FOREST INVENTORY ATTRIBUTE ESTIMATION USING AIRBORNE LASER SCANNING, AERIAL STEREO IMAGERY, RADARGRAMMETRY AND INTERFEROMETRY–FINNISH EXPERIENCES OF THE 3D TECHNIQUES

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Commission VI, WG I/2

KEY WORDS: Remote sensing, airborne laser scanning, aerial stereo imagery, SAR radargrammetry, SAR interferometry, random forest, sample plot, area-based approach, forest inventory, forest management planning, GIS

ABSTRACT:

Three-dimensional (3D) remote sensing has enabled detailed mapping of terrain and vegetation heights. Consequently, forest inventory attributes are estimated more and more using point clouds and normalized surface models. In practical applications, mainly airborne laser scanning (ALS) has been used in forest resource mapping. The current status is that ALS-based forest inventories are widespread, and the popularity of ALS has also raised interest toward alternative 3D techniques, including airborne and spaceborne techniques. Point clouds can be generated using photogrammetry, radargrammetry and interferometry. Airborne stereo imagery can be used in deriving photogrammetric point clouds, as very-high-resolution synthetic aperture radar (SAR) data are used in radargrammetry and interferometry. ALS is capable of mapping both the terrain and tree heights in mixed forest conditions, which is an advantage over aerial images or SAR data. However, in many jurisdictions, a detailed ALS-based digital terrain model is already available, and that enables linking photogrammetric or SAR-derived heights to heights above the ground. In other words, in forest conditions, the height of single trees, height of the canopy and/or density of the canopy can be measured and used in estimation of forest inventory attributes. In this paper, first we review experiences of the use of digital stereo imagery and spaceborne SAR in estimation of forest inventory attributes in Finland, and we compare techniques to ALS. In addition, we aim to present new implications based on our experiences.

1. INTRODUCTION

The retrieval of forest inventory attributes, which are needed in forest management planning, is carried out using airborne laser scanning (ALS)-based inventory methodologies in the Nordic countries. ALS-aided forest inventory methodologies were taken into practice shortly after demonstrative studies (e.g., Nilsson 1996, Næsset 1997a, b, 2002, Hyyppä and Inkinen 1999, Hyyppä and Hyyppä 1999). In operational mapping of forest inventory attributes, a two-stage procedure using ALS data and field-measured sample plots, i.e., an area-based approach (ABA, Næsset 2002), has become common and a reference against which other inventory methodologies are compared (White et al. 2013). The ABA can provide precise predictions for many required forest inventory attributes, including stem volume and height. ABA is sampling based, and it is possible to calculate accuracy statistics. In addition, forest attributes are predicted for a grid (e.g., 16x16 or 20x20 m), and thus, ALS-based inventory does not depend on stand boundaries.

ALS is a method based on light detection and ranging (LiDAR) measurements from an aircraft, where the precise position and orientation of the sensor is known, and therefore the position

(x, y, z) of the reflecting objects can be determined. Due to rapid adaptation to operational forest inventory, ALS has been the primary data source for three-dimensional (3D) information on forest vertical structure. However, there is an increasing interest in the use of high-spatial-resolution digital aerial imagery (digital stereo imagery, DSI) or radar imagery to generate 3D information analogous to ALS data to support forest inventory and monitoring (Wulder et al. 2013, White et al. 2013, Vastaranta et al. 2013, Holopainen et al. 2014). This interest in alternative technologies for acquiring accurate height information can be attributed to the need to control cost and improve temporal resolution at the same time. At the moment, imagery is about one-half to one-third of the cost of ALS data. High resolution radar data is still as expensive as ALS, but there is a drift toward a free-of-charge data policy. The strength of SAR is the temporal resolution. In theory, SAR data can be obtained on nearly a daily basis. In addition, alternative and complementary data sources are searched because certain forest inventory attributes remain difficult to obtain with ALS, such as tree species composition.

