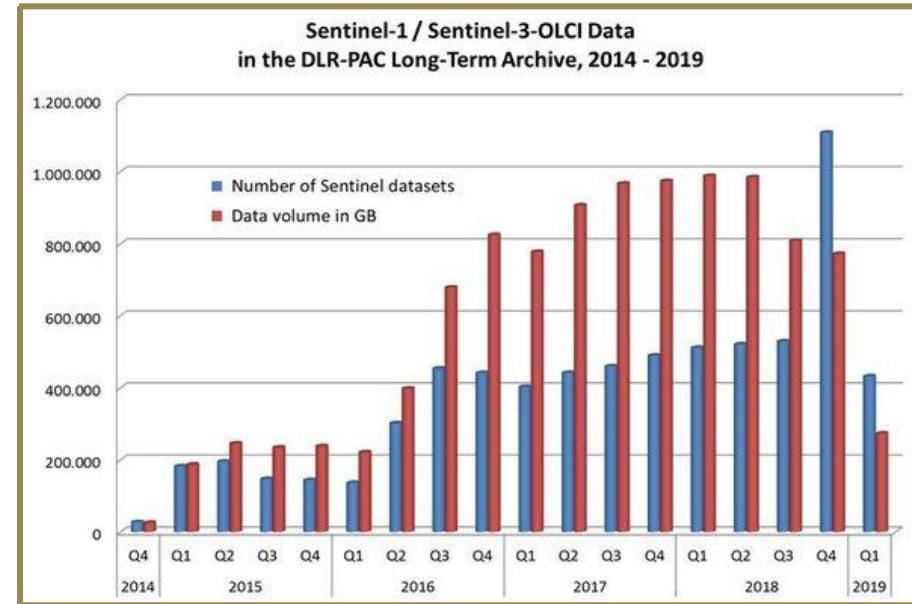
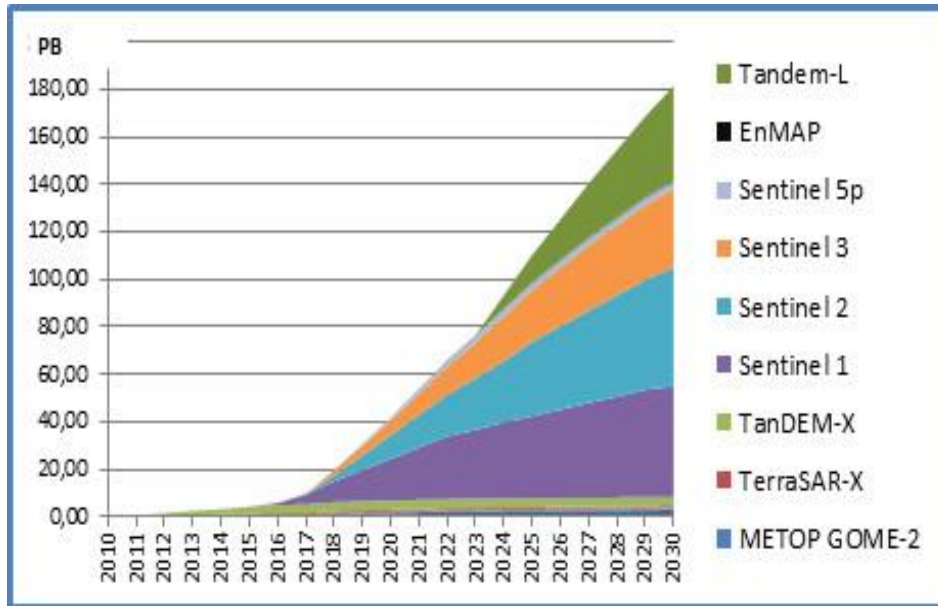
C.O. Dumitru, G. Schwarz, D. Ao, G. Dax, V. Andrei, C. Karmakar, and M. Datcu

