Major Woody Parts Segmentation in Forest Point Clouds with Normalized Intensity

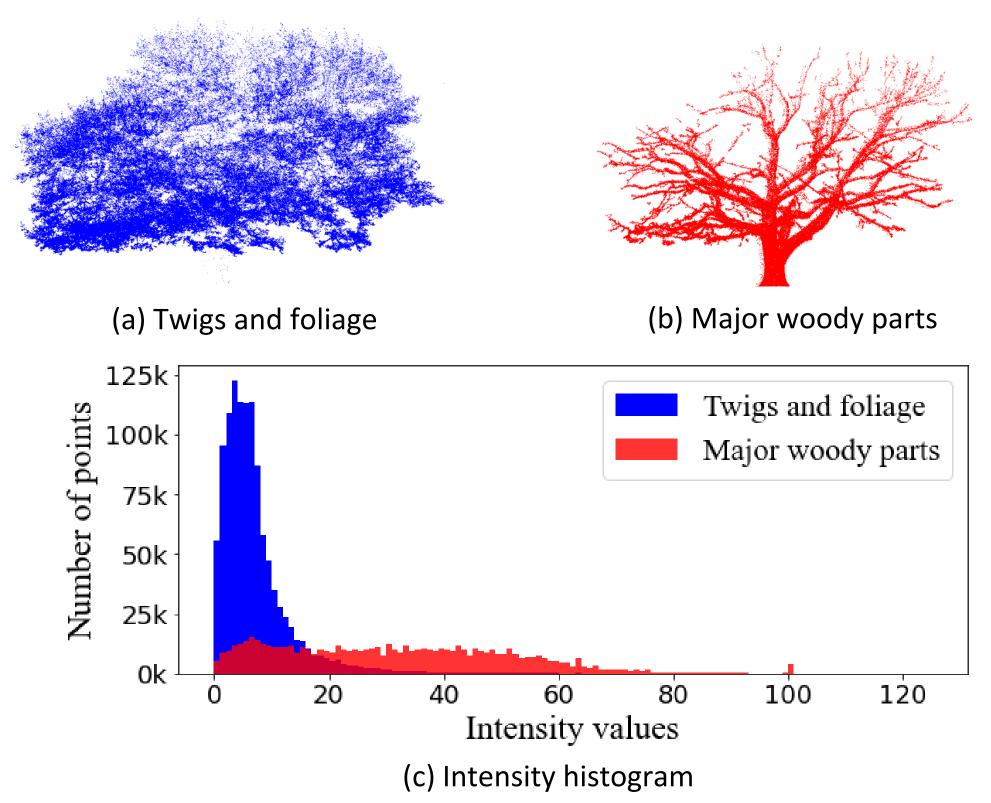
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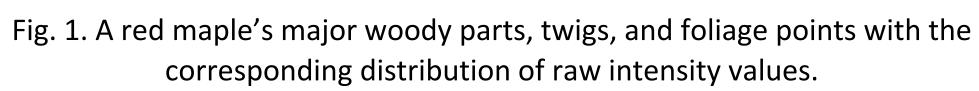


We present a fast and fully unsupervised method for segmenting major woody parts using LiDAR data from a backpack Mobile Mapping System (MMS). Our method combines automatic thresholding of normalized radiometric and geometric features to extract major woody parts in point clouds. Compared to existing methods, our approach achieves higher performance on 14 trees with different sizes and species in both leaf-on and leaf-off seasons. Unlike static Terrestrial Laser Scanning, our MMS LiDAR data is more efficient and accurate due to its non-destructive and dynamic scanning capabilities. Our results demonstrate the potential of our method for deriving structural and biophysical attributes of trees.

INTRODUCTION

- Major woody parts dominate the majority of biomass and timber and so play an important role in a tree.
- Terrestrial Laser Scanning (TLS) has been used for major woody parts segmentation in recent years.
- However, TLS is slow and not scalable to capture large areas in a reasonable time.
- LiDAR intensity hasn't been fully exploited; it shows significant differences between major woody parts, twigs, and foliage (shown in Fig. 1.)
- With a backpack LiDAR system (shown in Fig. 2), we propose a fully unsupervised method that combines normalized radiometric and geometric information.



