

Department of Geoinformatics and Remote Sensing Faculty of Geography and Regional Studies Warsaw University, Poland

Vegetation mapping of the High Tatras and the Karkonosze Mts. using hyperspectral images and neural networks

2013-07-23

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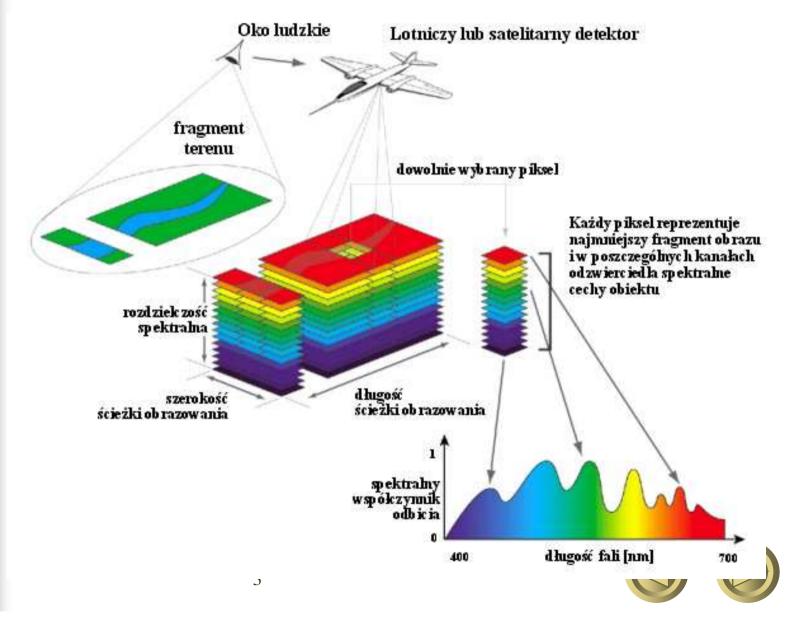
### The study aims at:

- Mountain ecosystem mapping
- Plant communities identification,
- Image classification using fuzzy ARTMAP and SVM methods,
- Analysis of airborne hyperspectral data potential for vegetation monitoring
- Presenting the HySens (the Tatras) and HyMountEcos (Krkonose/Karkonosze) projects.



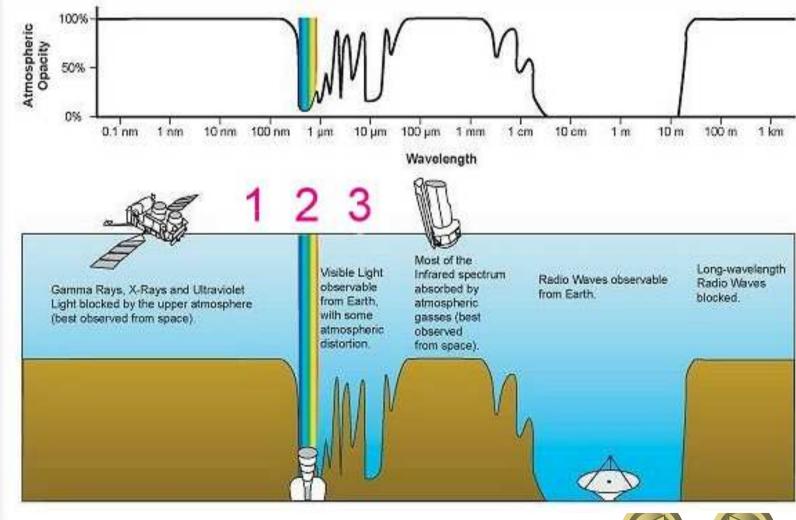


## Idea of remote sensing





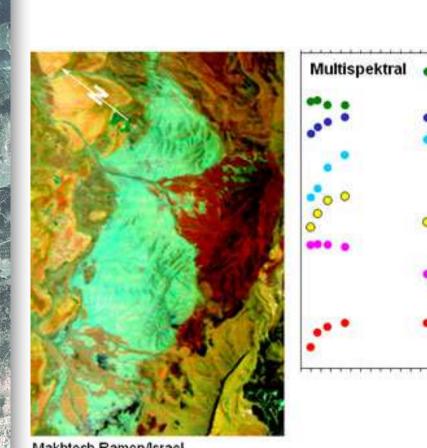
### Spectral ranges of remote sensing of environment