Knowledge for Tomorrow



Big Data & Sensors

The volume of data for different sensors over the years



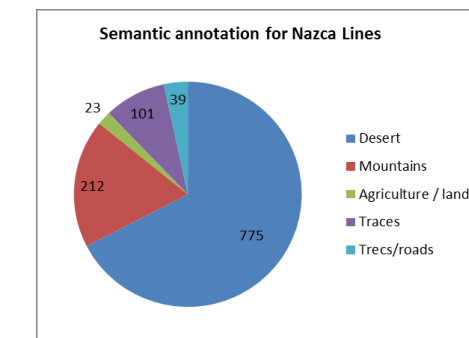
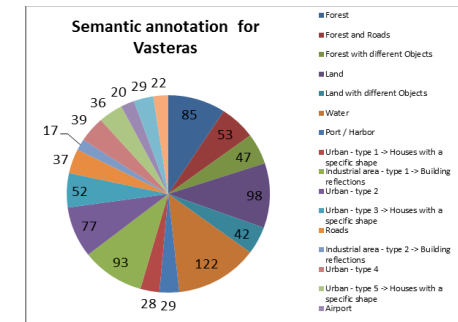
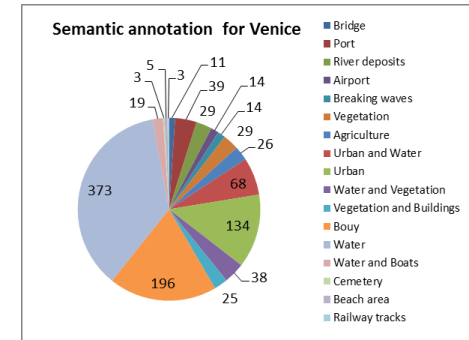
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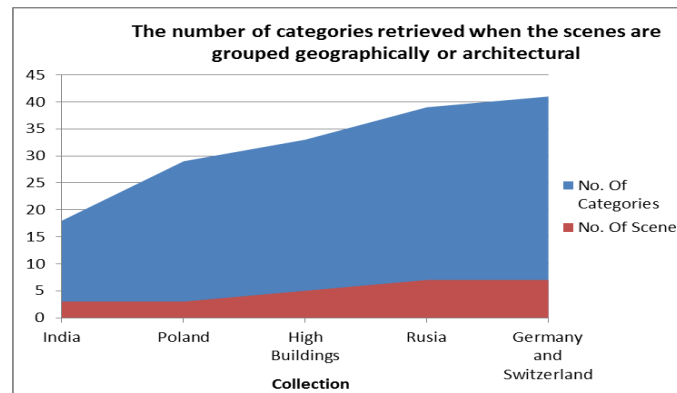


Applications: TerraSAR-X archive semantic catalogue

300 cities 1300 semantic labels



- Bangkok (Thailand);
- Shenyang (China);
- Nazca Lines (Peru);
- Havana (Cuba);
- Venice (Italy);
- Vasteras (Sweden);
- Oran (Algeria);
- Bogota (Columbia).



Applications: Protected areas in Europe - ECO-POTENTIAL

General description of Ecopotential protected areas

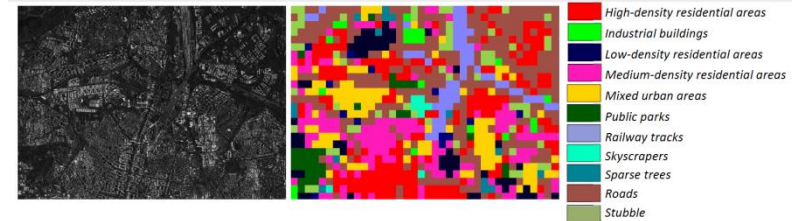


Fig. 21. b. TerraSAR-X quick-look view (left) and classification map (right) for an image of Madrid, Spain.

