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Comparison of Forest Attributes Derived from Two Terrestrial Lidar Systems

Mark J. Ducey, Rasmus Astrup, Stefan Seifert, Hans Pretzsch,
Bruce C. Larson, and K. David Coates

Abstract

Terrestrial lidar (TLS) is an emerging technology for deriving forest attributes, including conventional inventory and canopy characterizations. However, little is known about the influence of scanner specifications on derived forest parameters. We compared two TLS systems at two sites in British Columbia. Common scanning benchmarks and identical algorithms were used to obtain estimates of tree diameter, position, and canopy characteristics. Visualization of range images and point clouds showed clear differences, even though both scanners were relatively high-resolution instruments. These translated into quantifiable differences in impulse penetration, characterization of stems and crowns far from the scan location, and gap fraction. Differences between scanners in estimates of effective plant area index were greater than differences between sites. Both scanners provided a detailed digital model of forest structure, and gross structural characterizations (including crown dimensions and position) were relatively robust; but comparison of canopy density metrics may require consideration of scanner attributes.

Introduction

In the past decade, terrestrial lidar systems (TLS) have shown promise as tools for estimating standard forest inventory variables (e.g., Aschoff *et al.*, 2004; Henning and Radtke, 2006b; Maas *et al.*, 2008; Thies and Spiecker, 2004; Watt and Donoghue, 2005), and forest canopy structure and attributes (e.g., Clawges *et al.*, 2007; Henning and Radtke, 2006a; Hosoi and Omasa, 2009; Huang and Pretzsch, 2010; Moorthy *et al.*, 2008; Vierling *et al.* 2008; Zheng and Moskal, 2009). For standard forest inventories, research efforts have focused

on automatic detection of individual trees (e.g., Brolly and Kiraly, 2009; Maas *et al.*, 2008; Simonse *et al.*, 2003) and the subsequent estimation of parameters such as tree diameters and height (e.g., Aschoff and Spiecker, 2004; Henning and Radtke, 2006b; Hopkinson *et al.*, 2004; Maas *et al.*, 2008; Strahler *et al.*, 2008) and taper or other stem quality attributes (Henning and Radtke, 2006b; Lefsky and McHale 2008; Thies *et al.*, 2004). With regard to crown structure, research efforts have focused on measures such as gap fraction, plant area index, and leaf area index (LAI) at the plot level (Henning and Radtke, 2006a; Hosoi and Omasa, 2009; Strahler *et al.*, 2008; Zheng and Moskal, 2009), or individual tree level (e.g., Clawges *et al.*, 2007; Huang and Pretzsch, 2010; Moorthy *et al.*, 2008), or branch and leaf level (e.g., Eitel *et al.*, 2010; Hosoi *et al.*, 2011). Dassot *et al.* (2011) provide a recent review of past research and potential applications.

The effect of scanner setup pattern (single scan mode versus multiple scan mode) (Maas *et al.*, 2008; Van der Zande *et al.*, 2008) and scanner position in the canopy (Hosoi and Omasa, 2007; Van der Zande *et al.*, 2006) on the retrieved inventory and canopy structure parameters has received some attention in the literature. By contrast, the effect of the scanner systems themselves on derived forest attributes has received only minor attention. The only published comparison of scanner systems for forestry applications used different scanners on different plots, illustrating differences in performance but not allowing direct comparison (Maas *et al.*, 2008). Today, a wide variety of commercial TLS systems exist, and the systems applied in forest-related applications range from 2D systems with low resolution (e.g., Rossel *et al.*, 2009; Van der Zande *et al.*, 2006) to 3D systems with much higher resolution (e.g., Eitel *et al.*, 2010; Henning and Radtke, 2006a,b; Hosoi *et al.*, 2011). From detailed studies of canopy structure with hemispherical photography, it is apparent that factors such as resolution may be important for the derived canopy parameters (Englund *et al.*, 2000; Hale and Edwards, 2002; Jonckheere *et al.*, 2004). Differences in pulse penetration have been demonstrated between TLS and airborne lidar (Chasmer *et al.*, 2006; Hilker *et al.*, 2010 and 2012). For TLS applications, it has been suggested that scanner system specifications might be important for derived forest attributes (Clawges *et al.*, 2007; Zheng and Moskal, 2009), but the possible effects have not been formally investigated through field trials. Understanding the impact of scanner specifications on derived attributes is especially important for future deployment in successive surveys designed to estimate changes

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in crown and stem characteristics, as scanner technology is changing quite rapidly whereas forest growth is a relatively slow process.