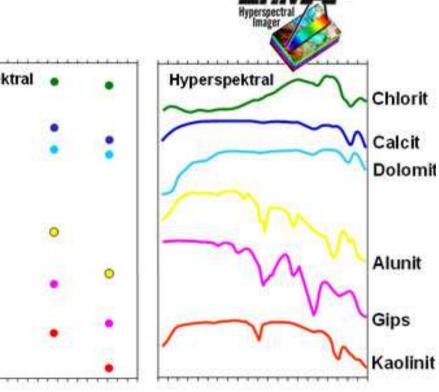




### Multi-, hyperspectral remote sensing



Makhtesh Ramon/Israel Farbdarstellung der Bänder 1, 20, 48

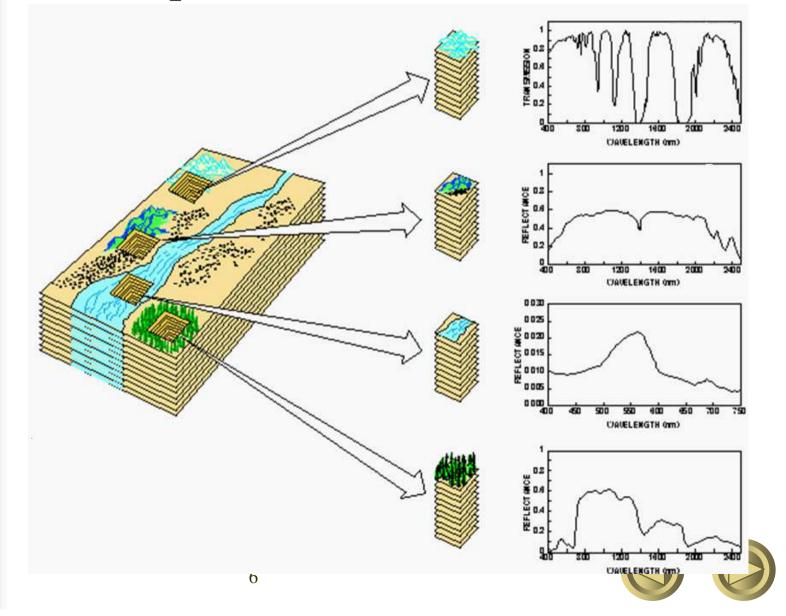




MAP



### Data acquisition





### **High Tatras** (Tatra National Park, M&B reserve)







2013-07-23





### **Research objects** (1) (42 classes)

- cryptogamic plant communities on scree initial phase
- epilitic lichen communities
- scree communities
- snow-bed communities
- subnivale swards
- alpine swards
- peaty and boggy communities
- avalanche meadows
- tall herb communities
- grassland communities after grazing
- subalpine dwarf scrub communities
- willow thicket
- mountain-pine scrub on silikat substrate
- mountain-pine scrub on calcareus substrate
- montane spruce forest
- lakes











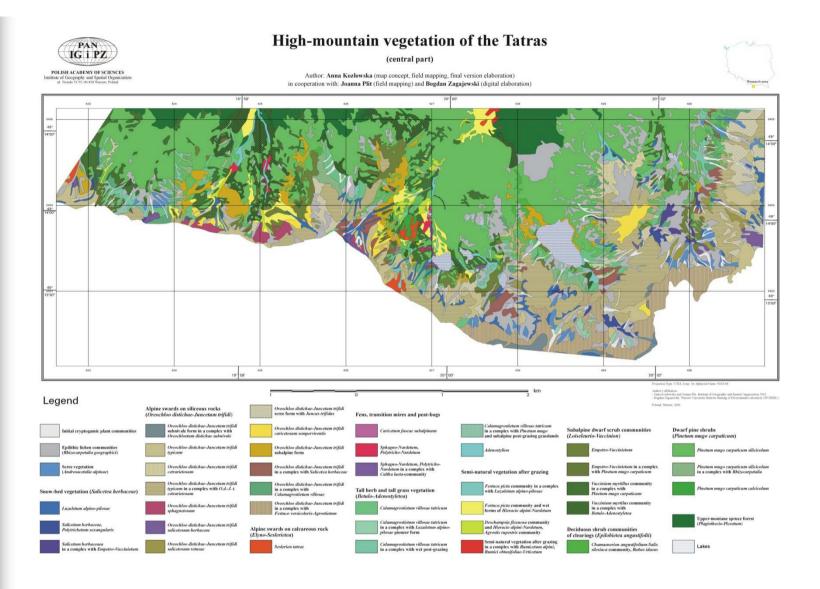














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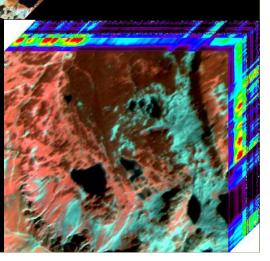


### **DAIS 7915 images**

4 flight lines 79 bands, 15 bit, 3 m pixel size,

*DAIS* 7915 *RGB* 22, 12, 1 compositions of flight lines 2., 4., 5. and 6.

