

UAV BASED MULTI SEASONAL DECIDUOUS TREE SPECIES ANALYSIS IN THE HAINICH NATIONAL PARK USING MULTI TEMPORAL AND POINT CLOUD CURVATURE FEATURES

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ABSTRACT:

Low cost UAV systems are a flexible and mobile platform for very detailed spatial high-resolution point cloud and surface height mapping projects. This study investigates the potential of the DJI Phantom 4 Pro 3D point clouds and derived crown surface height information in combination with RGB spectral information for mapping of deciduous tree species in the Hainich national park area. RGB image data was captured in August, early October and November 2018 to create a multi seasonal spectral dataset for a 100 ha test area. The flight campaigns were controlled from the Hainich flux tower platform in 40 m height owned and operated by University of Göttingen in the central part of the park area. Absolut georeferencing accuracy of the datasets was improved using 7 DGPS measured control points within the stand structure on small forest clearings. Image files and ground control points were processed to a dense point cloud model with 2.6 billion points (approximately 200k points per tree crown object) using the Agisoft Metashape cluster processing environment. Additionally, a digital surface model and a true ortho image mosaic with 3 cm spatial resolution was generated. For the differentiation of deciduous tree species, a reference data set with coordinates for the tree species *Fagus sylvatica* (beech), *Fraxinus excelsior* (ash), *Acer pseudoplatanus* (sycamore maple), *Carpinus betulus* (hornbeam) and dead trees and early defoliated trees was defined. The study site is however dominated by *Fagus sylvatica* and *Fraxinus excelsior*. We studied two different groups of features: tree crown surface height variability parameters using point cloud densities, point cloud height variance, local standard deviation of gaussian curvature, standard deviation of local point cloud roughness and multi temporal normalised spectral features using multi seasonal uncalibrated UAV RGB data. Analysis of feature separability showed that very high-resolution point cloud surface curvature properties with small neighbourhood radii can differentiate some tree species types but we also found multitemporal spectral ratios based on RGB data to be very successful in differentiating the main tree species.

Results of this work show that super fine very dense point cloud models and derived roughness measures of mixed forest stand surfaces hold valuable information for deciduous species discrimination and will likely also be very useful for morphological analysis of tree crown types.

1. INTRODUCTION

Small Off The Shelf (OTS) quadro copter systems provide a very cost effective, highly flexible, mobile platform for high resolution point cloud mapping. Various studies showed the full potential of these small and flexible platforms (Mlambo et al. 2017). An overview is also given in Colomina and Molina (2014). Especially in very complex 3D environments the automatic obstacle avoidance, low copter weight, long flight times and precise manoeuvring are important advantages of small OTS systems in comparison with larger octocopter systems. Especially for the mapping of 2.5D structures - mapping a continuous height surface of a stand and not volume 3D structures - the airborne Lidar and the image-based UAV Structure from Motion (SfM) approach create comparable surface representation results (Thiel and Schmulius 2016). Since the development of very fast Graphic Processing Units (GPU) with multiple processing cores the computing times for image matching have decreased considerably making projects with hundreds to thousands of

full-resolution 12-24 megapixel image files possible now. Studies in the past have shown that this is possible even with un-georeferenced images from various sources and with varying image quality (Snavely, Seitz and Szeliski 2008). Photogrammetric grade cameras have been used together with GPS and IMU (Inertial Measurement Unit) data to extract camera position and orientation data and various applications with very high-resolution airborne data have been published (Ofner et al. 2006). With SfM approaches the 3D modelling is done directly based on multiple overlapping images and position/orientation data is not necessarily needed. These concepts do not require any orientation information although orientation and GPS information is an advantage for fast data processing. The applied concepts usually use features that are independent of dissimilarities between overlapping images. Namely the SIFT and the SURF operator (Bay et al. 2008, Lingua, Marenchino & Nex 2009, Lowe 2004) play a critical role for the successful introduction in various 3D point cloud mapping software solutions. First publications go back into the 1990s and earlier (Lowe 1999 and Ulman 1979). The SfM

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concept is exemplary summarized e.g. in Carrivick et al. (2016). Detailed summary of the overall concept is also available in Smith et al. (2015). While the concept is very flexible it inherits some problems when point clouds are analysed and compared to ALS/TLS (Airborne/Terrestrial Laser Scanning) datasets. The density of the point cloud is not necessarily proportional linked to the surface density of a complex object surface. It is therefore difficult to compare the ALS/TLS point clouds of deciduous and coniferous trees with those coming from SfM concepts.