ALS is superior compared to aerial imaging or radar in the mapping of terrain height in forested areas. A laser pulse hit on the forest canopy can produce one or more returns. The first returns are mainly assumed to come from the top of the canopy and the last returns mainly from the ground. However, in many jurisdictions, detailed ALS-based digital terrain models/digital elevation models (DTM/DEM) are already available, and that enables also linking photogrammetric or SAR-derived heights to heights above the ground. In other words, in the mapping of forest inventory attributes, detailed DTM is a prerequisite if alternative techniques are used.

This paper summarizes the first results from the use of digital stereo imagery and spaceborne SAR in forest inventory attribute mapping in Finland on areas where detailed ALS-based DTM is available. Alternative 3D techniques are also compared to ALS in regard to data acquisition, processing, point cloud/DSM quality and in obtained estimation accuracies. In addition, we aim to present new insights based on our experiences. We begin with a short overview of the alternative 3D techniques (Section 2). Specific study areas and methods are then presented in Section 3. In Section 4, we compare and discuss data acquisition and processing, point cloud/DSM quality and obtained prediction accuracies in estimation of forest inventory attributes, followed by a conclusion in Section 5.

2. ALTERNATIVE 3D TECHNIQUES

2.1 Aerial imagery

The creation of image-derived digital surface models requires high-resolution aerial images with stereo coverage (Hirschmugl 2008, Leberl et al. 2010). When an object is imaged from two different perspectives, stereophotogrammetry enables the measurement of its three-dimensional position relative to a reference datum (e.g., sea level). The use of digital aerial cameras has enabled a substantial increase in the number of overlapping images that are acquired for ongoing forest inventory or monitoring programs. In forested regions, the availability of many overlapping images provides the multiimage information required to produce a DSM and reduces the impact of occlusions (i.e., shadows), which occur more frequently when there is less image overlap (Haala et al. 2010). The film-to-digital transition has resulted in improvements to the radiometric properties of the images, while advances in computing technology have made complex algorithms for image matching practical (Leberl et al. 2010). These technological advances have greatly enhanced the quality of DSMs derived from stereophotogrammetric processing, improving the characterization of detailed structures. The digital image resolution is defined by the ground sampling distance (GSD), which depends on various factors, most importantly the flying height and the specifications of the camera (instrument) used. Flying heights between 550 m and 4800 m have been used with 60-90% forward overlap and 30-60% side overlap for forestry applications, resulting in GSDs ranging from 0.05 m to 0.5 m (e.g., Hirschmugl 2008, Bohlin et al. 2012, Järnstedt et al. 2012, Nurminen et al. 2013).

2.2 Spaceborne SAR

Synthetic aperture radar (SAR) is active imaging radar operating in the microwave region of the electromagnetic spectrum. Spaceborne SAR systems are typically pulse-based radars, i.e., a pulse of microwave radiation is transmitted, and echoes coming back from the target area are received. As the satellite moves in its orbit, a two-dimensional image-like representation can be processed from the received echoes.

With current SAR systems, images with resolutions of about 1 m can be obtained from satellites orbiting at altitudes of several hundred kilometers. A major advantage of radar images, compared with optical region satellite images, has been their availability (temporal resolution) under varying imaging conditions.

Radargrammetry and interferometry are basic techniques used to derive 3D information from radar imagery. Radargrammetry is based on the stereoscopic measurement of SAR images in which, analogously to photogrammetric forward intersection, two or more radar images with different viewing perspectives are used to extract 3D information from the target area. The geometrical baselines of stereo pairs are typically in the order of hundreds of kilometers. Although radargrammetry has been a well-known technique for many decades, it has gained new recognition due to the new SAR satellites with enhanced spatial resolution and very high geolocation accuracy (Peerko et al. 2010, Raggam et al., 2010).

SAR interferometry is based on the phase differences between two complex SAR images acquired from slightly different view angles. Geometrical baselines are typically in the order of a few tens or hundreds of meters. Phase differences are then converted to height differences in the target area. By using a short radar wavelength such as the X band radiation, backscattering from the canopy is obtained in forested areas, and the canopy height can be estimated when ground surface elevations are known. It should be noted that due to signal penetration in the target, there is underestimation of the actual height (Praks et al. 2012).