Data cube of the Gasienicowa Valley (central part of the test area )





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### **Research algorithm**

#### Building of reference layers

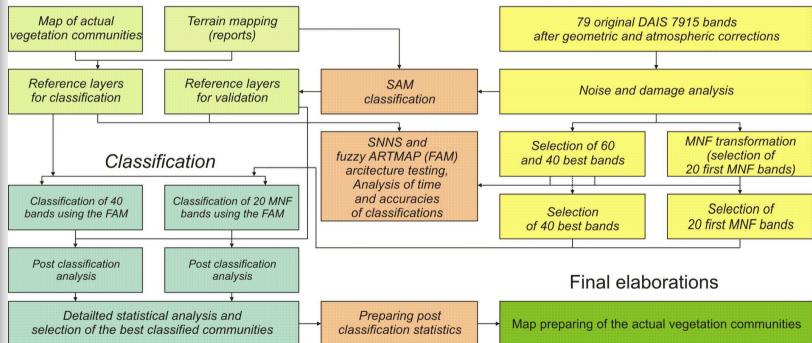
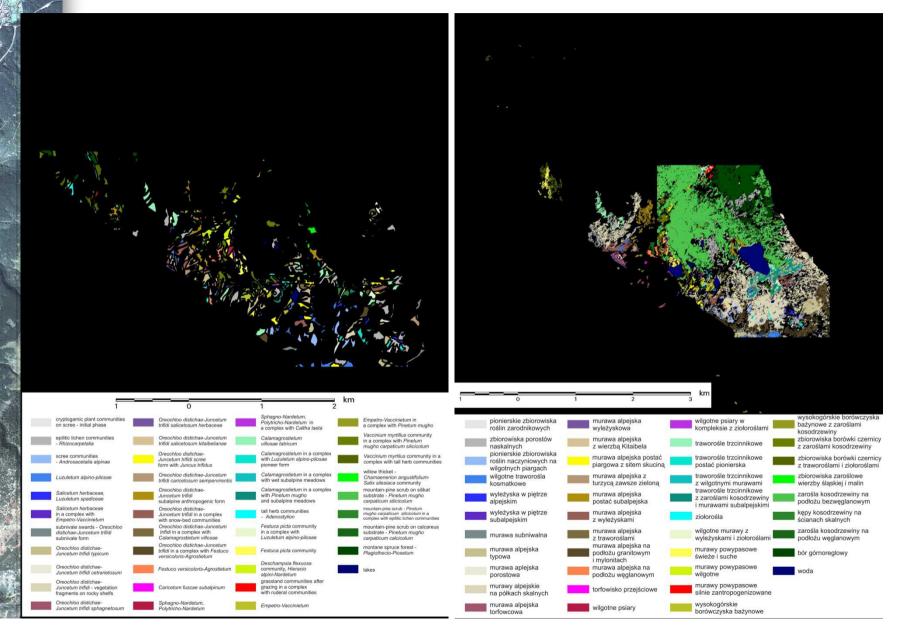


Image processing

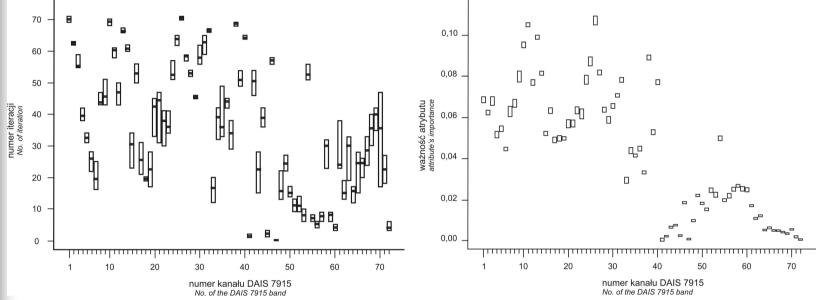
### **Teaching and validation sets**

### (on the base of field mapping and SAM classification)





### Image data quality analysis



Quality of DAIS 7915 bands: 26, 1, 10, 38, 13, 32, 40, 25, 31, 2, 14, 11, 30, 27, 3, 46, 24, 16, 28, 54, 39, 42, 12, 9, 29, 8, 36, 69, 20, 35, 21, 4, 44, 22, 34, 23, 68, 70, 37, 5, 67, 15, 61, 58, 63, 17, 6, 49, 66, 71, 19, 43, 65, 18, 7, 48, 62, 33, 64, 50, 52, 51, 53, 57, 59, 55, 56, 72, 60, 45

