## Classification of tundra vegetation in the Krkonoše Mts. National Park using APEX, AISA Dual and Sentinel-2A Data

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SCERIN-5 Capacity building workshop, Pécs Hungary, June 2017

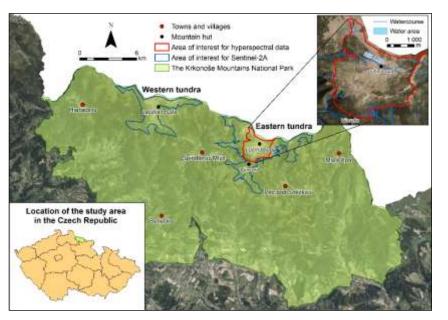
# Starting points

- Tundra ecosystems (alpine treeless) belong to the most valuable natural phenomena worldwide.
- Biotopes above the treeline are very sensitive to various types of environmental factors
- Changes can be very fast in these areas and their monitoring is very important
- Earth observation potentially powerfull tool for the monitoring

# Goals

- To evaluate and compare suitability of aerial hyperspectral data (AISA Dual and APEX sensors) with freely available Sentinel-2A data for classification of tundra vegetation cover in the Krkonoše Mts. National Park.
- Different classification methods (pixel and object-based) were used to find out which classification algorithm for which type of data can bring the most accurate classification results.
- We expected that the best accuracy will be achieved using hyperspectral data with higher spatial and spectral resolution (AISA Dual).
- Further assumption was that in the case of Sentinel-2A data with its limited spatial and spectral resolutions some vegetation (especially grassland) categories will not be distinguishable.

## Study area



Over the years affected by human impacts

The highest parts of the Krkonoše Mts. National above the treeline (1,300 m a. s. l.)

A unique ecosystem, southernmost relict area of the arctic-alpine tundra in Europe

Area of 47 km<sup>2</sup> - 7.4% of the total Krkonoše Mts. Area (Czech and Polish sides). Two parts: Western and Eastern.

As a result of palaeogeographical history the Krkonoše Mountains represent a "biodiversity crossroads" where Nordic and alpine flora and fauna coexist

Besides the mosses, lichens, and alpine heathlands, the prevailing vegetation types are: closed alpine grasslands dominated by *Nardus stricta*, subalpine tall grasslands, and *Pinus mugo* scrub

From the 9<sup>th</sup> century till the beginning of the 19<sup>th</sup> century expanding due to local agricultural practices that included deforestation and grazing

Since early 20<sup>th</sup> century this human impact has been reduced and the area became strictly protected as a nature reserve.



# Data and classification legend

	Number of used		Spatial ground	
Sensor	bands	Wavelength range	resolution	Acquisition date
APEX	288	400 nm - 2,500 nm	2 to 5 m	09/10/2012
AISA Dual	494	400 nm - 2,500 nm	1 to 3 m	06/19/2013
Sentinel-2A	10	400 nm - 2,300 nm	10 and 20 m	08/30/2015

#### Detailed legend

- 1. Block fields and anthropogenic areas
- 2. Pinus mugo scrub (Mountain pine)
- 3. Subalpine *Vaccinium* vegetation (Blueberries, cranberries and bog bilberries)
- 4. Closed alpine grasslands\*
- 4a. Nardus stricta stands (Matgrass)
- 4b. Species-rich vegetation with high cover of forbs
- 5. Subalpine tall grasslands\*
- 5a. Calamagrostis villosa stands (Hairy reed grass)
- 5b. Molinia caeruela stands (Purple moor grass)
- 5c. Deschampsia cespitosa stands (Tufted hair grass)
- 6. Alpine heathlands
- 7. Wetlands and peat bogs
- 8. Water areas (not for Sentinel-2A)

#### Simplified legend

- 1. Block fields and anthropogenic areas
- 2. Picea abies stands (Norway spruce)

3a. *Pinus mugo* scrub dense (more than 80 % of total cover)

3b. *Pinus mugo* scrub sparse (30 – 80 % of total cover)