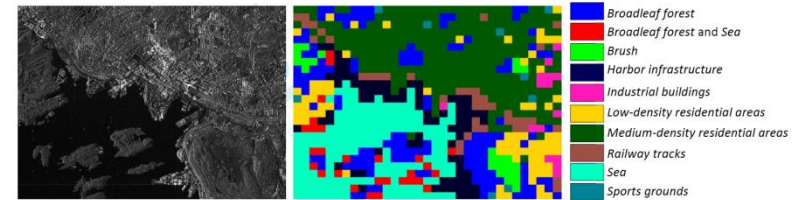
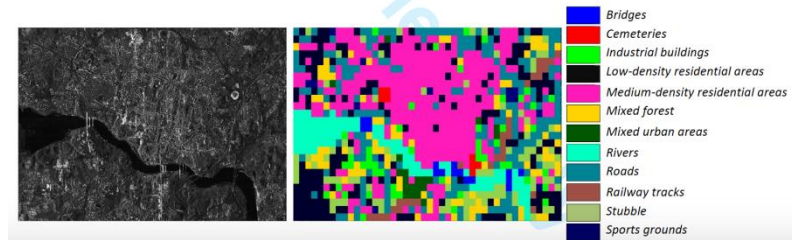
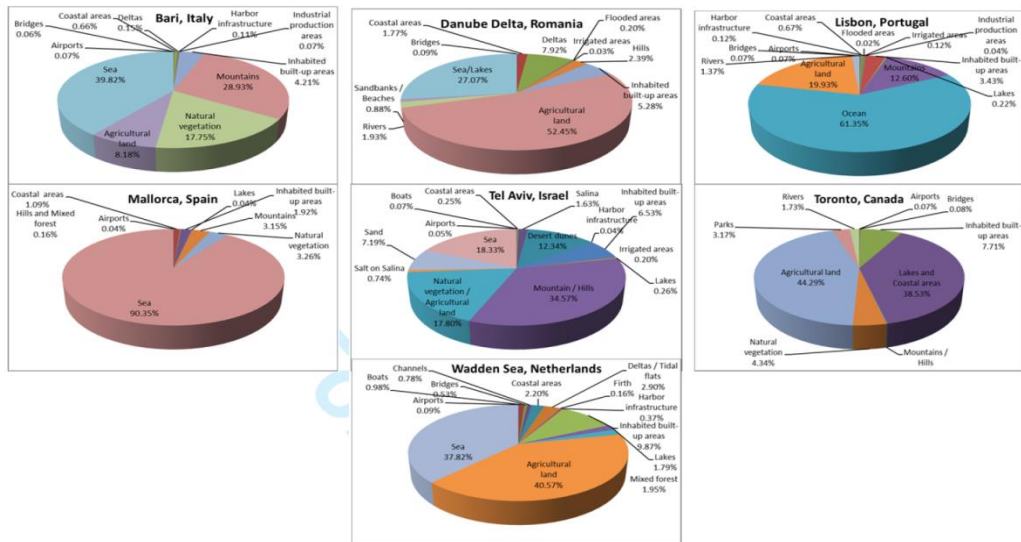


Fig. 21. c. TerraSAR-X quick-look view (left) and classification map (right) for an image of Oslo, Norway.



Data Mining Tools

Multimodal queries

The screenshot shows the Dolphin software interface. On the left, a 'Semantic' tree lists various categories, with 'Medium density residential areas' selected. Below the tree, a 'Query Expression' table is visible:

Parameter	Operator	Value	Connector
name	=	Storage.t	AND
name	=	Medium	AND

The main window displays a table of query results with columns: stion, label_id, label_name, goodness, coverage, trust, lastupdate, label_id, parentlabel, name, description, level, source, and stp. Below the table is a grid of 24 small satellite images corresponding to the results.

Query by example

The screenshot shows the Dolphin software interface for a 'Query by example'. It features a large satellite image on the left and a grid of smaller images on the right. The interface includes a 'Find Images' section, a 'Find by Name' section, and a 'Find by Example' section. The 'Find by Example' section has a 'System' dropdown set to 'Vector Distance' and a 'Labels' section with 'L1' selected. The 'Find by Example' section also has a 'Find by Example' button and a 'Labels' section with 'L1' selected.