3. STUDY AREAS AND USED METHODS

The use of aerial imagery and spaceborne SAR in forest inventory attribute estimation has been studied in three different study sites in Finland (Figure 1).

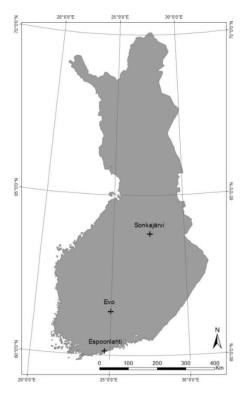


Figure 1. Locations of the study sites.

Karjalainen et al. (2011), Vastaranta et al. (2013a) and Vastaranta et al. (2014) have used study area locating in Espoonlahti, which has relative flat topography, mixed land uses and varying forest structures. The study area in Järnstedt et al. (2012), Vastaranta et al. (2013b) and Karila et al. (2015) has been Evo, which is dominated by managed coniferdominated forests. The study area in Sonkajärvi used by Nurminen et al. (2013) is dominated by managed pine forests.

ABA has been used for forest attribute estimation. The number of the used modeling plots has varied from 89 to 500, as the attribute estimation has been validated mainly at resolutions between 200 and 300 m^2 , corresponding to the field measured plots. Vastaranta et al. (2014) was the only study where results were validated at the stand level, as varying plot sizes were used in Nurminen et al. (2013).

In all of the studies, nonparametric estimation methods were used (Table 1). Random forest (RF) estimation was the mostused technique, and consequently, the results were also validated using RF's built-in cross-validation technique based on out-of-the-bag samples. Although most of the studies were independent, rather uniform methods have been used, which enables comparison of the used remote sensing data sets for forest attribute estimation.

	Study area	Number of plots	Prediction resolution	Estimation/ validation				
Airborne laser scanning								
Vastaranta et al. 2013a	Espoonlahti	110	200 m ²	RF, Out-of- the-bag				
Vastaranta et al. 2013b	Evo	500	300 m ²	RF, Out-of- the-bag				
Nurminen et al. 2013	Sonkajärvi	89	100–1257 m ²	RF, Out-of- the-bag				
Hyyppä et al. 2012	Evo	292		RF, Out-of- the-bag				

Yu et al. 2011	Evo	69		RF, Out-of- the-bag				
Aerial imagery								
Järnstedt et al. 2012	Evo	402	300 m ²	k-NN				
Nurminen et al. 2013	Sonkajärvi	89	100–1257 m ²	RF, Out-of- the-bag				
Vastaranta et al. 2013b	Evo	500	300 m ²	RF, Out-of- the-bag				
Radargrammetry								
Karjalainen et al. 2011	Espoonlahti	110	200 m ²	RF, Out-of- the-bag				
Vastaranta et al. 2013a	Espoonlahti	110	200 m ²	RF				
Vastaranta et al. 2014	Espoonlahti	207	Stand (mean 4.1 ha)	k-MSN				
Interferometry								
Karila et al. 2014	Evo	335	300 m ²	RF, Out-of- the-bag				

Table 1. Methods used in the estimation of forest inventory attributes.

4. COMPARISONS OF ALS AND ALTERNATIVE 3D TECHNIQUES

4.1 Data acquisition and processing

Image platforms are able to fly higher and faster than ALS platforms. Imaging instruments will typically have a field of view (FOV) of 75°; ALS FOVs (for forest applications) are $\leq 25^{\circ}$. Thus, for the same number of flying hours, image acquisition can cover a much larger area.

However, ALS systems have more flying hours per day, as they are insensitive to lightning conditions. Aerial imagery is strongly influenced by solar illumination and view angles (sun, surface and sensor geometry). Occlusions caused by shadows are particularly problematic for the generation of image-based point clouds in forest canopies.