- 4. Closed alpine grasslands dominated by *Nardus stricta*
- 5. Grasses (except Nardus stricta) and
- subalpine Vaccinium vegetation
- 6. Alpine heathlands
- 7. Wetlands and peat bogs

### **Classification legend**

Block fields and anthropogenic areas Pinus mugo scrub Subalpine Vaccinium vegetation Species-rich vegetation with high cover Nardus stricta stands of forbs Calamagrostis villosa stands Molinia caeruela stands

Alpine heathlands

Wetlands and peat bogs

Water areas



Deschampsia cespitosa stands





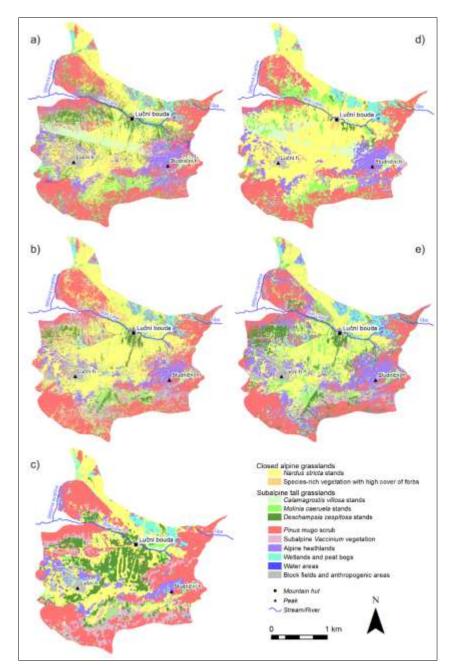
# Workflow

Legend type		Simplified legend					
Area of interest		Eastern Tundra					
Training data		51 polygons (11,388 m <sup>2</sup> ) collected in the field (2014 and 2015) and from orthoimages, specifically edited for each image data*					
Image data	APEX	APEX AISA Dual Sentinel-2A					
Number of bands	PCA 5, PCA 40, 288	PCA 7, PCA 40, 494	10	10			
Pixel-based classification	SVM NN	SVM NN MLC	SVM NN MLC	SVM NN MLC			
Object-based classification	SVM	SVM	]				
Validation data	72 polygons (17,129 from orthoimage	Re-classified field data from 2014 adapted for Sentinel pixel**					

\* Edited based on the pixel-size; for Sentinel-2A class "water areas" was not assessed.

\*\* Dataset originally created for Landsat 8 classification in Suchá et al. (2016) edited for Sentinel-2A pixel size.

#### Results for detailed legend (Eastern tundra, all data)



Land cover in the Eastern Tundra of the Krkonoše Mts. for the best classification results of per-pixel and object-based approaches:

a) per-pixel classification: APEX data, SVM classifier, 40 PCA bands;

*b) per-pixel classification: AISA Dual data, SVM classifier, 40 PCA bands* 

c) per-pixel classification: Sentinel-2A data, NN classifier (user defined);

d) object-based classification: APEX data, SVM classifier, 40 PCA bands;

e) object-based classification: AISA Dual data, SVM classifier, 7 PCA bands.