This study examines the potential of very small UAV-Systems (small sized Unmanned Aerial Vehicles below 5 kg) for detailed multi-seasonal tree canopy mapping using both spectral features and point cloud height-based features of a 100 ha area in the Hainich national park. While spatial very high-resolution analysis of tree species types has been conducted long before e.g. in Hirschmugl et al. (2007) there is not much research on point cloud densities and variabilities of the individual crown surface roughness and curvature of crown surface subsets for larger stands available so far. This is mainly due to the lack of availability of very dense point clouds with full modelling of the crown morphology for larger areas and continuous forest stand structures in the cm resolution range. This data could be very important for the validation of components of forest growth models (Pretzsch, Forrester and Rötzer 2015) or for measuring plant interception volume. A lot of work has been done investigating Laser scanning data for species differentiation (Hollaus et al. 2009, Korpela et al. 2010, Kim et al. 2009, Orka et al. 2009, Othmani et al. 2011, Person et al. 2004, Reitberger et al. 2006 and Brandberg 2007).

Gap fraction and tree roughness were investigated with TLS data by Antonarakis et al. (2010) and surface roughness on stand level scale was e.g. analysed with TLS/ALS data by Weligepolage, Gieske and Su (2012). Spatial very high-resolution TLS data f.e. was used by Othmani et al. (2013) to classify tree species by 3D bark texture. Published work by Pretzsch (2014) showed that in mixed stands the crown projection area of beech trees is less circular and crown eccentricity is increased. Fractal dimension of the mixed crown surface area and leaf area is changed “from lower space filling (Euclidian scaling) towards higher space filling of the crown volume (fractal scaling)” (Pretzsch 2014). In mixed stands this could modify the crown surface roughness of different tree species. Crown morphology and canopy structure as a result of environment and its impact on growth of the various species are at the same time the important driver and the product of stand dynamics (Pretzsch 2014). Crown overlap in multiple layers in mixed stands clearly complicates any analysis of crown surface parameters.

2. DATA AND STUDY SITE

The Hainich national park (“Nationalpark Hainich”) was founded December 31, 1997 and is the only national park in Thuringia. Main objective is the beech forest and fauna protection and the restoration of a primordial status. The park area covers 75 km² in the western German state Thuringia. The national park is a part of the largest deciduous forest region (160 km²) in Germany broadly named “Hainich”. The study site is situated in the central part of the national park and forms a rectangular 100 ha shape surrounding the flux measurement tower of the University of Göttingen (Figure 1). The forest is dominated by beech trees (*Fagus sylvatica*), ash trees (*Fraxinus Excelsior*) and sycamore maple (*Acer Pseudoplatanus*), other tree species are field maple, hornbeam

(*Carpinus Betulus*), wych elm (*Ulmus Glabra*), lime trees, Norway maple, cherry and oak. The forest is completely unmanaged and includes stands of flowers and fungi and various animal species (wildcats, various species of bats, wood peckers and wood beetles and also wild pigs).

For this work the DJI Phantom 4 Pro (Table 1) was used, flight altitude above crown level was 100 m, X/Y-overlap 74%, resulting spatial resolution was 0.03 m (3 cm).

	31.8.2018	15.10.2018	7.11.2018
No. of images	625	624	612
Resolution m	0.03	0.03	0.03
Altitude in m	100 m (relative to start altitude)		
Camera-system	Phantom 4 Pro, DJI FC6310, full frame equivalent: 24mm, f2.8, 20MP, 1-inch DJI FC6310 RGB (Bayer) sensor.		
Overlap in %	74/74		
DGPS points	7	0	0
Exposure (fix)	1/400s f5.6	1/400s f4	1/500s f5
ISO	400	200	400
Flight time	31.8.2018 12:04-12:51	15.10.2018 13:49-14:31	7.11.2018 12:23-13:04
File format	JPEG with EXIF GPS location & barometric height & WGS84 altitude		

Table 1: Hainich flight plan information for the August, September and November flights over the “Huss”-area.