The objective of our study was to investigate the effect of two TLS systems on selected forest inventory and canopy structure variables. We met our objective by (a) scanning two different forest stand types with two different TLS systems, (b) comparing the stem mapping and tree diameter distribution derived from the two systems, and (c) comparing the canopy attributes, derived gap fractions, and LAI from the two stands.

Methods

Study Sites

Field data for the study were collected in two forest stands (Table 1) around Smithers in northwestern British Columbia, Canada (54°35'N, 126°55'W). The study area has a continental climate with cold temperatures and snowy winters and is part of the Sub-Boreal Spruce, moist cold subzone (SBSMC2; Banner *et al.*, 1993; Meidinger and Pojar, 1991). Lodgepole pine (*Pinus contorta* var. *latifolia* Engelm.) dominated stands are most abundant on the poorer and drier soils while interior spruce (a complex of white spruce *Picea glauca* (Moench) Voss and Engelmann spruce (*P. engelmannii* Parry ex Engelm.)), and subalpine fir (*Abies lasiocarpa* (Hook.) Nutt.) normally dominate mature forests on moisture-receiving sites. Sites intermediate in soil moisture and nutrient availability often support mixed forests of all three conifer species.

We selected two mature forest stands as a study sites (Table 1) representing a contrast in terms of soil nutrient availability and associated species composition and density. Site 1 is a nutrient-poor site dominated by pine and with a relatively open canopy, and Site 2 is a nutrient-rich site dominated by subalpine fir and spruce with a denser, more vertically-complex canopy. At both sites, trees were stem-mapped and measured for their diameter at breast height (DBH) in the immediate vicinity of the central scan point.

Laser Scanning

We used a Riegl LMS-Z360i and a Leica ScanStation to carry out scans at multiple positions on both study sites. The two scanners differ in key characteristics (Table 2), which potentially can affect the estimation of key canopy metrics. For each study site, one scan position was located in the plot center, and the remaining scan positions were located around it. The edge scan positions were located so that 10 to 20 trees were completely scanned from all sides and the crown shape and height of the trees immediately adjacent to the plot center could be seen from multiple scan points. Both scanners were deployed over identical scan points. However, due to differences in tripods and other aspects of scanner ergonomics, the

TABLE 2. COMPARISON OF KEY SCANNER SPECIFICATIONS

Scanner	Riegl LMS-Z360i	Leica ScanStation
Scanner type	Time of flight	Time of flight
Return	Last return/First return*	First return
Laser class	1	3R
Color (wavelength)	1550 nm	532 nm
Resolution (in this study)	0.12 degrees	0.20 degrees
Beam divergence	2 mrad**	0.2 mrad
Beam footprint at 10 m	20 mm	6 mm

*Site 2 was scanned twice with the Riegl LMS-Z360i in order to obtain datasets with last and first return.

**The scanner incorporates an integrated beam focusing system. In our study this was set at infinity.

height above the scan point benchmarks differed between scanners.

In the case of the Riegl LMS-Z360i, the angular distance between two distance measurements was set to 0.12 degree, which corresponds to a point spacing of approximately 2 cm if the target is 10 m from the scanner. The Riegl LMS-Z360i scanner has a horizontal scan capability of 360 degrees and a vertical field of view of 85 degrees. To provide coverage of the total canopy, the scanner was tilted, and a second scan was made from each scan position to obtain the upper tree sections. Before scanning, retro-reflective targets were positioned around the scanning area. Each target was scanned with the highest resolution of the scanner from each scan position. As a postprocessing step, the co-registration of the scan points was made by automatically matching the centroids of the retro-reflective targets. The central scan was always oriented to magnetic north by using a compass. To obtain a vertically oriented local coordinate system, a bubble level was used to ensure that the scanner was positioned upright. At Site 1, the Riegl was used only in last return mode (i.e., the farthest return distance associated with an individual pulse was converted to an x, y, z coordinate); at Site 2, the Riegl was used in both first (i.e., nearest) and last return mode to compare performance.