#### Results for detailed legend (Eastern tundra, all data) AISA -2A

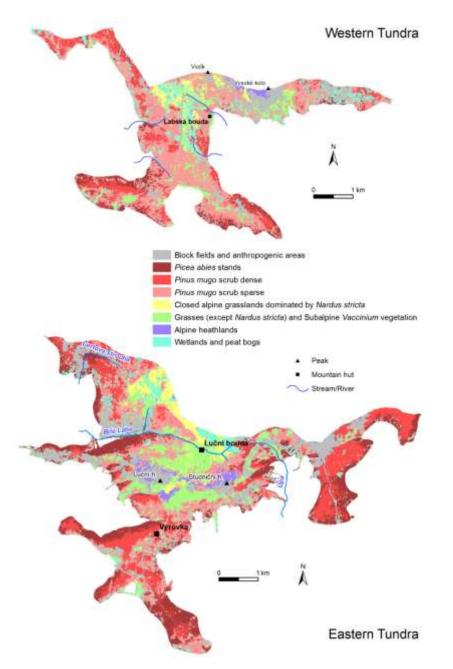
#### APEX

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		Overall			Overall	Карра		Overall	
	Classification	accuracy	Карра	<b>Classification method</b>	accuracy (%)	coefficient	Classification	accuracy	Карра
	method	(%)	coefficient	РСА	- 40 bands		method	(%)	coefficient
PCA - 40 bands				SVM	84.31	0.81	All bands 10 and 20 m		
	SVM	82.59	0.79	OBIA SMV RBF	80.66	0.77	NN user defined	58.27	0.52

Data and classification method	pixel-based classification						OBIA	
	APEX (SVM 40 PCA bands)		AISA (SVM 40 PCA bands)		Sentinel-2A (NN)		AISA (SVM 7 PCA bands)	
Class	PA (%)	UA (%)	PA (%)	UA (%)	PA (%)	UA (%)	PA (%)	UA (%)
1. Block fields and anthropogenic areas	98.60	100.00	99.25	98.58	92.68	95.00	99.93	95.11
2. Pinus mugo scrub	99.86	94.78	99.96	98.45	100.00	88.51	100.00	99.36
3. Subalpine Vaccinium vegetation	8.54	100.00	63.90	50.19	65.38	45.95	53.15	87.66
4a. Nardus stricta stands	73.44	86.01	83.73	71.38	46.02	54.17	79.24	73.27
4b. Species-rich vegetation with high cover of forbs	86.84	44.59	55.32	60.00	50.00	35.29	81.22	33.06
5a. Calamagrostis villosa stands	63.95	49.74	55.03	87.29	31.82	43.75	76.20	82.62
5b. Molinia caeruela stands	64.54	59.87	66.78	75.22	15.00	60.00	79.15	44.94
5c. Deschampsia cespitosa stands	87.31	68.25	85.10	85.81	57.50	26.44	63.49	89.76
6. Alpine heathlands	90.36	82.14	81.60	83.80	37.78	42.50	66.11	73.13
7. Wetlands and peat bogs	58.56	80.30	63.46	86.74	56.76	91.30	40.24	85.07
8. Water areas	100.00	100.00	98.80	100.00	х	х	100.00	100.00

#### Results for simplified legend – Sentinel-2A



Classification method	Overall accuracy (%)	Kappa coefficient					
all bands 10 a 20 m simplified legend							
MLC	77.73	0.74					
SVM	70.99	0.67					
NN (default)	76.21	0.73					

### Conclusions

- Best classification results for the hyperspectral data with the highest spectral and spatial resolution, i.e. AISA Dual data, comparable for APEX data
- Best results both types of hyperspectral data: SVM classifier
- Best results Sentinel-2A data in the case of simplified legend, NN and MLC methods achieved better results than SVM.
- Important definition of legend categories different for different spatial resolutions
- We have to improve classification accuracy of grassland categories
- Results for Sentinel-2A promissing, especially for Sentinel-2A in tandem with Sentinel-2B in time series
- Next improvement UAV with hyperspectral sensor high spectral and time resolution, biophysical parameters (chlorophyll, fAPAR, biomass, LAI etc.), upscaling to Sentinel
- Earth observation powerful tool for tundra ecosystem monitoring, management and presevation

### THANK YOU FOR YOUR ATTENTION!



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Article in European Journal of Remote Sensing: http://www.tandfonline.com/doi/full/10.1080/22797254.2017.1274573

#### ACKNOWLEDGEMENT

We a sup of

This research was made possible by the Charles University in Prague project GAUK No. 938214 and Ministry of Education, Youth and Sports of the Czech Republic project NPU I LO1417 Our thanks belong also to botanists Stanislav Březina and Jan Šturma for their help during the fieldwork.

SCERIN-5 Capacity building workshop, Pécs Hungary, June 2017