KDD and semantic extraction

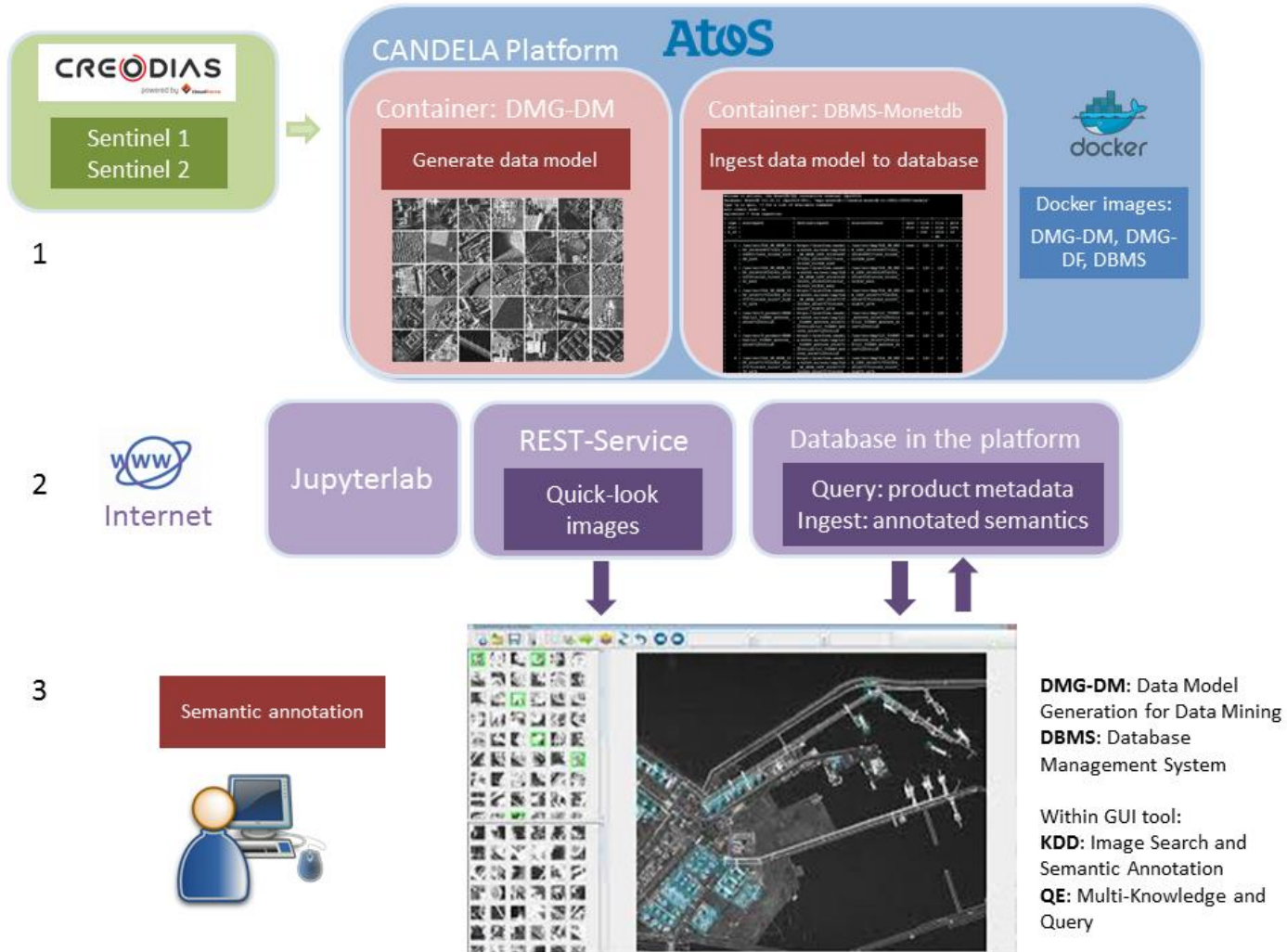
The screenshot shows the Dolphin software interface for 'KDD and semantic extraction'. It features a grid of 8 satellite images. Below the grid are buttons for 'Int', 'Run', 'Save as...', 'Classify...', and 'Show classification'. The 'Results' section at the bottom shows a grid of 8 small images, likely representing the extracted features or classifications.