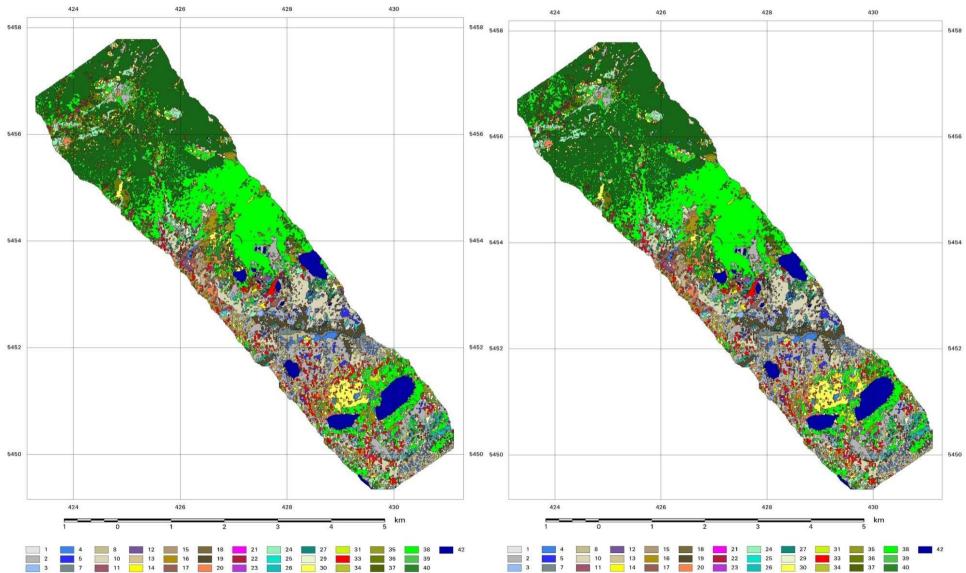
2013-07-23



## **Results** (line #5, 5000 iterations)

40 bands

20 MNF bands



	Band #6				Band #5				Band #4		Band #2				
	40 bands      20 MNE        87.40%      86.4        (362757/415058)      (358569/		F bands 40 bands		20 MNF bands		40 bands		40 bands		20 MNF bands				
Overall					86.96%		85.4			07%	88.79%			45%	
Accuracy			(358569/414637)		(286175/329088)		(281235/329065)		(224941/252538)		(352691/397228)		(335445/397228)		
Карра								,							
Coefficient	0.8429		0.8.	0.8310		0.8425		0.8248		0.8581		0.8634		0.8103	
Class	Prod. Acc.	User Acc.	Prod. Acc.	User Acc.	Prod. Acc.	User Acc.	Prod. Acc.	User Acc.	Prod. Acc.	User Acc.	Prod. Acc.	User Acc.	Prod. Acc.	User Acc.	
Class #1	87.8	75.1	88.3	73.9	92.4	80.1	92.4	80.1	-	-	88.9	69.4	83.3	65.37	
Class #2	78.5	74.3	78.9	72.9	83.6	82.5	79.0	81.6	84.1	68.7	79.6	82.0	75.13	76.39	
Class #3	96.0	54.9	90.7	53.5	92.8	73.2	92.8	71.7	-	-	96.5	66.8	93.73	61.7	
Class #4	97.4	68.0	97.5	67.7	93.5	79.1	93.5	78.8	95.8	70.7	93.8	76.5	90.31	45.14	
Class #5	95.6	72.1	95.6	72.1	95.2	57.3	95.2	53.4	-	-	89.9	74.5	75.7	70.02	
Class #7	90.7	65.9	90.7	64.1	95.8	53.2	94.6	52.9	-	-	90.1	75.5	87.06	64.67	
Class #8	89.4	74.8	89.5	74.2	89.1	79.1	89.2	76.9	94.0	93.0	92.2	81.0	89.43	73.07	
Class #9	-	-	-	-	-	-	-	-	98.3	66.6	-	-	-	-	
Class #10	73.3	92.6	73.4	91.7	80.7	94.9	77.9	94.6	86.5	89.3	78.7	95.2	73.45	93.93	
Class #11	-	-	-	-	91.4	98.1	91.4	97.1	-	-	-	-	-	-	
Class #12	-	-	-	-	90.5	89.3	90.5	89.3	-	-	-	-	-	-	
Class #13	91.7	73.0	80.5	69.0	94.9	45.2	94.9	45.2	-	-	89.9	81.8	73.33	76.02	
Class #14	82.2	64.0	54.4	52.9	78.6	80.5	78.6	80.5	94.4	32.3	83.8	77.1	75.34	59.61	
Class #15	87.9	83.2	70.7	82.1	79.4	68.4	79.4	68.2	97.9	63.6	87.1	81.6	59.1	72.76	
Class #16	76.2	94.4	65.0	94.7	78.5	83.4	78.5	81.9	90.6	77.6	83.9	91.6	80.15	88.77	
Class #17	81.4	82.5	68.9	80.1	68.8	91.4	68.9	88.3	95.0	90.6	91.3	72.2	87.02	65.3	
Class #18	85.9	73.6	86.0	71.5	71.5	93.2	71.5	93.2	95.4	84.0	81.7	72.5	76.16	67.77	
Class #19	84.8	76.8	84.8	76.2	72.8	86.3	72.8	85.4	90.1	88.4	90.4	82.6	85.69	77.46	
Class #20	95.1	85.2	61.8	86.3	89.8	66.6	83.2	63.8	-	-	91.9	89.1	83.7	87.22	
Class #21	-	-	-	-	95.7	84.6	95.7	84.6	-	-	97.8	83.5	88.89	85.71	
Class #22	85.6	80.7	74.0	75.2	90.8	56.4	87.8	55.7	-	-	91.3	84.0	87.78	80.63	
Class #23	93.8	78.2	83.5	67.4	94.1	55.1	94.1	54.5	-	-	84.9	85.3	73.14	85.09	
Class #24	82.1	73.4	78.2	71.9	73.7	84.6	73.8	81.6	91.0	70.0	80.3	71.2	66.55	60.86	
Class #25	95.8	67.9	91.6	65.8	86.0	63.0	86.0	58.1	-	-	93.6	70.0	89.45	61.91	