In the case of the Leica ScanStation, the angular distance between the two distance measurements was set to 0.20 degree, which corresponds to a point spacing of approximately 3.5 cm if the target is 10 m from the scanner. The Leica has a horizontal scan capability of 360 degrees and a vertical field of view of 135 degrees (from 45 degrees below the horizontal, up to the zenith), and hence only one scan was needed at each scan point. Co-registration of scan points was achieved by conducting a closed traverse, using a circular target at the foresight and backsight position for each scan point, and using the built-in dual-axis tilt compensator of the scanner to ensure vertical accuracy.

Analysis

Postprocessing

To ensure differences in key results reflected scanner differences rather than postprocessing differences, we employed identical postprocessing procedures for both scanners, with each step performed by members of a single research team using data from both scanners, with the exception of co-registration of points within a site, which was done automatically within the Leica software. First, all data were exported to ASCII files. Next, all data for each scanner at each site were combined into single files that included the (x, y, z) coordinates of all contacts, the identity and location of the scanner

TABLE 1. STUDY SITE CHARACTERISTICS

Site	Site 1	Site 2
Site type	Nutrient poor	Nutrient rich
Basal area (m ² /ha)	42	59
Stems/ha	1517	1644
Species composition (pine, spruce, fir)	51%, 27%, 22%	9%, 28%, 63%
Age range (breast height) (years)	9–71	92–182
Median age (years)	57	150
Top height	25 m	35 m
No. of scans	5	4

position associated with the contacts, and coordinate information for each probe that did not return a contact. The files were translated into a common coordinate system for both scanners, and served as the basis for all further analyses.

Visualization

We used visual comparison of a series of images derived from scanner data as an initial basis for comparison to facilitate better understanding of the quantitative differences that might be uncovered later. Visualization is inherently subjective, but the ability of the human visual system to perceive patterns in optical data remains unsurpassed by quantitative and computational techniques, despite decades of research (Neumann *et al.*, 2007; Todd, 2004). Visualization of scanner data thus represents an avenue toward understanding what quantitative algorithms might be able to do with such data, and discerning what patterns in the data might be important for current algorithms.

As a first step in scanner comparison, we created range images from the data from each scanner at each scanning position. Contacts occurring at a range of >25 m from each scan point were not included in the range image. Range images were constructed by rasterizing the data with 1,600 pixel resolution in the 360 degree horizontal range, using the closest contact in each pixel. Visual inspection of matching images provided an immediate and intuitive basis for understanding the differences in quantitative metrics calculated from the data.

As a further aid to visual interpretation, we extracted matching slices through the point clouds of the combined scan data for each scanner at each site. These included the following: a vertical slice 2 m wide that passed through the plot center, which was useful for understanding horizontal and vertical penetration and subsequent reconstruction of three-dimensional stand structure; a horizontal slice 0.1 m thick, centered at breast height (1.3 m) above ground surface at the plot center, which was useful for understanding differences in reconstructed forest inventory attributes; and a horizontal slice 2 m thick, centered at mid-canopy height (10 m) above ground surface at the plot center, which was useful for understanding likely differences in reconstructions of crown size, shape, and position.

Scanner Penetration

For each scanner, we quantified penetration through the canopy horizontally and vertically. We used the cumulative fraction of contacts within a given distance from each scanner position (including only those contacts originating from that position, and excluding probes that did not yield a return) as a measure of penetration. Penetration was quantified in terms of horizontal and vertical distances because we hypothesized that the anisotropic structure of the forest canopy might lead to differences in scanner performance in each dimension.

Stem Mapping

To ensure an objective comparison between the scanners and the modes of measurement, we implemented an identical algorithm to extract stem positions and stem diameters. The algorithm is based on the tendency for stems to completely block ground-level returns that are behind them from the vantage point of a single scan position. By contrast, low-lying twigs and foliage only partially occlude the ground behind them. Moreover, as branches are normally not connected to the ground, this will lead to measured ground points farther away from the scanner than the branch. Thus, a narrow void within a single scan within which no ground-level returns occur is likely to contain a tree stem at the apex of the void pointing toward the scan position.

Once a void is detected, indicating a possible stem location, the algorithm includes two further steps. First, the points above the ground near the apex of the void are clustered based on their distance. Clusters with enough points are used as potential stem positions. Starting with these potential stem positions, a second step is performed to extract a diameter, if possible. For the stem diameter estimation, not only one scan from one position is used, but also all available scans merged to one point cloud. A subset of points around each potential stem position is selected. These points are grouped to layered grids. Only points that are in grid cells which have corresponding occupied cells in at least four different layers are selected, indicating a vertically-oriented object. With this layering, horizontal structures are removed, as they occupy only one or two layers locally. For example, such objects could include a whorl of horizontally-oriented branches surrounding a vertically-oriented stem. Next, a distance-based clustering is applied with an ellipsoidal distance function. This is done because stems have higher vertical spread than horizontal spread; the ellipsoidal distance function leads preferentially to the vertical rather than horizontal expansion of the cluster. Finally, for each cluster, a circle is fitted and the best fitting circle is used as a stem diameter.