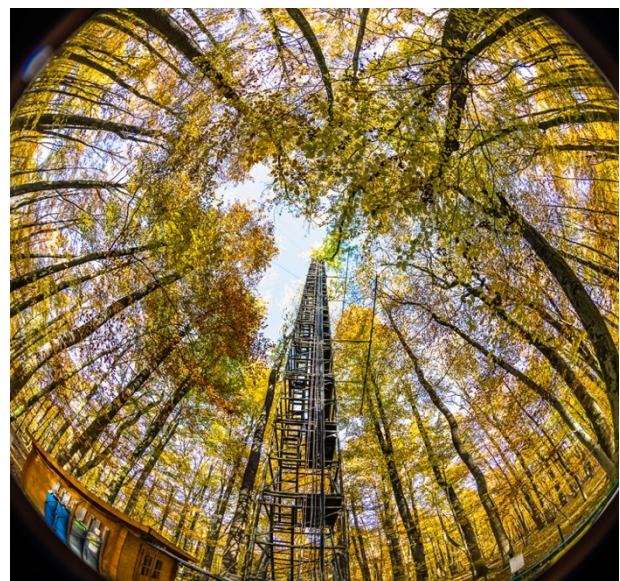


Figure 1: Flux tower installation in the Hainich national park area photographed in November 2018 from the forest ground level (fisheye).

3. METHODS AND RESULTS

3.1 UAV Data Pre-Processing

SfM (Structure from Motion) methods for data processing to georeferenced dense point clouds, digital surface model and ortho image mosaic level were used in an Agisoft Metashape cluster environment (www.agisoft.com, Agisoft LLC, St.Petersburg, Russia). Description of the Agisoft Metashape workflow can be found e.g. in Turner, Lucieer and Wallace (2014) or in Dandois and Ellis (2013). Agisoft implements digital photogrammetry techniques together with computer vision methods.



Figure 2: Subset of the study area with RGB spectral properties (uncalibrated) from 31.8., 15.10. and 7.11.2018 and digital surface model (DSM) below.

After initial feature matching an iterative bundle adjustment is applied and positions of the cameras and the 3D positions of the features are measured.

The August dataset was spatially corrected using DGPS (Stonex GNSS) measured reference GCPs. These GCPs were measured on highly reflective ground targets and later found in the relevant images and integrated into the Agisoft workflow as reference points. The October and November data mosaics were geometrically co-registered to the August reference dataset. Agisoft data processing was done using the full resolution image files for initial image alignment and dense point cloud matching (compare with the processing configuration in Table 2). All three dataset types – the SfM acquired point clouds, the digital surface model (DSM) and the ortho mosaic were exported to standard file formats (LAZ – compressed LAS format and PCIDSK 16 bit raster format) (Table 3 and Figure 2).

The point cloud model from August was used for the morphological crown analysis, the multitemporal dataset from August, October and November was used together for crown based spectral and textural feature analysis.

A Digital Canopy height Model (DCM) was produced applying simple terrain model subtraction. The digital terrain model (DTM) was calculated based on local minima mapping in 120 m search distance of the point cloud model. The point cloud model was exported to LAZ format including colour information and height-based ground/vegetation surface classification information. Reference data for 50 tree crowns was provided by the Hainich national park administration and extended visually to 100 tree crowns in the dataset, partly

using the Huss-area reference information from 2013 (Figure 5) from Huss and Manning (2016).



Figure 3: Perspective view of the dense point cloud model (full resolution point cloud with 2.23 billion points on approx. 100 ha) with attributed ortho mosaic RGB values. Hainich national park (partly showing the “Huss area”).

Python Task	Metashape Processing Parameter	
Chunk.match photos Chunk.align cameras ()	Accuracy:	High (full resolution image files)
	Generic/Reference preselection:	yes/yes
	Key point limit:	0 (no limit)
	Tie point limit:	4000
	Adaptive camera model fitting:	yes
Chunk.build Depthmaps / chunk.build densecloud ()	Quality:	“Ultra quality” (full resolution files)
	Filtering	Disabled (no point filtering)
	Point color	yes
Chunk.build dem / chunk.build orthomosaic()	Blending mode	Mosaic
	Enable hole filling	yes
	Interpolation	yes
	Coordinate system	WGS84/UTM32N
Tweaks	Enable fine level task sub-division Metashape.app.gpu mask=3	

Table 2. Agisoft Metashape 1.5.1. cluster processing setup for the sub processes “Align Photos”, “Build Dense Cloud”, “Build DEM”, “Build Ortho Mosaic”.