Class #26	89.7	67.5	73.8	63.8	97.8	30.8	97.8	27.5	-	-	91.1	81.4	77.13	80.44	
Class #27	-	-	-	-	83.7	87.5	83.8	87.1	-	-	76.5	84.5	38.02	72.89	
Class #28	83.9	79.5	64.4	76.5	-	-	-	-	-	-	87.2	72.4	88.27	71.57	
Class #29	97.8	73.9	91.8	73.4	84.0	67.5	84.3	65.1	92.7	89.8	76.9	78.5	76.56	67.43	
Class #30	80.8	57.4	74.8	48.5	76.7	80.6	76.7	79.1	-	-	87.9	60.0	82.75	53.85	
Class #31	85.5	71.0	69.7	66.5	81.7	76.4	81.7	76.4	80.1	59.1	89.4	59.6	79.8	47.8	
Class #33	96.1	75.9	96.6	73.9	96.5	64.3	95.7	59.5	98.8	63.3	93.5	66.7	94.21	60.23	
Class #34	-	-	-	-	85.6	96.2	85.6	96.2	-	-	-	-	-	-	
Class #35	84.8	72.0	85.7	67.6	81.4	82.7	81.5	81.7	98.1	50.5	86.8	69.5	78.5	63.75	
Class #36	75.3	73.7	63.6	73.3	73.8	84.0	66.0	78.5	68.4	76.4	71.1	78.5	62.06	61.84	
Class #37	92.6	66.0	63.3	60.4	94.3	28.3	95.1	18.7	74.7	51.1	92.4	49.6	65.56	38.72	
Class #38	95.8	93.9	96.0	93.4	93.4	96.2	92.3	95.3	91.8	92.1	94.6	96.8	94.09	95.04	
Class #39	85.2	72.6	85.4	71.9	91.1	67.2	91.1	67.2	-	-	93.8	67.7	89.97	60.01	
Class #40	-	-	-	-	95.9	37.4	95.9	37.2	-	-	96.0	86.0	92.02	76.38	
Class #41	81.3	97.0	81.4	97.0	-	-	-	-	80.9	96.0	87.7	96.0	77.16	96.16	
Class #42	97.4	98.1	97.5	98.0	97.8	97.8	97.8	97.8	98.7	99.6	96.7	98.3	96.78	97.85	



### Conclusions

- Hyperspectral remote sensing techniques and ANN show potential for mountainous vegetation mapping in large scales, what can bu used for monitoring of high-mountain ecosystems,
- 39 of 41 vegetation classes were classified,
- The best results (producer and user accuracies > 90%) are observed for: #8 Oreochloo distichae-Juncetum trifidi typicum; #11 Oreochloo distichae-Juncetum trifidi sphagnetosum; #16 Oreochloo distichae-Juncetum trifidi subalpine anthropogenic form; #17 Oreochloo distichae-Juncetum trifidi in a complex with snow-bed communities; #18 Oreochloo distichae-Juncetum trifidi in a complex with Calamagrostietum villosae; #38 mountain-pine scrub on silicate substrate - Pinetum mugho carpaticum silicicolum; #42 lakes
- The best average fourty-band set's results: 92.8% (producer accuracy) and 84,2% (user acc.). The worst: 84.2% and 67.5%.
- The best average twenty-MNF-band set's results: 86.1% and 79.9%, and the worst 74.7% and 63.7%
- The best results are observed for: oat crops (99,8 %), stubbles (96,6 %), grasslands (94,8 %), deciduous forest (93,9 %), and the worst for tree clumps (38,8 %), orchards (44,7 %) and side roads (56,1 %),
- Twenty-MNF-band set achieves ~7-9% worst results, but classifications were 2-3 times faster