Canopy Attributes

For each scanner position, we calculated canopy gap fraction as the simple proportion of probes that failed to return a contact within each of 16 equally spaced inclination angle classes. The resulting canopy gap fractions were then used as input into the leaf area inversion model described by Norman and Campbell (1989) using the same 16 leaf inclination angle classes for estimation. Although more sophisticated approaches are available with TLS data, as with hemispherical photography (e.g., estimating canopy gap fraction in multiple azimuthal regions (cf. Lang and Yuequin, 1986); estimation of leaf area in voxelized subspaces (Henning and Radtke, 2006a; Hosoi and Omasa, 2009)), we focused on this relatively simple procedure as one that would highlight any differences due to scanner attributes without miring the analysis in algorithmic complexity. The procedure yielded estimates of effective plant area index (PAI_{eff} ; cf. Chen *et al.*, 1997) and mean tip angle (MTA) for each scanner location, for each scanner. These estimates were tested for statistically significant differences using mixed-effects ANOVA, with scanner (Leica, Riegl last return, or Riegl first return) and scanner by site interaction as fixed effects, and with site and position within site as random effects. These models were compared to null models that only included the random effects. Models were fit by using restricted maximum likelihood, even though ANOVA statistics estimated using this method are approximate; had any tests been ambiguous, they would have been repeated using bootstrapped likelihood ratio statistics. Statistical analysis was conducted using the NLME library (Pinheiro *et al.*, 2009) in the R statistical package (R Development Core Team, 2009).

Results

Visualization

Range images for the two scanners at the plot centers of both sites are shown in Plate 1. Some general attributes of the scans can be discerned from these images, and are consistent across the range images from all scan points (not shown). First, as might be anticipated, the finer beam of the Leica scanner results in a greater level of apparent detail, even though the angular resolution of the scan is lower. For example, the Leica is better able to resolve both dead and foliated lower branches of nearby trees than the Riegl in either mode. Such features appear thicker and “padded” in the Riegl scans, in

Plate 1. Range images (in spherical coordinates) from the plot center for both scanners: (a) Site 1, Leica, (b) Site 1, Riegl - last return, (c) Site 2, Leica, (d) Site 2, Riegl - first return, and (e) Site 2, Riegl - last return. Red pixels are closest to the scanner with cooler colors indicating greater distance.

comparison with the Leica. Visually, the range maps from the Riegl used in last-return mode are more similar to those of the Leica than are those from first-return scans, but these differences are qualitative and hence difficult to quantify and are open to subjective interpretation. It is also apparent that the Riegl first-return and last-return scans can differ substantially

in their rendering of semi-transparent nearby objects, such as foliated branches and small trees, as well as fine-textured objects such as dead branches. Such objects are much more opaque and heavily padded in first-return scans, but can also be almost transparent in last-return scans. The range images also show the difference in intensity of scan patterns near

the zenith: the Leica provides a very high intensity pattern (a consistent separation in spherical coordinates θ and ϕ , implying adjacent probes are actually closer together as elevation increases toward the zenith), whereas the Riegl maintains a more uniform separation (which graphs as a more dispersed pattern in spherical coordinates).

The qualitative characteristics seen in the range images carry forward into attributes of the point clouds. Figure 1 shows vertical slices through the point clouds. In these images, the finer-beamed Leica scanner is able to resolve finer detail than the Riegl, but the Riegl is better suited for describing the tops of all trees, especially those in Site 2, when it was used in last-return mode. However, the images share many

similarities. In general, the same trees are recognizable with the same shapes and in the same positions. This implies that competent, flexible algorithms should be able to discriminate and measure gross structural attributes (e.g., number, size, and spatial distribution of tree crowns) in similar ways using either scanner.

Figures 2 and 3 show horizontal slices through the point clouds, centered at breast height above the plot center. Here, the differences are more striking. The fine-beamed Leica scans clearly show individual tree trunks, with many stems represented as hollow circles surrounded by a narrow ring of returns. This is also the case for trees that have branches and foliage surrounding the stem (see, e.g., Figure 3).