Datasets	Dataset Specifications
Ortho Mosaic	Spatial resolution: 3 cm, WGS84 UTM32, Bayer RGB, 7 DGPS GCPs RMSE: 11,5 cm. Okt./Nov. data is polynomial co-registered to the August ortho mosaic.
DSM /DTM and derived nDSHM	Spatial resolution: 3 cm (32R), UTM32N, WGS84
Point cloud model (full/medium/low resolution, low res. ground points classified using 120 m regional local min filtering)	Full: 2.600 000 000 points on 100 ha = 26 mio/ha (approx. 2600 points/m ²) Approx. 200 000 -260 000 points/tree crown, Medium: 273 million points Low: 16 million points (laz file format, CloudCompare binary entities format), WGS84 UTM32N.

Table 3: UAV based georeferenced datasets processed from 612/625 images using Agisoft Metashape.



Figure 4: UAV start from the flux tower platform 40 m above ground level in the Hainich national park.

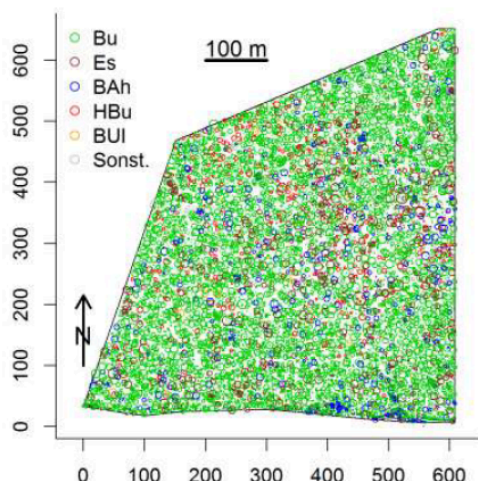


Figure 5: “Huss”-area (defined in summer 1999 by J. Huss and Smaltschinski – University of Freiburg) and individual tree crown reference information in the Hainich national park (status 2013). The main tree species (Bu/Es/BAh) (Nationalparkverwaltung Hainich, 2016. Exkursionsführer – Wissenschaft im Hainich, Hainich-Tagung 27-29.4.2016 Bad Langensalza, PDF download).

3.2 High Resolution Point Cloud and Multi-temporal Image Band Ratio Features

Image object and point cloud feature analysis was started after initial object-based shadow masking and ortho image sub-segmentation into 6.25 m² crown objects. The nine spectral channels (uncalibrated RGB information from all flight campaigns) were used to calculate a multitemporal shadow mask based on mean brightness value thresholds in every month for image objects larger 0.1 ha. The shadow mask defined gaps in the stand structure and was later excluded from further tree species crown analysis. The remaining crown surface area was re-segmented into a grid cell system of 2.5 x 2.5 m². Reference tree crown objects for each species were selected using data from the national park

administration and extended with data from the Huss-area and by manual delineation of additional trees for the class “dead trees”. For these objects we analysed different spectral and image-morphological as well as height-variance specific features, point cloud height variance and surface roughness features. The following features (Table 4) were zonally calculated for the dominating tree species: 1. beech (*Fagus sylvatica* - BU), 2. ash (*Fraxinus excelsior* - ES) and 3. sycamore maple (*Acer pseudoplatanus* - BAH), 4. hornbeam (*Carpinus betulus* - HBU), 5. dead tree crowns (Dead) and 6. early November defoliated tree crowns (mainly ash) – DF). Other tree species are field maple, wych elm (*Ulmus Glabra*), lime trees, Norway maple, cherry and oak trees.

Feature type	Description
Stdev (SD) object RGB	Stdev (SD) from 8Bit uncalibrated DN (RGB)
Mean object RGB DN	8Bit uncalibrated DN
Haralick texture measures: Homogeneity & Angular Second Moment (Ortho mosaic)	Omni directional and in 0°, 45°, 90° and 135 ° (object based)
Normalized Difference Green-Red Index (ortho mosaic) (NDGRI).	$NDGRI = \frac{(Gr - Re)}{(Gr + Re)}$
Multi Temporal Green Red Index Difference (MTGRID), three point convex-concave autumn coloration index (-1 to 1).	$MTGRID = \frac{(NDGRI_{aug} + NDGRI_{nov})}{2} - NDGRI_{oct}$
Local gaussian surface curvature per point in 1, 2 and 6 m radius neighbourhood, resampled point cloud to 16 mio. points, SD.	K based on a sphere surface, $K = 1/r_1 * 1/r_2 = 1/r^2$, inverted, unsigned, calculated from quadric function fitting.
Surface roughness (in a re-sampled point cloud)	Point distance to the plane of neighbourhood points.
Point density (number/area)	Points per m ²
Point height variability (240 million / 2260 million) SD.	Height SD in 2.5 x 2.5 m zones.