Hyperspectral remote sensing for Karkonosze ecosystems – EUFAR's HyMountEcos Project

Zagajewski B., Kupkova L., Markowicz K.M., Kozłowska A., Adamczyk J., Albrechtova J., Będkowski K., Bilip M., Chiliñski M., Jarocińska A., Kycko M., Lhotakova Z., Marcinkowska A., Mierczyk M., Nasiłowska S., Ochtyra A., Knapik R., Oprządek M., Pabjanek P., Potuckova M., Przewoźnik L., Raczko E., Sendyk A., Slacikova J., Stachlewska I.S., Tobiasz M., Wojtuń B., Zawadzka O., Żołnierz L.



# HyMountEcos project Project acronym HyMountEcos Project title Hyperspectral Remote Sensing for Mountain Ecosystems

- Type Scientific project
- Scientific theme Assessment of the hyperspectral techniques potential for mountain ecosystems monitoring
- Main scientific field and Specific discipline Earth
  Sciences & Environment / Ecosystems & Biodiversity



# Scientific problems

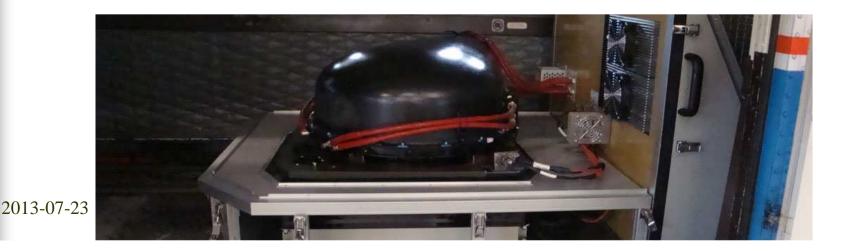
- monitoring of mountain ecosystems of the Giant Mountains (Karkonosze/Krkonoše) National Park
- to assess influence of management practice.
- mountain ecosystems mapping and inventarization,
- an analyses and evaluation of forest ecosystems conditions/health,
- an analyses of ecosystems species composition and invasive species introduction
- an analysis of soil contamination.





## **APEX Performance**

- Spectral Range 380.5 971.7 nm VNIR 941.2 2501.5 nm SWIR
- Spectral Bands Up to 532 (Default: 312)
- SSI 0.45 7.5 nm 5 10 nm
- FWHM 0.7 9.7 nm 6.2 12 nm
- Spatial 1000 Pixel
- Ground Resolution 0.5 1.75m @ 1000 3500m AGL
- FOV 28 Degrees