Spatio temporal patterns discovery

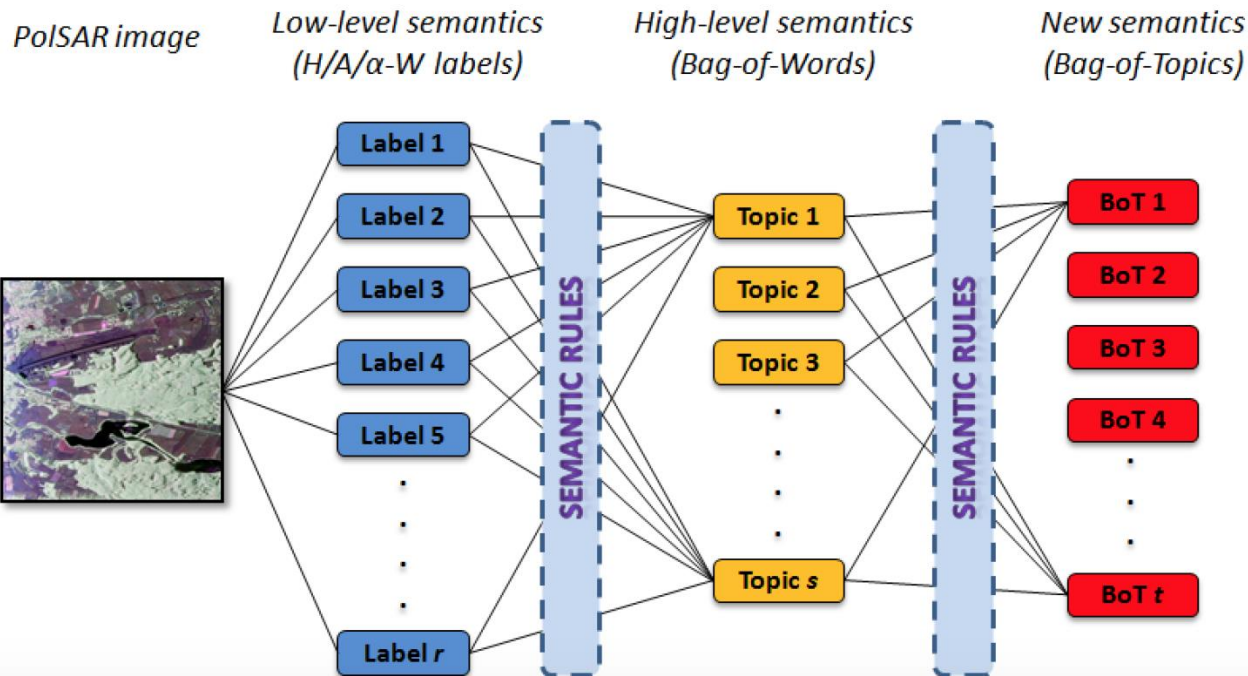
The screenshot shows the Dolphin software interface for 'Spatio temporal patterns discovery'. It features a line graph at the top with 'DjvTresh-' at 0.72 and 'DjvTresh+' at 0.51. The graph shows a fluctuating line with a peak around 20001025. Below the graph is a grid of 8 satellite images, with the bottom row labeled 'Image-' and 'Image+'. The bottom right corner of the grid shows the date '00/10/15'.