Table 4: Spectral, point height and surface shape-based feature descriptions of the Hainich forest area for reference tree species plots.

3.3 Results

Spectral separability analysis was calculated for the 6.25 m² sized image objects with J-M separability – (Jeffries-Matusita Distance - Bruzzone et al. 2005). The Normalized Difference Green-Red Index (NDGRI) (Table 4) (2) also named GRND (Green-Red Normalised Difference) (Tucker 1979, Hunt et al. 2015) showed best separability results with the November data (Table 5). The spectral data from November clearly differentiated ES, BU and HBU and to a certain degree also BAH. This was expected since the timing of the reflectance change in autumn is very much tree species specific. Dead and November-defoliated crowns clearly are distinguishable using the full NDGRI series (Figure 6). Point density and point cloud height variability (standard deviation) did not differentiate the different crown species types in August (compare with Figure 7). Multitemporal point cloud height analysis however indicated that late autumn tree crown point clouds show the starting defoliation with increased point height variability with some deciduous tree species types.

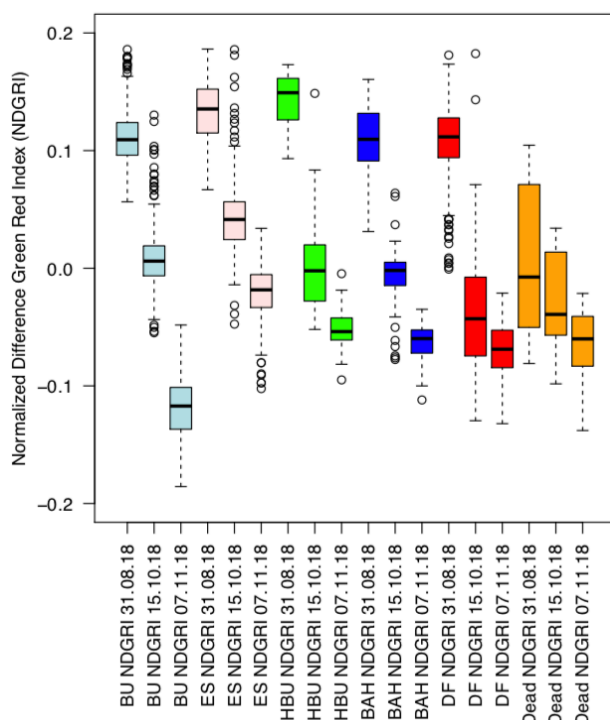


Figure 6: Normalized Difference Green-Red Index values for August, October and November RGB data of the dominating tree species (beech (BU), ash (ES), maple (BAH), hornbeam (HBU) and November-defoliated (DF)/ dead crown class.

J-M	HBU	BAH	ES	DF	DEAD
BU	1,37 0,16 0,47	1,20 0,08 0,04	1,72 0,37 0,21	0,82 0,50 0,07	0,84 0,35 1,15
HBU	-	0,12 0,14 0,45	0,42 0,32 0,05	0,18 0,20 0,43	0,12 0,11 1,44
BAH	-	-	0,83 0,61 0,05	0,08 0,21 0,00	0,12 0,16 0,99
ES	-	-	-	0,79 0,91 0,21	0,58 0,76 1,32
DF	-	-	-	-	0,02 0,03 0,94

Table 5: Separability analysis (Jeffries-Matusita Distance) for the main tree species (Normalized Difference Green-Red Index for November (above), October(mid) and August (below) 2018 data).

Results from the multi temporal point cloud analysis are not included in this paper but will be published at a later stage of this project.

3.3.1 Point cloud height variability

Results from the August point cloud height variability analysis indicate a higher variability for HBU compared to other tree species. Exceedingly high variability with August defoliated crowns (Figure 7) is expected. Point density per sqm however is difficult to attribute to different vegetation types and vegetation densities. Variability in the medium resolution point cloud model (273 million points) is between 100 and 800 points per sqm but variability clearly overlaps for the different main tree species (Figure 7) and can also be attributed to illumination condition changes variability of feature density and variations in image overlap (image overlap