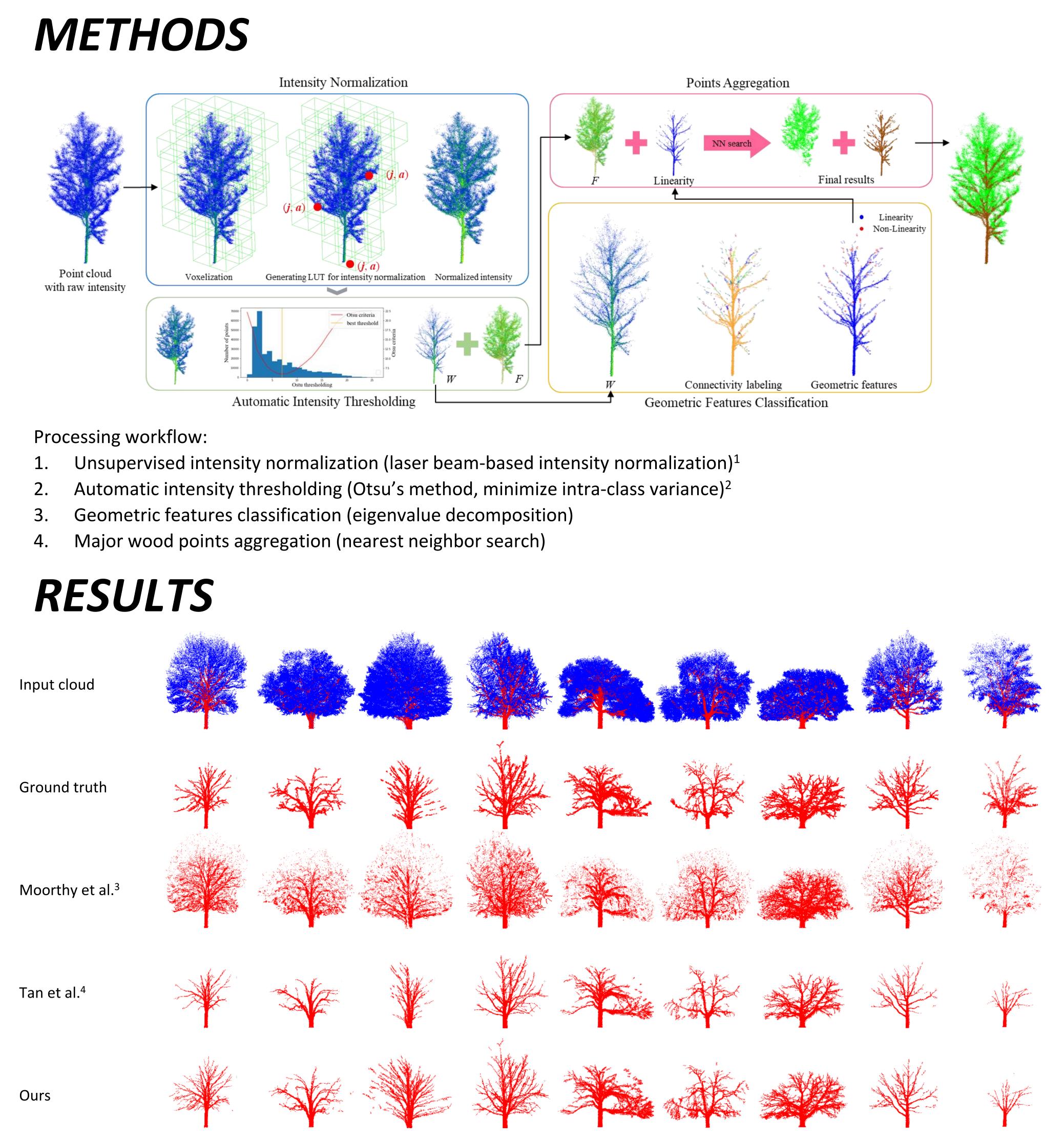


Integration of the backpack LiDAR system: Antenna

- LiDAR scanner (Velodyne VLP-16 HiRes)
- Micro computer (Raspberry Pi)
- GNSS/INS system (NovAtel SPAN-CPT)
- Camera (Sony α7RII + 32mm lens)

Fig. 2. The backpack system with its individual components.