CANDELA



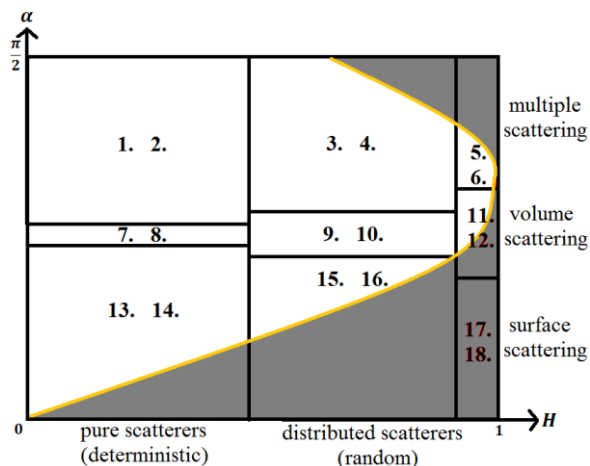
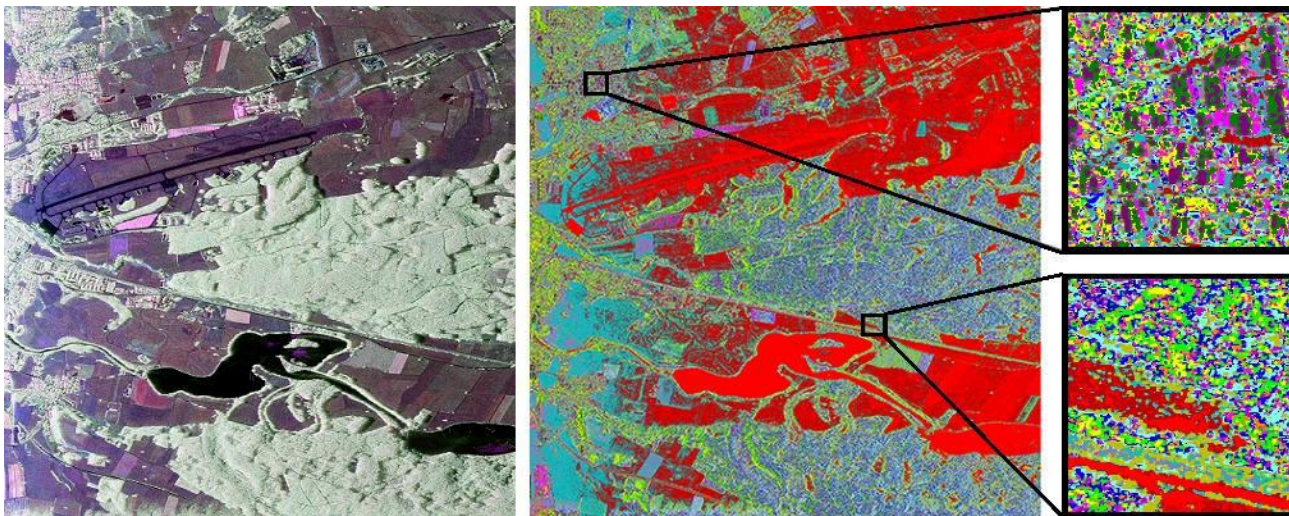
Discovery of semantic relationships: PolSAR L-Band



28\36



PolSAR L-Band Low-level semantics – H\A\alpha classes



Label	Semantics
L1	Low-entropy multiple scattering with low anisotropy
L2	Low-entropy multiple scattering with high anisotropy
L3	Medium-entropy multiple scattering with low anisotropy
L4	Medium-entropy multiple scattering with high anisotropy
L5	High-entropy multiple scattering with low anisotropy
L6	High-entropy multiple scattering with high anisotropy
L7	Low-entropy volume scattering with low anisotropy
L8	Low-entropy volume scattering with high anisotropy
L9	Medium-entropy volume scattering with low anisotropy
L10	Medium-entropy volume scattering with high anisotropy
L11	High-entropy volume scattering with low anisotropy
L12	High-entropy volume scattering with high anisotropy
L13	Low-entropy surface scattering with low anisotropy
L14	Low-entropy surface scattering with high anisotropy
L15	Medium-entropy surface scattering with low anisotropy
L16	Medium-entropy surface scattering with high anisotropy

