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3D Modeling of Urban Tree Crown Volumes Using Multispectral LiDAR Data



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- <http://scsc.xmu.edu.cn/>



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Outline

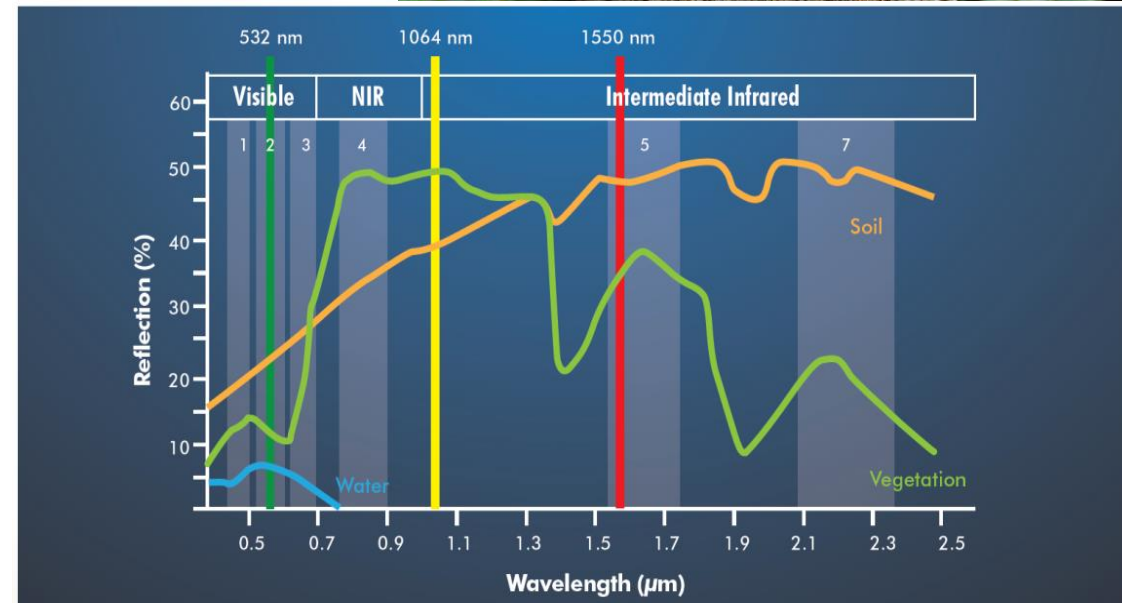
- Introduction
- Objectives
- Methodology
- Results and Discussion
- Concluding Remarks
- Our Software Solution