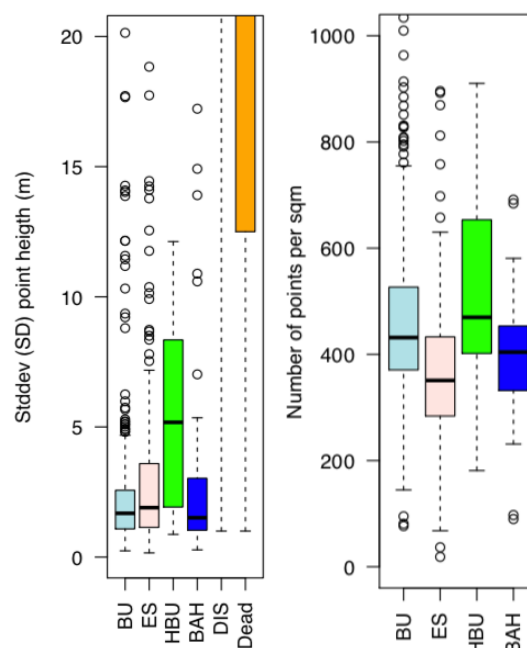


Figure 7: Variability of the point cloud height information sampled over 2.5 x 2.5 m² plots for the dominating tree species (beech (BU), ash (ES), maple (BAH), hornbeam (HBU) and defoliated / dead tree classes.

for this project was however kept constant at 74% along track and 74% across track).

Point density is therefore a weak spatial feature for this kind of analysis and also shows multi modal distributions.

3.3.2 Local gaussian surface curvature

For this analysis we also investigated gaussian surface curvature K and surface roughness descriptions calculated for different point cloud neighbourhood distances (radiuses of the neighbourhood). Gaussian curvature is usually calculated as a product of k_1 and k_2 ($K = k_1 * k_2$). The mean curvature is calculated as $H = \frac{1}{2} * (k_1 + k_2)$. Gaussian curvature can be either negative (hyperbolic surface) or positive (elliptic surface). In our study we used an implementation with k_1 and k_2 identical in a concept with a computed radius from a best fitting of a quadric to the neighbourhood points ($K = 1/r_1 * 1/r_2$).

Low neighbourhood distance calculations create a stronger variability between different surface curvatures on a surface. This can be explained with the smaller neighbourhood kernel that can also describe stronger curvature values (smaller sphere radiuses) but can also be influenced by noise. Very large neighbourhood spheres likely will even-out small local surface height variabilities and noise and will generally also create smaller curvature radius values. Since the gaussian surface calculation for very dense point clouds with larger radius definitions is computational very expensive we reduced the surface density to a thinned point cloud with 16 million points (random thinning to 1 point per 10x10 cm²). We measured curvature for a 1 m, 2 m and 6 m sphere radius for every point for all deciduous tree species training areas (Figure 8-9). Surface curvature variability was measured using “Standard Deviation” (SD) of the curvature within every 2.5 x 2.5 m² tree crown reference plot area. The resulting SD is provided in Figure 8 and 9. Boxplots show information about the data distribution with indicated median, upper and lower quantile and maximum/minimum of the data distribution (for

values within the 1.5 factor limit of the inter-quartile distance) – values outside this range are shown as outliers. Since the measured values of curvature are not normal-distributed the boxplot visualisation is not fully adequate here. In Figure 8 the median values are clearly shifted and various outlier clusters are visible. We also provide histograms in A1 (Appendix) that show in detail the multi modal data distribution of the point cloud curvature SD values and point cloud height SD values. A radius of 1 meter around every point indicates as expected also slightly better separability compared to 2 m and 6 m radius but the absolute variability also increased. The 6 m radius is clearly already detecting the full crown surface curvature and is not sensible anymore to surface roughness (figure not provided due to paper length restrictions).