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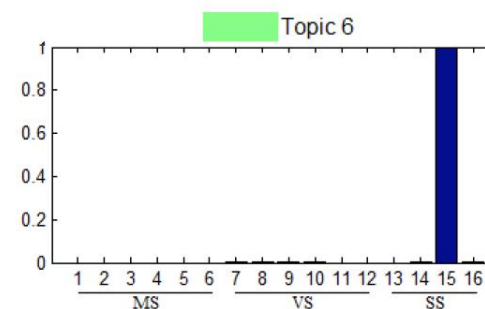
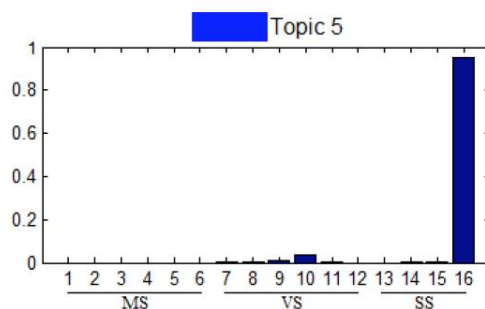
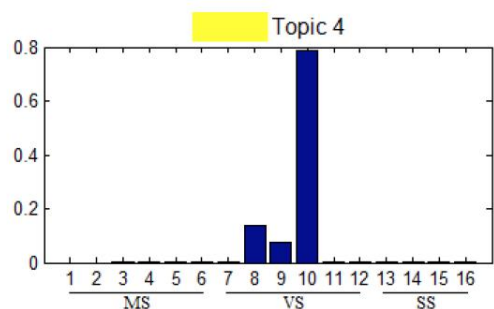
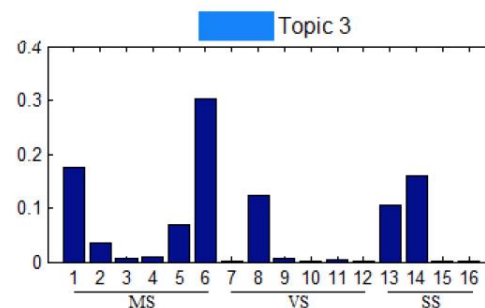
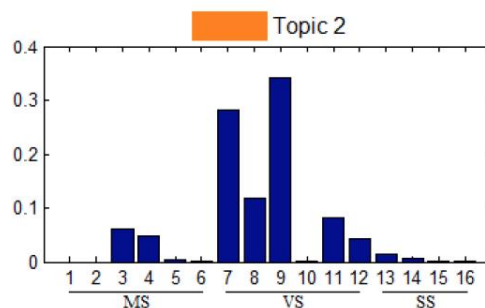
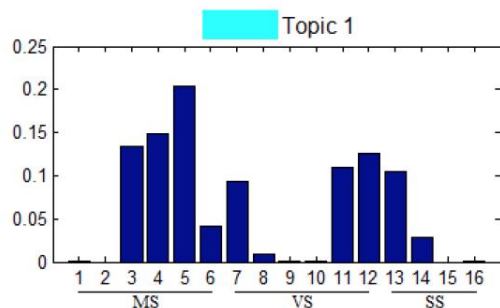
[Discovery of Semantic Relationships in PolSAR Images Using Latent Dirichlet Allocation](#)

IEEE Geoscience and Remote Sensing Letters 2017

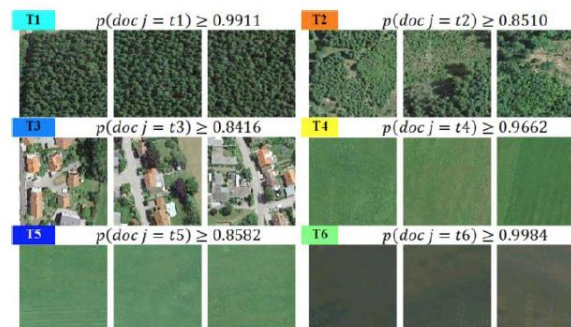


High-level semantics – LDA topics for PolSAR L-Band

$$p(w_{dn} | a, B) = \int p(\theta_d | a) \left[\sum_{z_{dn}} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn}, B) \right] d\theta_d$$



Topic	Semantics	Semantic rule
T1	Woody vegetation	53% MS + 34% VS + 13% SS
T2	Mixed woody vegetation & shrubs	12% MS + 86% VS + 02% SS
T3	Artificial, man-made structures	60% MS + 14% VS + 26% SS
T4	Herbaceous vegetation	00% MS + 100% VS + 00% SS
T5	Smooth surface	00% MS + 05% VS + 95% SS
T6	Specular surface	00% MS + 00% VS + 100% SS



Trends

- Enforce the market/business of AI and Data Analytics in EO
- Grow multi-disciplinary and disruptive thinking HR
- Apply AI where it is needed and works such to obtain more than 20% *better*
- Target to implement solutions for the big, required, but unsolved yet technologies for global and long term EO market:
 - Joint multi-mission and multi-source data valorization
 - Satellite Image Time Series valorization
 - Multisource model assimilation and physical parameters inversion
 - Predictive models and forecast Earth dynamic
 - Quantum technologies
- Elaborate an overall EO intelligence for the system of systems:
- DATA – SENSOR – MISSION – ARCHITCTURE



Threats

- **DNN: in 2019 more than 500 papers/month**
- **Research is often wasted effort**
- **ML faces a deep reproducibility crisis**
- **Training data is as important as the learning algorithm**
- **ML finds any pattern in data, it may be irrelevant**
- **We need the actual patterns of the Earth processes**
- **Big EO Data accentuate the crisis**



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