Figure 2. Horizontal slices through the point cloud, 0.1 m thick, centered 1.3 m above ground surface at the plot center; axes show coordinates in meters, with tripod locations as shaded triangles: (a) Site 1, Leica, (b) Site 1, Riegl - last return, (c) Site 2, Leica, (d) Site 2, Riegl - last return, and (e) Site 2, Riegl - first return.

Corresponding tree stems in the coarser-beamed Riegl scans remain obvious, although they are not as simply rendered. Where the scan interacts with foliage and branches at breast height, there can be a characteristic smearing of these

structures radially away from the scanner. These artifacts are known as intermediate points, and result from multiple reflections within a single impulse. If the reflecting surfaces in one laser impulse have smaller distance differences than the

Figure 3. Close-up of the horizontal slice at 1.3 m above plot center, showing the differences in rendering of individual tree stems within the point cloud, with tripod locations as shaded triangles: (a) Site 1, Leica, (b) Site 1, Riegl - last return, (c) Site 2, Leica, (d) Site 2, Riegl - last return, and (e) Site 2, Riegl - first return.

length of the laser impulse, it will not be possible to separate them and an average distance will be measured. This could lead to points appearing in the point cloud that fall between the positions of the physical objects in the scan. These features imply that algorithms for identifying and mapping individual tree stems might work differently with data from the

two scanners, and careful cross-calibration might be needed in an inventory context.

Finally, slices through the point clouds at mid-canopy position (Figure 4) show reasonable qualitative agreement between scanners. The Leica continues to show a greater degree of fine detail, such as the characteristic star-shaped

Figure 4. Horizontal slices through the point cloud, 2 m thick, centered 10 m above ground surface at the plot center; axes show coordinates in meters, with tripod locations as shaded triangles: (a) Site 1, Leica, (b) Site 1, Riegl - last return, (c) Site 2, Leica, (d) Site 2, Riegl - last return, and (e) Site 2, Riegl - first return.

branching pattern of some lodgepole pine crowns, while the Riegl does not show such detail as readily. The positions and sizes of crowns recorded by the two scanners are generally consistent, although the Riegl shows the same slight padding

of branches and other fine features as in other visualizations, leading to a slight loss of negative space between crowns. Otherwise, for gross structural characterization of the canopy, the two systems appear to give similar results.

Scanner Penetration

For horizontal penetration, both the Leica and the Riegl used in last-return mode showed consistent differences (Figure 5). For example, at Site 1 the median horizontal distance to return for the Leica was 2.81 m, whereas for the Riegl it was 4.33 m. At Site 2, the corresponding distances were 3.55 m and 8.92 m, respectively. However, the Riegl used in first-return mode had similar penetration to the Leica, with a median horizontal distance to return of 3.94 m.

Vertical penetration showed a different pattern than horizontal penetration (Figure 5). The Leica and the Riegl in last-return mode had similar vertical penetration (3.02 m as opposed to 3.15 m median vertical distance to return at Site 1, and 4.15 m as opposed to 4.03 m at Site 2). However, the Riegl in first-return mode showed much less vertical penetration (2.34 m at Site 2).

Stem Mapping

The results of stem position detection in both stands are given in Table 3. Both site conditions and scanner attributes appear to have impacted the number of detected stems, including the

propensity for false positives and false negatives. At Site 1, more potential tree positions were found with the Leica scanner, and also more trees could be matched with manually extracted stem positions using the Leica. At Site 2, the Leica again performed better than the Riegl in first-return mode, but the Riegl in last-return mode performed best. Site 2 had more dead and living branches occluding the stems, leading to false positives. The broader beam of the Riegl scanner may have contributed to false positives due to the intermediate points between the real stems. However, the false positives also included the poles to hold the reflectors or registration targets.

Using the potential stem positions as a starting point, stem diameters for Site 2 were extracted. The results are summarized in Figure 6. There was a tendency for both Riegl scanner modes to slightly underestimate stem diameters, but the Leica showed greater variability. The Riegl in last-return mode showed the best performance, as judged by mean absolute deviation (MAD). However, none of the scanners or modes provided sufficiently reliable diameter measurements using the algorithm employed here to substitute for traditional measurements using a tape or calipers.