Callery Pear European Beech Ginkgo

Ash

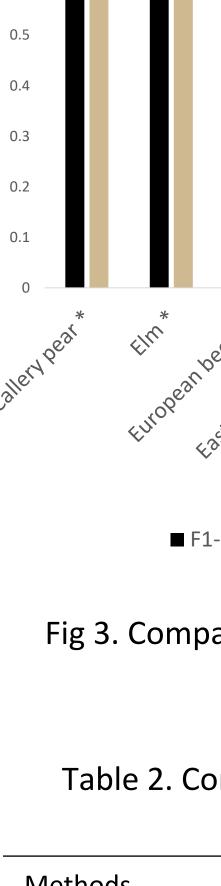
Pin Oak

Red Maple

Red Oak

Sweetgum





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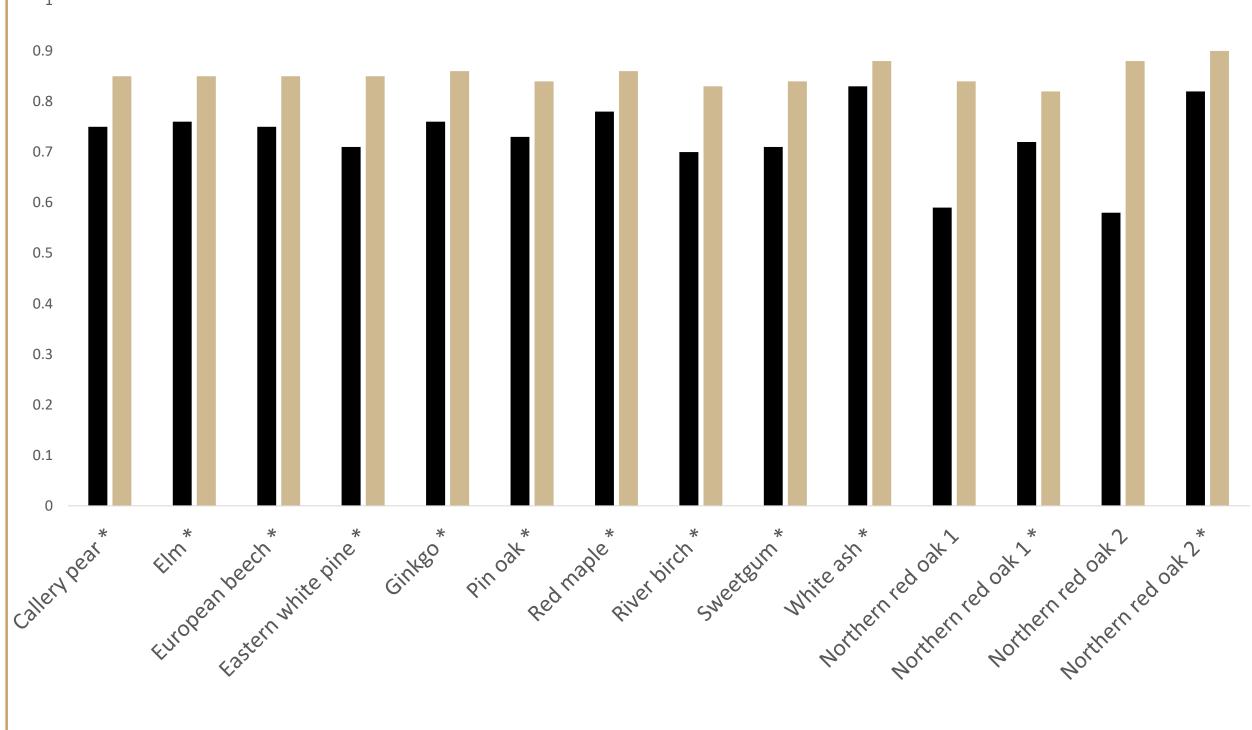


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Elm



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■ F1-score (before intensity normalization) F1-score (after intensity normalization)

Fig 3. Comparison between before- and after-intensity normalization in different tree species (* represents leaf-off season)

Table 2. Comparison between the proposed method and others in prior works

| Running time | Precision | Recall | F1-score |
|--------------|-----------|--------|----------|
| 3.5 hours | 0.46 | 0.55 | 0.50 |
| 15 seconds | 0.68 | 0.72 | 0.70 |
| 8 seconds | 0.87 | 0.86 | 0.86 |

DISCUSSION

approach is able to improve the accuracy of major ody parts segmentation across different tree species in erent seasons.

research provides new potential for quick data uisition and segmentation.

plan to expand this approach to major woody parts mentation at the plot- and stand-level.

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