In terms of flight planning, the advantage clearly lies with imagery, particularly in steep and/or complex terrain, primarily due to the higher altitude at which imagery can be acquired relative to ALS. The higher altitude associated with image acquisition, combined with the larger field of view of imaging instruments, also results in greater spatial coverage for imagery relative to ALS (given the same number of flying hours). One key advantage of ALS over imagery, however, is the flexibility associated with ALS acquisition. Imagery is strongly influenced by solar illumination and view angles, and occlusions or shadows in the forest canopy can greatly limit image matching capabilities. As a result, the number of hours available for image acquisition on any given day is limited, particularly at high latitudes like in Finland. This limitation is partially offset by the higher altitudes, faster flying speeds and greater field-of-view associated with imaging platforms. In contrast, ALS is not influenced by the presence of shadows or daylight and therefore affords more hours for acquisition. As an active sensing system, ALS data can theoretically be acquired at night, but in practice this is uncommon. In addition, although both image and ALS platforms are adversely affected by inclement weather (e.g., precipitation), as imaging platforms typically fly at higher altitudes, there is a greater opportunity for haze or cloud to negatively impact image acquisition. The time required to go from acquisition to point cloud is shorter for ALS, although this advantage has narrowed over time with the advent of fully digital photogrammetric workflows.

Nowadays, SAR data are available internationally from a large number of satellites with different frequency bands, polarizations and variable imaging geometries (see reviews in, e.g., LeToan et al. 1992, Ouchi 2013, Kaasalainen et al. 2015). Thus, in theory, the strength of SAR compared to aerial imagery or ALS is the easy data acquisition and also the temporal resolution. SAR data can be obtained on nearly a daily basis and also under varying imaging conditions. Userready products, such as DSMs or point clouds, can be easily processed with methods similar to those that are used with aerial imagery.

4.2 Point cloud or DSM quality

Compared to ALS and digital stereo imagery (DSI), the obtained point densities are far sparser. DSMs with resolutions ranging from 25 m² to 100 m² are usually derived using radargrammetry or interferometry (Figures 2 and 3). Even higher resolution is possible using the latest satellite data. DSMs based on ALS or digital stereo imagery are more detailed; resolutions vary between 0.25 m² to 1 m² (Figures 4 and 5). Another difference is that canopy heights can be measured accurately straight from the ALS or image point cloud or DSM, as due to radar signal penetration in the target, there is underestimation of the actual height (Praks et al. 2012).

Additional information compared to DSM can be obtained from aerial imagery or ALS point clouds. With regard to resolution, the image-based point cloud is capable of a greater point density than ALS data, for a given cost, as a function of the GSD and the number of independent three-dimensional pixel matches (Leberl et al. 2010). However, research into the impact that increasing ALS point density has on the accuracy of forest inventory attribute estimation would suggest that the increased point density afforded by image-based point clouds may not be of any particular advantage for the area-based approach (e.g., Treitz et al. 2012, Jakubowski et al. 2013).

However, the imagery-derived height information primarily characterizes the outer canopy envelope; the detection of small canopy openings is limited. The lack of penetration and insensitivity to small canopy openings limits the variety of metrics that may be generated from the digital stereo imagery when compared to the broad range of metrics that may be calculated from the ALS data (Vastaranta et al. 2013).

Radargrammetry may be capable of capturing slightly more variation in canopy height than interferometry. In the interferometric processing, pixels are averaged, and continuity of phase is favored, but radargrammetry is based on (independent) 3D measurements of tie-points between images, which may lead to an enhanced height variation response when multiple stereo-pairs are available with varying incidence angles and viewing directions.