TABLE 3. RESULTS OF STEM MAPPING

Site 1: 71 manually extracted trees			
Scanner	Potential Positions	Correct Matches	Trees Not Found
Leica	80	29	42
Riegl (first return)	68	21	50
Site 2: 45 manually extracted trees			
Scanner	Potential Positions	Correct Matches	Trees Not Found
Leica	73	25	20
Riegl (first return)	96	29	16
Riegl (last return)	72	30	15

Figure 6. Comparison of diameter extraction at Site 2 for trees that were detected by all three scanner/mode combinations: (a) Leica, (b) Riegl - first return, and (c) Riegl - last return.

Canopy Attributes

Gap fraction as a function of zenith angle is shown for the two scanners in Figure 7. At Site 1, the Leica had a higher mean gap fraction for angles near the zenith than the Riegl used in last-return mode, but a lower mean gap fraction for angles greater than 60 degrees from zenith (i.e., less than approximately 30 degrees from horizontal). At Site 2, the pattern was similar. Here, the Leica had a higher mean gap fraction than the Riegl used in either mode for angles less than 50 degrees from zenith, but a lower gap fraction for angles more than 50 degrees from zenith.

The impact of different estimates of gap fraction on effective plant area index and mean tip angle are shown in Table 4. The Leica provided consistently higher estimates of plant area index than the Riegl. These differences were highly significant (likelihood ratio chi-square 21.73, $p < 0.0001$). Differences in mean tip angle, although marginally significant from a statistical perspective (likelihood ratio chi-square 6.75, $p = 0.034$), are so small (only a fraction of a degree between scanners and sites) as to be practically inconsequential. Both the binning of zenith angles during analysis, and the approximate nature of the statistical hypothesis test, can be considered as arguments against any attempt to interpret such trivial differences in mean tip angle.

Discussion

Although both of the scanners employed in our study are relatively high-resolution scanners, capable of acquiring point

clouds with millions of points in a single hemispheric scan, the differences upon visual inspection of range maps as well as scatter plots of slices through point clouds are striking and informative. The scanners differ in numerous design attributes, including laser wavelength and power, beam divergence, and data pre-processing. Because the latter occurs within the scanner unit before export and often is proprietary information, it is difficult to evaluate its full impact. However, at least some of most obvious visual differences are attributable to beam diameter, which depends on diameter of the beam exiting the instrument and on beam divergence. The comparison shows that a small beam diameter leads to better penetration through low branches and understory vegetation (Figure 3) potentially giving small beam diameter scanners a greater potential for deriving stem properties. Significant divergence in beam diameter exists between scanners employed in forestry applications, and the presented results show that beam diameter is an important parameter to consider when choosing an instrument and interpreting derived variables. Even though there is significant divergence in beam diameter between the two compared scanners (Table 2), both scanners have narrower beams at operational distances of 10 to 20 m than for example the ECHIDNA instrument used by Jupp *et al.* (2005) and Strahler *et al.* (2008).

The ability of a scanner to penetrate a forest stand will be a critical determinant of multiple aspects of performance. Greater penetration should, in principle, allow recovery of stand structure from fewer scan positions or from positions placed farther apart. All other things being equal, a scanner

Figure 7. Canopy gap fraction as a function of angle of inclination at each scan position (dashed lines) and average overall positions (solid line) for both scanners at both sites: (a) Site 1, Leica, (b) Site 1, Riegl - last return, (c) Site 2, Leica, (d) Site 2, Riegl - last return, and (e) Site 2, Riegl - first return.

with greater penetration should be able to map and identify individual trees at a greater distance from the scan position. Greater penetration should also allow better reconstruction of the attributes of upper canopies, although if such penetration occurs because returns are not being provided for lower canopy elements, there may be tradeoffs in resolving different canopy layers. We speculate that the observed differences in

patterns between horizontal and vertical penetration (Figure 5) are caused by differences in the geometric configuration of canopy elements and their interaction with scanner properties. For vertical penetration, the horizontal sprays of conifer foliage and fine branches, which are relatively efficient for capturing sunlight, may also be relatively opaque to the Riegl when used in first-return mode, but can be penetrated by the

TABLE 4. ESTIMATES OF EFFECTIVE PLANT AREA INDEX (PAI_{eff}) AND MEAN TIP ANGLE (MTA) FOR EACH SCANNER AND SITE

Scanner		Site 1		Site 2	
		Mean	Std. Dev.	Mean	Std. Dev.
Leica	PAI_{eff}	2.08	0.09	2.28	0.11
	MTA	23.5	2.6	25.1	0.3
Riegl (Last Return)	PAI_{eff}	1.31	0.15	1.68	0.24
	MTA	23.4	0.4	25.6	0.2
Riegl (First Return)	PAI_{eff}			1.74	0.43
	MTA			25.6	0.1

fine beam of the Leica or, with reasonable frequency, by a sufficient signal to allow a more distant detection when the Riegl is used in last-return mode.