## Data acquisition (10.09.2012)





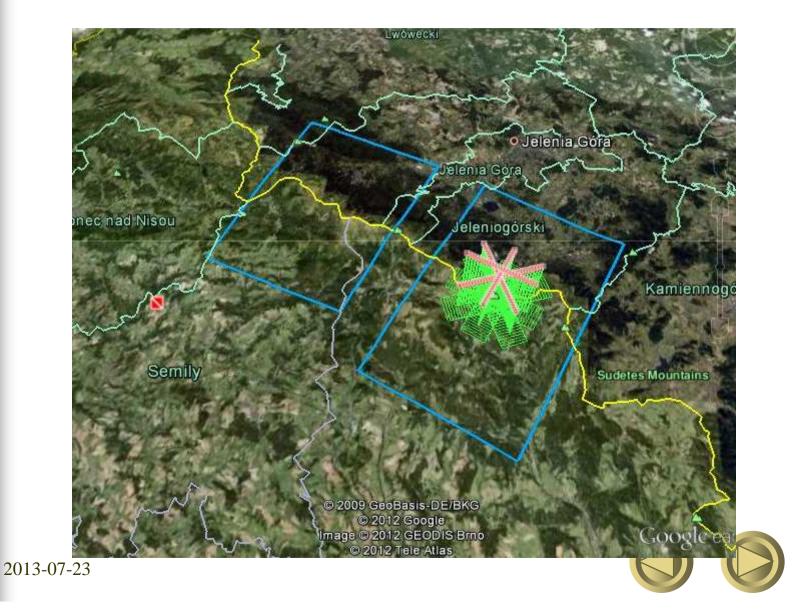






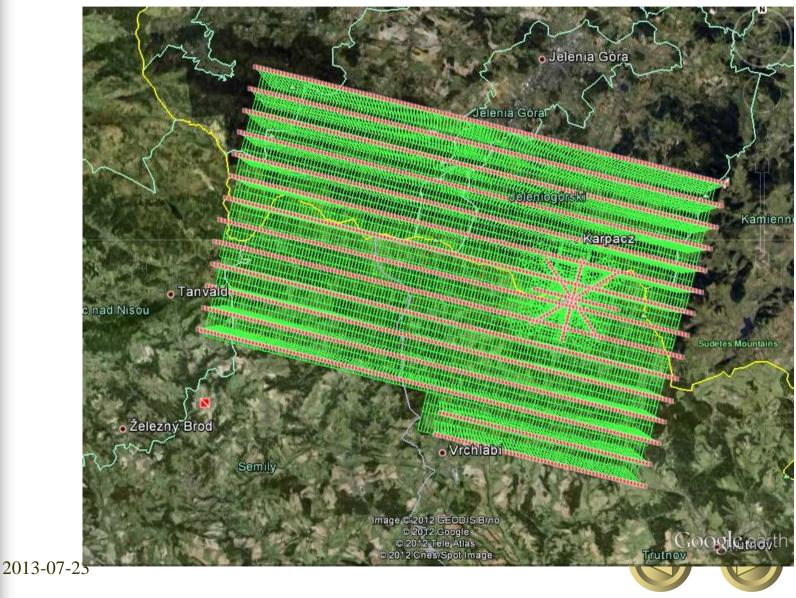


## Flight plan (BRDF)



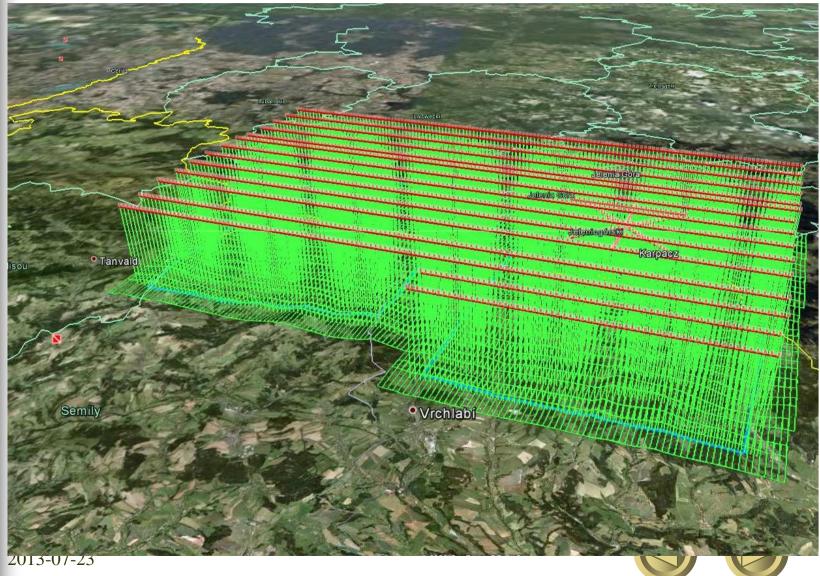


## Flight plan 2





### Flight plan 3



# Output of the project

- Mountain ecosystems mapping and inventarization,
- Analyses of ecosystems species composition and invasive species introduction,
- Analyses and evaluation of forest ecosystems conditions/health (biophysical parameters like chlorophyll content, LAI, water content).
- Processing chain for mountain ecosytems monitoring using hyperspectral technologies and potential/feasibility assessment of hyperspectral data/technologies for the mountain ecosystems analysis and monitoring.