4.3 Forest inventory attribute prediction accuracies

Based on Finnish experiences, 3D metrics derived from aerial stereo imagery can be used in the estimation of forest inventory attributes, and obtained accuracies are close to ALS in boreal forest conditions (Table 2). However, the imagery-derived height information primarily characterizes the outer canopy envelope; the detection of small canopy openings is limited. The lack of penetration and insensitivity to small canopy openings probably limits the prediction accuracy in multilayered stands. On the other hand, it should be noted that aerial images also provide spectral information that is useful in tree species classification, and the height information is as accurate as that obtained from ALS. A review of the potential of aerial image-derived point clouds for forestry purposes can be found in White et al. (2013). From 5 to 15 percentage points lower root mean squared errors are obtained using spaceborne radargrammetry and interferometry. Considering the resolution of the radargrammetric or interferometric DSM, the obtained results are surprisingly close to the results obtained with ALS or imagery. Based on the Finnish experiences, radargrammetry has provided slightly more accurate estimation of forest inventory attributes than interferometry, and that is probably due to its capability of capturing height variation, as the DSM generation is based on 3D measurement with multiple imaging geometries instead of averaging over multiple pixels, as in interferometry.

		Above-			Mean		
	Stem	ground	Basal-	Mean	diameter		
	volume	biomass	area	height			
Airborne laser scanning	RMSE-%						
Vastaranta et al. 2013a	23.4	24.7	***	***	***		
Vastaranta et al. 2013b	17.9	17.5	17.8	7.8	19.1		
Nurminen et al. 2013	20.7	***	***	6.6	11.4		
Hyyppä et al. 2012	20.3	***	***	6.1	16.1		
Yu et al. 2010	20.9	***	***	6.4	10.9		
Aerial imagery	RMSE-%						
Järnstedt et al. 2012	40.4	***	36.2	28.2	25.3		
Nurminen et al. 2013	22.6	***	***	6.8	12		
Vastaranta et al. 2013b	24.5	23.7	23.6	11.2	21.7		
SAR radargrammetry	RMSE-%						
Karjalainen et al. 2012	34		29	14	19.7		
Vastaranta et al. 2013a	29.9	30.2	***	***	***		
Vastaranta et al. 2014	16.3	16.1	12	6.7	***		
SAR interferometry	RMSE-%						
Karila et al. 2015	32	***	29	20	***		

Table 2. Obtained accuracies in forest attribute estimation using various 3D remote sensing methods in study sites around Finland.

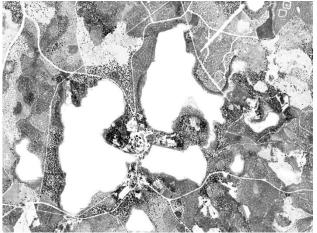


Figure 2. Canopy height model (1 m resolution) derived from airborne laser scanning data. Black areas indicate high vegetation. Three lakes (white) can be seen in the middle.

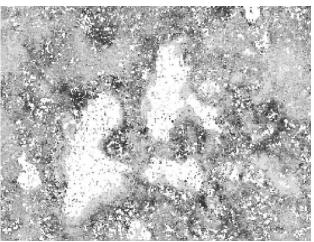


Figure 5. Canopy height model (10 m resolution) derived using radargrammetry.

5. CONCLUSIONS



Figure 3. Canopy height model (1 m resolution) derived from airborne stereo imagery.

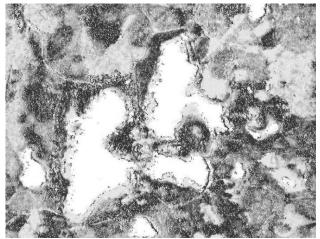


Figure 4. Canopy height model (10 m resolution) derived using interferometry.

We are able to summarize findings from several studies with similar reference measurements, estimation methods and validation techniques. In addition, most of the studies were implemented in the same study area, enabling fair comparisons.

In general, it would appear that for even-aged, single canopy layer stands, such as those that dominate the study areas, prediction accuracies are surprisingly close to each other, considering the differences in level of detail. ALS data sets used in Vastaranta et al. (2013) included 10 pulses per m^2 , enabling production of 10–40 points per m^2 , characterizing ground and vegetation. As in Karila et al. (2015), interferometric DSM had a resolution of 4x4 m, meaning there were only around 20 height observations per plot. This also indicates that the main predicative power comes from 3D, i.e., the inclusion of the tree height information, and it does not matter so much what is the final point density. Therefore, the results are also comparable to Hyyppä and Hyyppä (1999) and Hyyppä et al. (2000).

