# Comparison of Multi-Platform LiDAR Sensors for Efficient Forest Inventory

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Forest Growth & Yield Lab



MOSCOW, IDAHO

### **Motivation and Objectives**

### 📌 Gap

Lack of rigorous, multi-sensor (TLS–MLS–DLS) comparisons across structurally diverse forest conditions.

### Solution

Conduct a systematic evaluation of platform-specific performance in both simple and complex forest structures.

### 🌍 Objectives

- 1. Assess sensor accuracy for tree count, DBH, and height.
- 2. Compare data collection and processing time under varying structural complexity.



### **Study Sites**



#### MPB-PSP Network

- Long-term PSP establishment since 1960, old-growth forest,
- 5–10-year remeasurement,
- Nested plot design (Tree, regen, sapling),
- PSP-9 (4 subplots) were selected, and plot sizes are 1012 m<sup>2</sup>.



#### Vanderwell (VAN) CTs

- Private land owned,
- White spruce plantation (former farmland), 30 years old,
- 500 m<sup>2</sup> and nine rectangular plots,
- Randomized complete block with control, 33%, 50% RD.



#### Blue Ridge Lumber (BRL) CTs

- Blue Ridge, Alberta (West Fraser Division),
- Naturally regenerated lodgepole pine stand, 50 years old,
- 500 m<sup>2</sup> and nine rectangular plots,
- Randomized complete block with

control, 33%, 50% RD.



### **Materials and Surveying Procedure**

#### **Surveying Procedure**





DJI M300 RTK - R2A lidar Collected in 2024

x5

Point Rate: 200,000 points/s

**Drone LS DATA** 

Flight parameters:

80/80 m overlaps, 4 m/s flight speed at 80 m altitude

> Point Density: ~900 points/m<sup>2</sup>

#### **Mobile LS DATA** GeoSLAM's Zeb Horizon - SLAM

Collected in 2024

Scanning rate: 300,000 points/s.

Range: 100 m.,

Operation: Same path was scanned twice

points/m<sup>2</sup>

Point Density: ~4,500



#### **Terrestrial LS DATA**

Leica RTC360

Collected in 2024

Scanning rate: 2 million points/s.

Range: 120 m.

**Operation:** Double-scan, high density

×10 → Point Density: ~45,000 points/m<sup>2</sup>



### Method: General Workflow





### Method: Drone (D)-based Lidar Sensor

- 1. Data collection using DLS,
- 2. Pre-processing (Noise filter, ground classification, height normalization)
- 3. Detect tree locations based local maxima filtervariable window,
- 4. Feature extraction
  - a.  $\text{Height}_{\text{tree}} = Z_{\text{max}} Z_{\text{ground}}$ ,
  - b. Tree matching procedure (for species)
  - c. DBH was estimated by Alberta-specific H-DBH models (Huang, 2016)
- 5. Validation and accuracy assessment



#### **Rules:**

Candidate Search Radius: 5-m. horizontally (X and Y)

Candidate Vote: 3-m. vertically (Z)

Candidate Testing: Closest height



### Method: Terrestrial (T) and Mobile (M)-based Lidar Sensors

Data Preparation	Feature Extraction	Accuracy
1- Acquisition	4- Stem detection (Hough Transformation: TreeLS-R)	7- Automated Tree Matching: 5-m. horizontally (X and Y) 3-m. vertically (Z)
2- Registration (ICP: CloudCompare)	5- Height derivation <b>(A)</b> (Z <sub>max</sub> - Z <sub>ground</sub> : Custom-R)	8- Validation (RMSE, Bias)
3- Data Preparation (SOR, CSF, IDW: lidR-R)	6- DBH estimation <b>(B)</b> (RANSAC and Circle Fitting: Custom-R)	
(a) TLS (a.1)	Highest section Highest Section Section width (B)	
(b) MLS	Lowest section	

### Method: Time Investments

PHASE	PRIMARY TASKS	
Data Collection	• Set up RTK/PPK	
	<ul> <li>Deploy &amp; survey sphere targets</li> </ul>	
	<ul> <li>Acquire DLS flight lines</li> </ul>	
	<ul> <li>Scan TLS positions</li> </ul>	
	• Walk MLS scan path	
Compilation & Pre-processing		
	• Export LAS files	
	<ul> <li>Register TLS/MLS scans</li> </ul>	
	<ul> <li>Co-register all sensors</li> </ul>	
	<ul> <li>Quality control checks</li> </ul>	
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Post-processing & Reporting		and a fact of the fact and the fact of the
	<ul> <li>Batch metrics extraction</li> </ul>	<pre>control</pre>
	<ul> <li>Generate CHM &amp; tables</li> </ul>	<pre>intermedia Control Contro</pre>
	• Export reports & logs	<pre>i metadotari - metadotari - metadotari i metadotari i metadotari - metadotari i metadotari</pre>
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### **Result:** *Tree Detection Accuracy*





### **Result: Accuracy of Height Across Sensors and Forest Conditions**





### **Result: Accuracy of DBH Across Sensors and Forest Conditions**



### **Result:** *Time investment*



### **Challenges:** Field Conditions and Sensor Limitations



#### **CONSEQUENCES:**

↑ Data volume → increased memory & storage demands

- ↑ Processing time → bottlenecks
  in operational workflows
  - ↑ Computational load → need for high-performance systems



## Summary of Findings and Practical Recommendations

### Conclusion

- TLS was the most accurate across all sites, especially for DBH (R<sup>2</sup> > 0.95), but required the longest time (~2.5× DLS in PSPs-complex sites).
- DLS was the most time-efficient (~2.5–2.8 hours/plot) but less accurate in complex understory, particularly for DBH.
- **MLS** offered a **balance** between accuracy and efficiency, though still affected by positional noise in denser stands.
- **Post-processing was the dominant time phase**, reaching up to **60% of total time** with MLS in VAN and 50% with TLS in BRL.

### Signal Recommendation

- Use **DLS** for rapid large-area inventories where ground detail is less critical.
- Use **TLS** for high-precision needs and detailed stem mapping.
- Use **MLS** as a logistically flexible option for operational inventories.
- Consider **hybrid sensor integration** (e.g., TLS/MLS for stems + DLS for canopy).
- Focus on automating post-processing to reduce overall workflow time.





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### THANK YOU







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