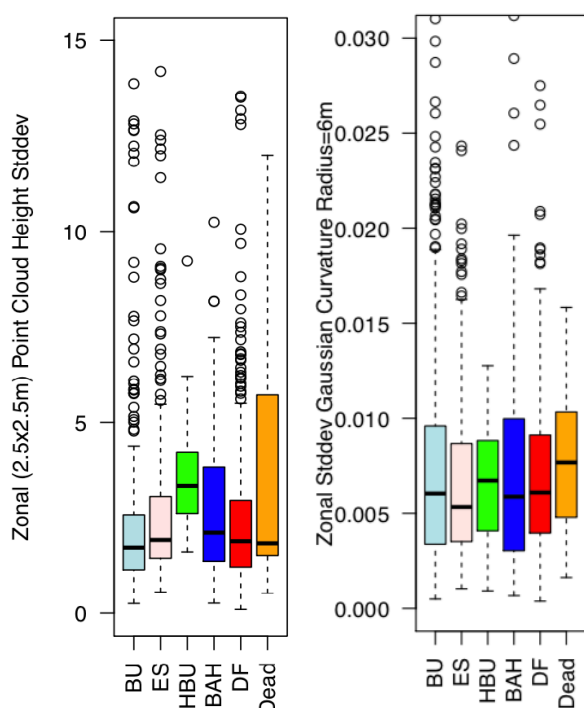


Figure 8: Local point cloud height variability SD (left) and gaussian curvature measures SD (right) investigating a radius of $r=6m$ neighbourhood sphere. Values in $1/r$ (radius) using quadratic fitting method. Zonal curvature variability (SD) calculated on $2.5\text{ m} \times 2.5\text{ m}$ crown surface regions.

The position of the training areas relative to the crown surface border is critical for crown surface shape parameterisations. Gaussian surface curvature also is best calculated on continuous surface types. SfM derived tree crown point clouds usually show a three-dimensional character with penetration into the crown surface. We did not account for this variability. The surface thinning was done using a random point selection. Relative position to the crown centre is also a relevant property of the point curvature value especially for the curvature values with large neighbourhood spheres. However, in Figure 9 we can demonstrate that gaussian curvature SD is clearly linked to the 3-dimensional character of the crown surface and changes with different crown species. Curvature and height statistics were calculated on the basis of the August flight campaign point cloud and crown closure was high at this moment. The point cloud height variability for the different tree species indicates higher variability with dead and defoliated crown types (class “Dead” in Figure 8).

November defoliated tree crowns are not showing this variability since the height map comes from August.

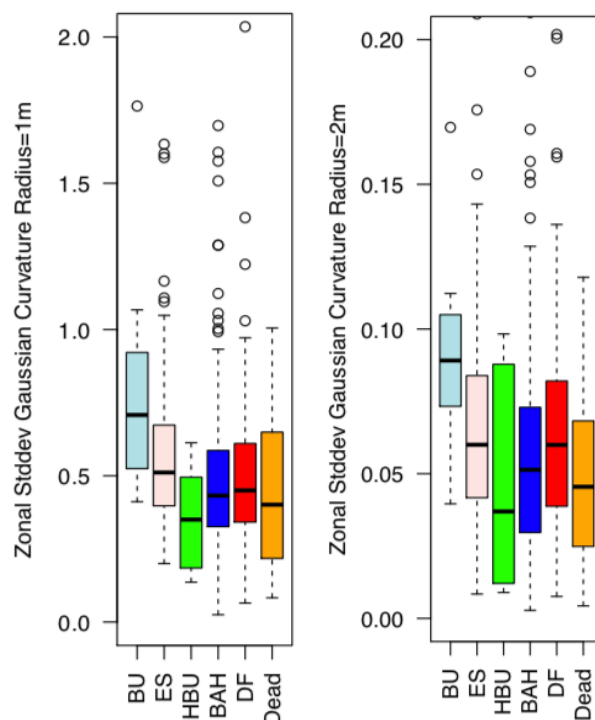


Figure 9: Local gaussian curvature measures investigating a radius $r=1m$ (left) and $r=2m$ (right) neighbourhood sphere. Values in $1/r$ (radius) using quadratic fitting method. Zonal curvature variability (SD).