It should be kept in mind that in all of the studies, a rather large number of plots were used, and the predictions were validated using the same plots, as the models were calibrated. Although one of the advantages of the RF is that it is not overfitting and it can provide validation statistics using built-in cross-validation, the obtained accuracies are probably on the positive side. However, the same methods were used in most of the studies, and thus, comparison between the remote sensing materials and prediction accuracies should be realistic. Still, we assume that if a smaller amount of plots per hectare is used, as is the case in operational accuracies, the accuracy of radarbased predictions will decrease more than ALS or image-based predictions. The reason for this is that ALS and image-based point clouds are capturing the height and density variation in more detail, even enabling direct measurements for vegetation height. Thus, those can provide better physical correspondence with point clouds and the actual attributes of interest.

3D data enable higher prediction accuracy of the forest inventory attributes when compared to 2D data (optical satellite imagery, radar intensity). With these more costefficient 3D techniques, forest managers could potentially realize more frequent inventory cycles, providing more up-todate information. For the purposes of forest mapping and monitoring, it has been suggested in Finland that ALS data could be acquired at regular, but extended, time periods (i.e., every 10 or 20 years, depending on forest and management considerations), with forest information updated periodically using alternative point clouds (Vastaranta et al. 2013, Holopainen et al. 2014) derived from optical satellite, aerial or radar imagery acquired in between regular forest inventory cycles. In theory, these point clouds could be used for areabased estimations in a fashion analogous to ALS data, as also proved here in many studies where forest inventory attributes have been predicted using ALS, imagery or SAR data. It should be pointed out that all of these alternative techniques require ALS-based DTM/DEM to obtain accurate estimates for forest inventory attributes.

We remain circumspect on recommendations regarding the use of alternative 3D techniques in other forest types, particularly in mixed-aged, multi-layered stands, and other management regimes. The relatively homogeneous and well-managed stand conditions found in the study areas result in limited variance and facilitated strong relationships between DSMs and the ground plot measurements.

Looking forward, it certainly appears that there are opportunities especially for DSI in forest management and planning. The capacity of DSI to achieve estimates with accuracies close to ALS needs to be tested in a broader range of forest conditions, particularly in larger, more complex stands (White et al. 2013).

Opportunities for monitoring and inventory update applications are foreseen for jurisdictions where large-area ALS coverages have been acquired, imagery or other 3D data is routinely collected, and numerous well-distributed ground plots are present. We may assume that in most of the cases even SAR could be used for forest attribute updating (time interval T1-T2). Mapping in T1 is done with ALS to obtain DTM and forest attributes, but in T2, SAR-derived DSM could be used. In forest inventory updating, SAR competes with repeated ALS data acquisition, digital aerial stereo imagery derived DSM and DSMs derived from optical satellite images. All these methods can provide prediction accuracies close to ALS and improvements to the current large-area mapping methods (e.g., methods where field plot data is generalized over large areas using Landsat imagery).

SAR-based DSMs are less detailed than ALS DSMs, but they could be more cost effective when large areas need to be covered. The advantage of SAR DSMs over those from optical imagery is in their availability regardless of weather and time of day. In addition to forest attribute updates, there is growing interest toward natural hazard monitoring systems. Due to good temporal resolution, SAR techniques have an operational potential to monitor forest changes such as wind or snow damages. However, for monitoring of non-stand replacing disturbances, such as snow damage and other individual treelevel changes, further research is needed with SAR.

From the user's point of view, which one of the mentioned 3D remote sensing methods is selected to be used in forest inventory attribute updates depends on data availability,

acquisition and preprocessing costs. User-ready products, such as DSMs or point clouds, can be easily processed with available software (imagery and SAR) and used for area-based estimation of forest inventory attributes.

ACKNOWLEDGEMENTS

The Academy of Finland is acknowledged for its financial support in the form of the projects "Science and Technology Towards Precision Forestry" and "Centre of Excellence in Laser Scanning Research (CoE-LaSR)." EU is acknowledged for its financial support in the form of the ESA Advanced SAR project.

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