Given the employed algorithms, which were developed specifically for the comparison task, neither of the scanners performed particularly well in the stem mapping and diameter extraction tasks. We note that the algorithm was not specifically optimized for any of the scanner setups, as doing so would have compromised the comparison. The stem mapping and diameter extraction results are consequently not an indication of the potential for TLS for these operations and significantly better performance has been shown in numerous studies (e.g., Brolly and Kiraly, 2009; Henning and Radtke, 2006b; van Leeuwen *et al.*, 2011; but see Thies and Spiecker (2004) for a study with similarly low detection rates). Our study does not include a formal comparison between algorithms, so we do not advocate for or against our algorithm here. However, the differences in performance between scanners suggest that point cloud attributes that depend in turn on scanner characteristics do matter for stem mapping and related inventory tasks. Maas *et al.* (2008) also suggested that beam divergence might impact the accuracy of tree diameter recovery. Hence, scanner-specific algorithm optimization is likely to be necessary as long as algorithms are not specifically designed to account for specific scanner properties and their interactions with forest canopies. The poor stem mapping and diameter extraction performance can partly be attributed to lack of scanner optimization, but also to the selected stand conditions with a high degree of low branching (Figures 1 and 2), a condition that has been underrepresented in field tests of TLS and inventory attributes.

Differences between effective PAI estimates for the two scanners were significantly different, and were larger than the estimated differences between the two sites, thus highlighting the need to account for scanner specifications in interpretation of results. The similarity in estimated plant area index between the two sites was unexpected given the striking differences in soil resources. However, given that the figures reported are effective PAI (Chen *et al.*, 1997), actual differences may have been masked by differences in shoot-level clumping between pine (which dominated the canopy of Site 1, which had poorer soils) and spruce and fir (which are dominant at Site 2, which had richer soils). The greater clumping of spruce and fir needles should have led to a more substantial underestimate of LAI and PAI at that site.

A potential advantage of TLS over hemispherical photography is that in principle it is unnecessary to set threshold values to calculate gap fraction, as must be done for photographs (Jonckheere *et al.*, 2005). Thus, gap fraction estimates and derived quantities such as PAI should be more consistent between observers, and should depend less on sky conditions. However, the results obtained in our study suggest that scanner attributes, including beam size and maximum range, may be important for interpreting gap fractions and values calculated from them.

It had been previously hypothesized, but not demonstrated in a controlled study, that details of the scanner might be important for derived forest attributes (Clawges *et al.*, 2007; Zheng and Moskal, 2009). Conversely, fine-beam TLS systems are capable of imaging with increasingly high resolution and spatial accuracy (Dassot *et al.*, 2011). Thus, it might be tempting to believe that while details of the point clouds from different scanners might vary, the results in terms of derived characteristics would be quite small, especially in comparison to the substantial differences sometimes observed between terrestrial and airborne scanning (e.g., Chasmer *et al.*, 2006; Hilker *et al.*, 2010 and 2012). Moreover, the emerging uses of this technology for ecological investigations (see, e.g., Vierling *et al.*, 2008; Dassot *et al.*, 2011; van Leeuwen *et al.*, 2011) and declines in scanner cost may encourage new users to explore TLS who are not dedicated lidar researchers and who might, therefore, not find the consequences of differing TLS configurations intuitive. Our results should serve as a caution against naive comparison of results derived from data from different TLS scanners.

In conclusion, we share the optimism of several authors (e.g., Henning and Radtke, 2006a; Strahler *et al.*, 2008; Dassot *et al.*, 2011; van Leeuwen *et al.*, 2011) that terrestrial laser scanning will be an important tool for describing forest attributes in the future, but highlight that scanner attributes such as beam size, wave length, and maximum range may have strong implication for the derived results. In this, we echo the conclusions of Chasmer *et al.* (2006) who emphasize the need to understand how specific laser scanners (whether airborne or terrestrial) interact with vegetated canopies in order to improve the reliability and consistency of scanner-derived forest attributes.

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