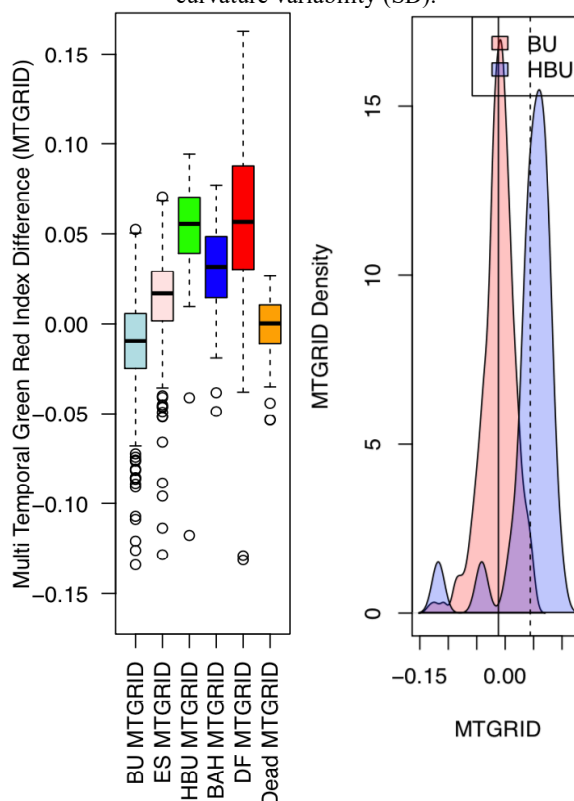


Figure 10: Multi Temporal Green Red Index Difference (MTGRID) feature for the quantification of the green-red spectral defoliation change signature (negative= convex shape, positive=concave shape).

3.3.3 Multi Temporal Green Red Index Difference

We developed a Multi Temporal Green-Red Index Difference feature (MTGRID) (1) based on the NDGRI (2) to quantify species specific differences of the change in green and red reflectance in the autumn colouring period between August and November.

$$MTGRID = \frac{(NDGRI_{aug} + NDGRI_{nov})}{2} - NDGRI_{oct} \quad (1)$$

where NDGRI = normalized difference green red index.

This difference index measures the amount of deviation of the October spectral response measured with a normalized Difference Green Red Index (NDGRI):

$$NDGRI = \frac{(Gr - Re)}{(Gr + Re)} \quad (2)$$

where
Gr = green channel
Re = red channel

from the mean of August and November – showing deviations from the expected (but not necessarily true) linear trend between August and November given that the time difference is identical and spectral change is linear with time. Negative values describe a convex progression from August to November whereas positive values indicate a concave change. This difference index is based on ratio indices and therefore creates transferrable and robust result – however the temporal dimension is critical for a comparison with other data sets and MTGRID should be calibrated for non-equal acquisition intervals with a dedicated weighting coefficient if used for comparisons. Results (Figure 10) indicate a good separability of the different main tree species using MTGRID.

4. SUMMARY & CONCLUSIONS

This study demonstrates that multi seasonal (late-August, mid-October, early-November) RGB data from Bayer-sensors with limited spectral resolution and very high-resolution point cloud height information from low flying UAV-systems can be used to extract fine scale information to differentiate deciduous tree species types. We found multi-seasonal spectral data dimensionality compressed to a multi temporal difference ratio (MTGRID) to be sufficient to differentiate 4-6 deciduous tree crown types. Spectral data from three flight campaigns was converted to a normalized difference multi seasonal green-red ratio. Additionally, local crown surface roughness was parametrized using a local roughness and a local gaussian curvature descriptor in three search distances. These features are promising for multi seasonal forest stand surface shape description that could potentially also characterise the progression of defoliation in late autumn. This study was computing hardware limited since derived 3D-features for very large 3D point clouds for a full forest stand region are computational expensive to process. We therefore reduced the point cloud density for the gaussian surface curvature analysis and for roughness calculations from 2.2 billion points to 16 million points. This clearly limited the potential roughness parametrisation. So far, we did not integrate multi seasonal point cloud analysis but change of surface properties due to defoliation are likely powerful features. Summer 2018 was an exceptional warm and dry summer in Germany and defoliation started much earlier than in previous years. Results of the multi temporal analysis are likely not transferrable to other years – even given the same

acquisition times are used. More reference data will be included in this study to differentiate more crown species types in 2019. More work is also planned with morphological crown features from leaf-off data captured March 2019.

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APPENDIX

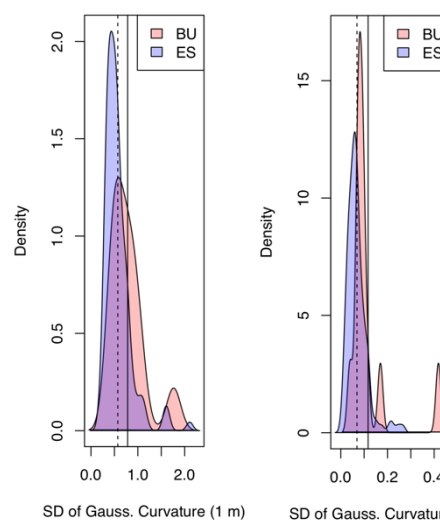


Figure A1: Histograms of SD for 6.25 m² tree crown sub regions of gaussian curvature for 1 m and 2 m neighbourhood radius (beech and ash trees).