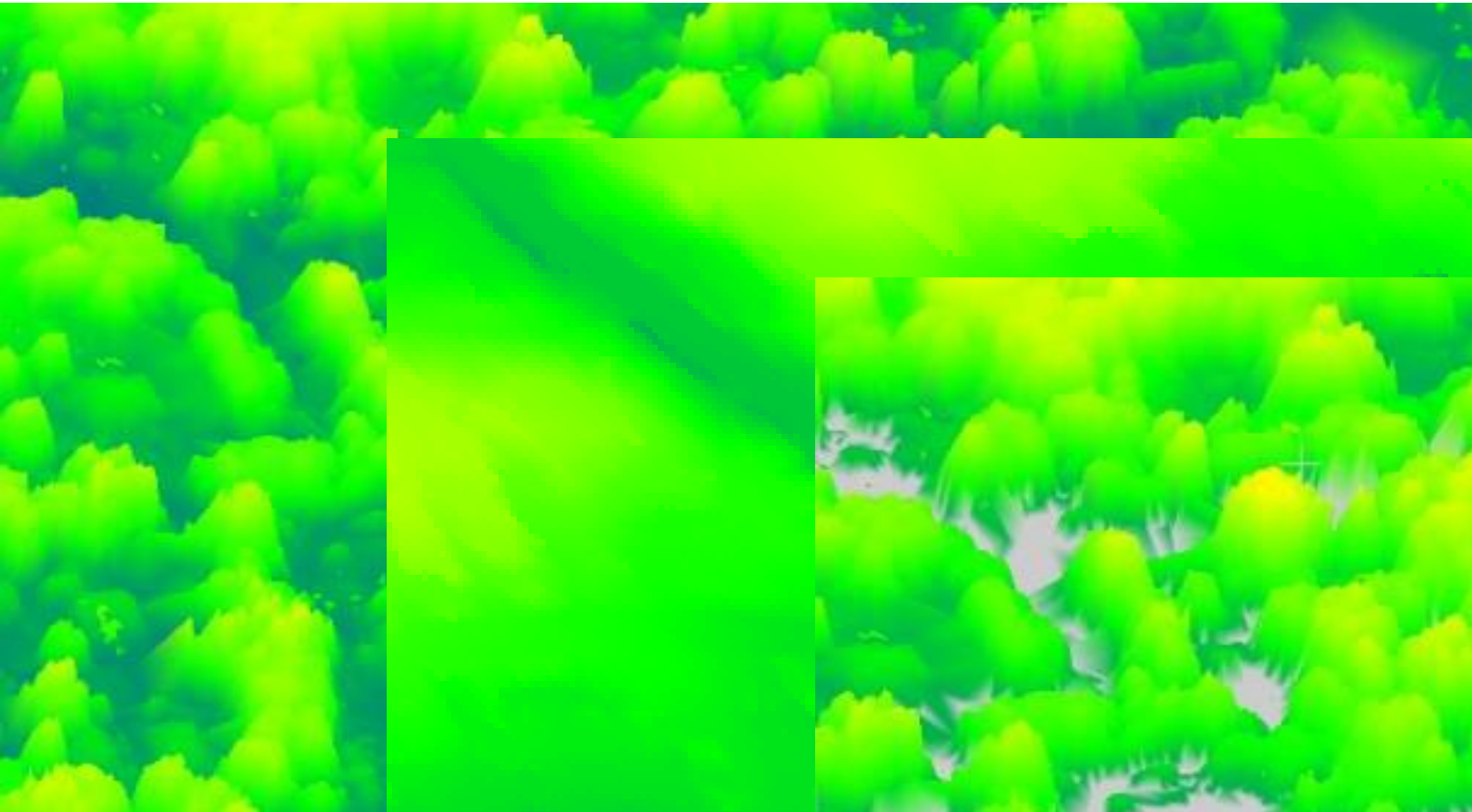
Introduction: Motivation of the Study

How to measure urban tree crown volumes?

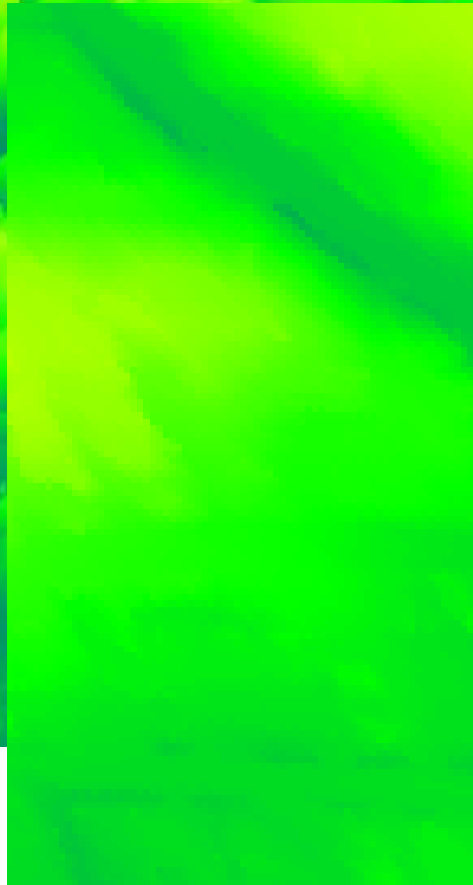
- Field Measurements
 - Expensive and time consuming
 - Destructive
- Remote Sensing Techniques
 - Multispectral imagery
 - Airborne LiDAR or ALS
 - Multispectral ALS**