## Vegetation mapping

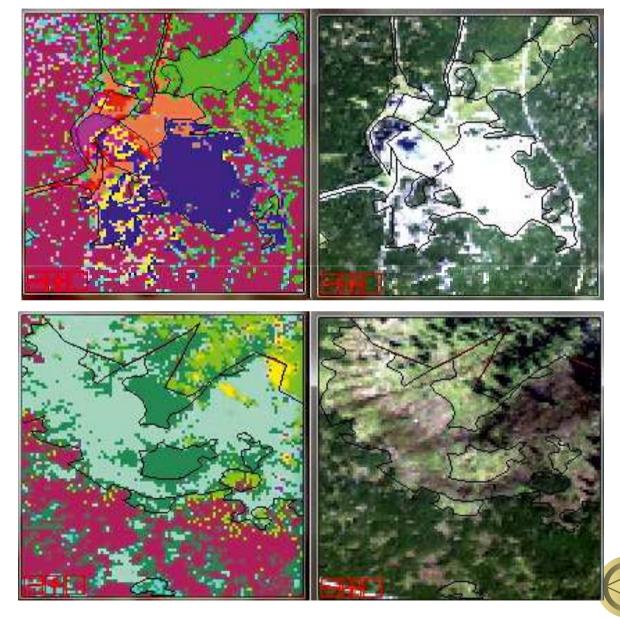


 Vaccinium myrtillus, 2. Associations from Rhizocarpion alpicolae alliance, 3. Associations from Umbilicarion cylindricae alliance, 4. Carici (rigidae)-Nardetum, 5. Associations from Artemisietea vulgaris class, 6. Pinetum mugo sudeticum, 7. Associations from Calamagrostion alliance, 8. Associations from Scheuchzerio-Caricetea nigrae class, 9. Deschampsia caespitosa, 10. Athyrietum distentifolii, 11. Luzulo nemorosae-Fagetum (typical), 12. Abieti-Piceetum (montanum), 13. Calamagrostio villosae-Piceetum (division of fern), 14. Calamagrostio villosae-Piceetum typicum, 15. Area without vegetation.





## Validation process







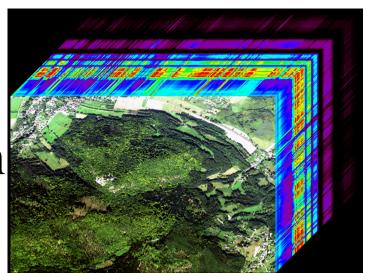
## Accuracy analysis

Class	Prod. Acc. (%)	User Acc. (%)	Class	Prod. Acc. (%)	User Acc. (%)
Class	77,24	73,85	9	73,23	76,86
2	99,06	99,76	10	0,24	3,23
3	93,33	54,55	11	77,02	77,32
4	89,37	96,35	12	65,63	69,44
5	89,13	94,25	13	99,14	92,74
6	99,14	94,86	14	81,59	45,56
7	93,17	96,22	15	83,76	90,48
8	83,72	97,3			

Overall Accuracy = 79,1259% Kappa Coefficient = 0,7736



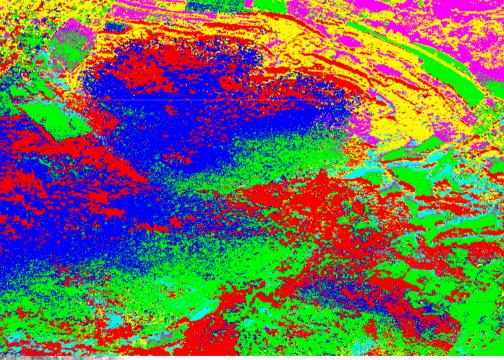
## Results of tree species classification

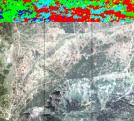


	Class	Pine	Beech	Larch	Adler	Birch	Spruce	
	Pine	91,71	0,00	0,00	0,00	0,00	0,85	
	Beech	0,00	88,04	9,87	14,20	10,39	0,47	
	Larch	0,00 1,15		58,48	4,32	16,23	0,57	
188 . A.	Adler	0,00	3,17	3,29	81,48	1,08	0,19	
A STATE	Birch 0,00		7,06	13,16	0,00	59,31	1,32	
	Spruce	8,29	0,58	15,19	0,00	12,99	96,60	

Overall Accuracy: 82,64%

Kappa Coefficient: 0,77







## Conclusions

- Dominant (spruce, beech and larch) tree species are usually better classified than those that are sparser.
- Best results were obtained for spruce (above 90% on both areas) and pine (above 80%).
- Worse accuracy for larch was mainly caused by small sample of data from which learning and verification polygons were taken.
- Results for deciduous trees were bit less accurate ranging from 59% for birch to 88% for beech. Given classifications were made on diverse test sites with mixed forest composition
- The best results are observed for associations from Rhizocarpion alpicolae alliance 99,06 % (prod. a), 99,76% (user a), the worst Athyrietum distentifolii 0,24% (p.a.), 3,23 % (u.a.). It was caused e.g. by homogeneity of polygons representing each class.





# Thank you very much for your attention



Special thanks to:

- Lucie Kupkova and colleagues from the Charles University
- Andreas Mueller and the DLR team for the DAIS data
- Koen Mueleman and the VITO and DLR teams for the APEX data
- Paolo Gamba for the fuzzy ARTMAP simulator
- Management Boards of the Tatra, Krkonose and Karkonosze National Parks



2013-07-23