









Mobile Laser Scanning for Estimating Tree Structural Attributes in a Temperate Hardwood Forest

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Context

Mr. Oak



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Investigation

GeoSLAM Zeb-REVO



Limited

range

B 230 (E) 220 210 200 MLS 303300 203305 30310 30315 30320 2033;









GeoSLAM Horizon & Hovermap (Emesent)



GeoSLAM Horizon

SLAM issue







Promising New Generation

Objectives

Evaluate the **potential of high-density mobile laser scanning (MLS) data** to estimate several forest inventory attributes in a mature hardwood stand, such as (i) tree **height**, **crown dimensions** and **diameter**, and (ii) **merchantable wood volume** using quantitative structural modeling (QSM) algorithms



Destructive Sampling



Terrestrial Laser Scanning



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Material

Study Site



Mobile Laser Scanning (MLS)



<u>SLAM-based MLS</u>: Simultaneous Localisation and Mapping <u>Characteristics</u>: 1,8 kg; 1H autonomy (x number of batteries) <u>Acquisition time</u>: **45 min (1 ha)** – <u>Automatic preprocessing</u>: 3H

New Brunswick

1 ha hardwood

75-160 years 500 trees/ha 26 sample trees

Leaf-off (Oct.2021)

(destructive measures)

© Hovermap

Terrestrial Laser Scanning (TLS)



<u>TLS</u>: multi-scan acquisition (5 scans) <u>Characteristics</u>: 5 kg; 4H autonomy (x number of batteries) <u>Acquisition time</u>: **12 Hours (9 sample plots)** – <u>Preprocessing</u>: 4.5H

Data Collection



Methods

Overall Data Processing Workflow



Methods

Manual Individual Tree Segmentation



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Filtering Process

Crown Range Filter (30 m)

• TLS

- MLS no filter
- MLS filtered

Filters: R package VoxR (Lecigne et al., 2018)

Sor Filter & Range Filter (15 m)

Quantitative Structural Modeling

Quantitative Structural Modeling (QSM) : *TreeQSM* (Raumonen et al., 2013) QSM parameters optimization: Martin-Ducup et al. (2021)



Truncated QSMs





Comparison of (**A**) tree height (H); (**B**) crown projected area (CPA); (**C**) crown volume (CV) and (**D**) diameter at breast height (DBH) for the 26 TLS and MLS trees. The blue line represents the linear regression. The grey zone is the 95% confidence band for predictions. The dashed black line represents the 1:1 line.

Operational Merchantable Wood Volume

• **Destructive sampling** (main stem and branches)

Α

С



• Allometric model (main stem; Li & Weiskittel)

 $d = \alpha_0 D^{\alpha_1} H^{a_2} X^{\beta_1 z^4 + \beta_2 (1/e^{D/H}) + \beta_3 X^{0.1} + \beta_4 (1/D) + \beta_5 H^{Q} + \beta_6 X}$



(**A**,**B**) Comparison of operational merchantable volume derived from the destructive field inventory (FI) (i.e., stem and branches with small end diameter outside bark (DOB) \geq 8 cm and length \geq 244 cm) with the QSM-derived value (i.e., branching order 0–3 with small end DOB \geq 8 cm and length \geq 244 cm) from (**A**) TLS and (**B**) MLS data. (**C**,**D**) Comparison of the merchantable stem volume derived from Li and Weiskittel's taper model (i.e., main stem with small end DOB \geq 8 cm) with the QSM-derived value of the stem (i.e., branching order 0 with small end DOB \geq 8 cm) from (**C**) TLS and (**D**) MLS data. *n* = 26 trees. The blue line represents the linear regression. The grey zone is the 95% confidence band for predictions. The dashed black line represents the 1:1 line.

Merchantable Wood Volume by Branching Order



- Order 0 (main stem)
- **Order 1** (1st level branches)
- Order 2 (2nd level branches)
- Order 3 (3rd level branches)



(A–D) Comparison of merchantable volume (i.e., stem and branches with small end diameter outside bark (DOB) ≥ 8 cm) derived from TLS and MLS QSMs presented by branching order (with small end DOB ≥ 8 cm): (A) order 0 = main stem; (B) order 1 = first-level branches; (C) order 2 = second-level branches; (D) order 3 = third-level branches. *n* = 26 trees. The blue line represents the linear regression. The grey zone is the 95% confidence band for predictions. The dashed black line represents the 1:1 line.

Merchantable Wood Volume (Total: order 0-3)



- Order 0 (main stem)
- **Order 1** (1st level branches)
- Order 2 (2nd level branches)
- Order 3 (3rd level branches)



(A) Comparison of total merchantable volume (i.e., stem and branches with small end diameter outside bark (DOB) \ge 8 cm) derived from TLS and MLS QSMs (i.e., branching order 0–3 with small end DOB \ge 8 cm). *n* = 26 trees. The blue line represents the linear regression. The grey zone is the 95% confidence band for predictions. The dashed black line represents the 1:1 line. (B) Boxplots of the proportion of merchantable volume by branching order (0 = main stem; 1 = first-level branches; 2 = second-level branches; 3 = third-level branches) from TLS QSMs. Red triangles represent the %RMSE by branching order.

Discussion

Key findings

- SLAM-based Mobile Laser Scanning (MLS) was efficient to scan a 1 ha stand
- Key inventory attributes (e.g. tree Height, CPA, DBH) were estimated with high accuracy
- Operational merchantable volume estimated from MLS data was close to the field measurements
- Branching volume derived from MLS was overestimated when compared to TLS

- Applicability of SLAM-based MLS data

- Practical and simple to use in forest environment
- Flexible data acquisition pattern interesting to acquire various size of sample plots
- Minimize signal occlusion
- Short acquisition time makes it more flexible against wind
- Automatic preprocessing

Limits

- Observations were done under favorable forest conditions (flat terrain, little low vegetation)
- Further study is required to understand the impact of walking pattern acquisition on data quality and data completeness
- Manual tree segmentation: not applicable on large scale
- QSMs algorithms: complex to parametrize, work at the tree-level; require clean point clouds

- Analyse the impact of walking pattern acquisition on tree structural attributes accuracy



Next Steps

- Investigate automation processes at the plot level (ITD, tree metrics, QSMs) ...



SLAM-based MLS data segmented from Computree (<u>https://computree.onf.fr/?page_id=497</u>)

Next Steps

- ... from a variety of hardwood stands



McCoy Brook Forest (New-Brunswick)



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Raw MLS data from a 1 ha uneven-aged northern hardwood stand using a similar 20 m x 20 m walking pattern acquisition (Slice of 4 m x 140 m)



What for?

- Enhancing forest inventories
- Supporting operation planning
- Assisting harvesting activities
- Improving inventory tracking in the supply chain...

... a first step towards a better assessment of

- Volume, biomass
- Species
- LAI, competition, allometric models, growth, ...

Thank you for your attention! Questions? <u>bastien.vandendaele@usherbrooke.ca</u>

More info: <u>https://doi.org/10.3390/rs14184522</u>

Open to Collaborations

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Canada https://hardwoodsnb.ca/en











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