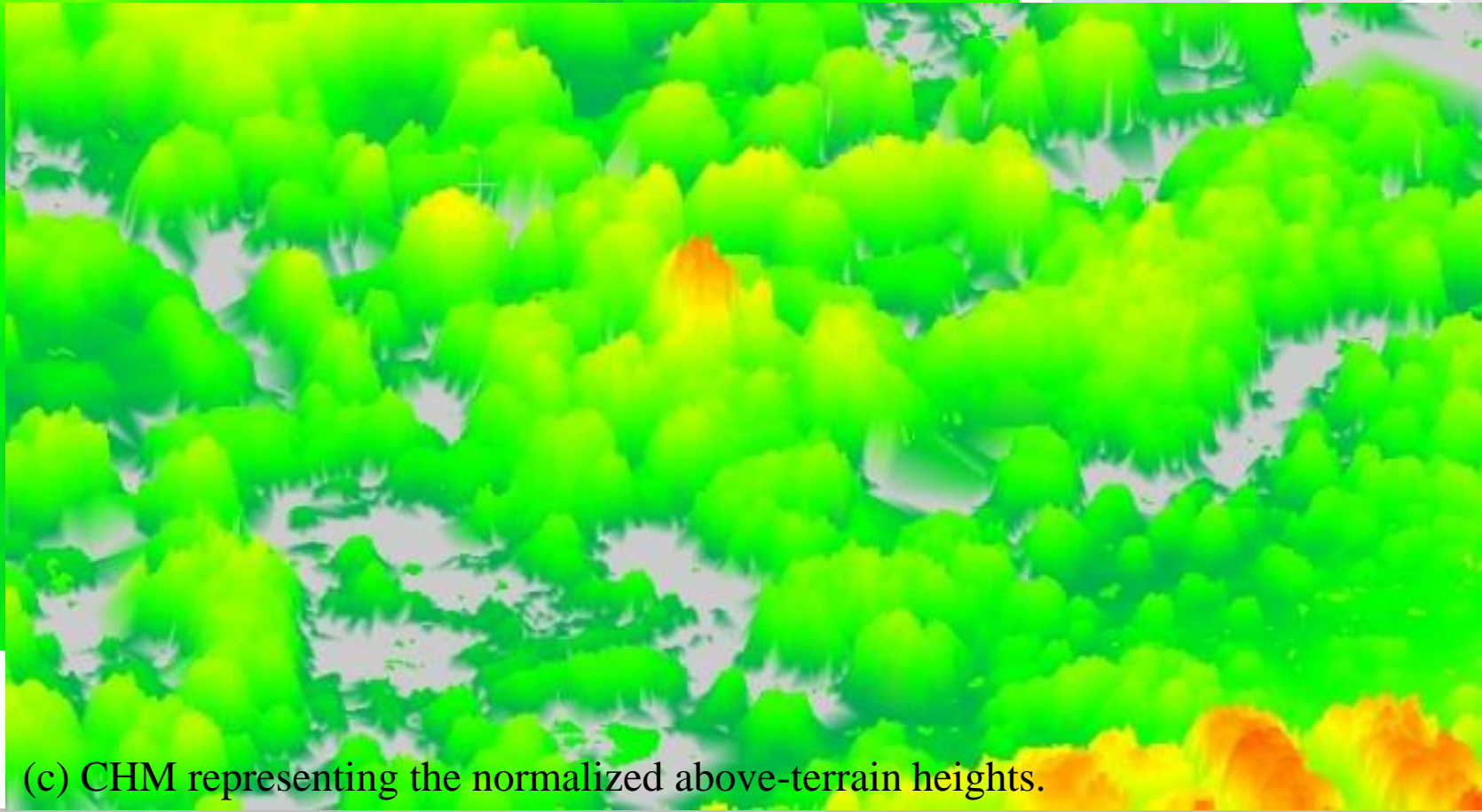
Introduction: ALS-derived products



(a) DSM



(b) DTM



(c) CHM representing the normalized above-terrain heights.



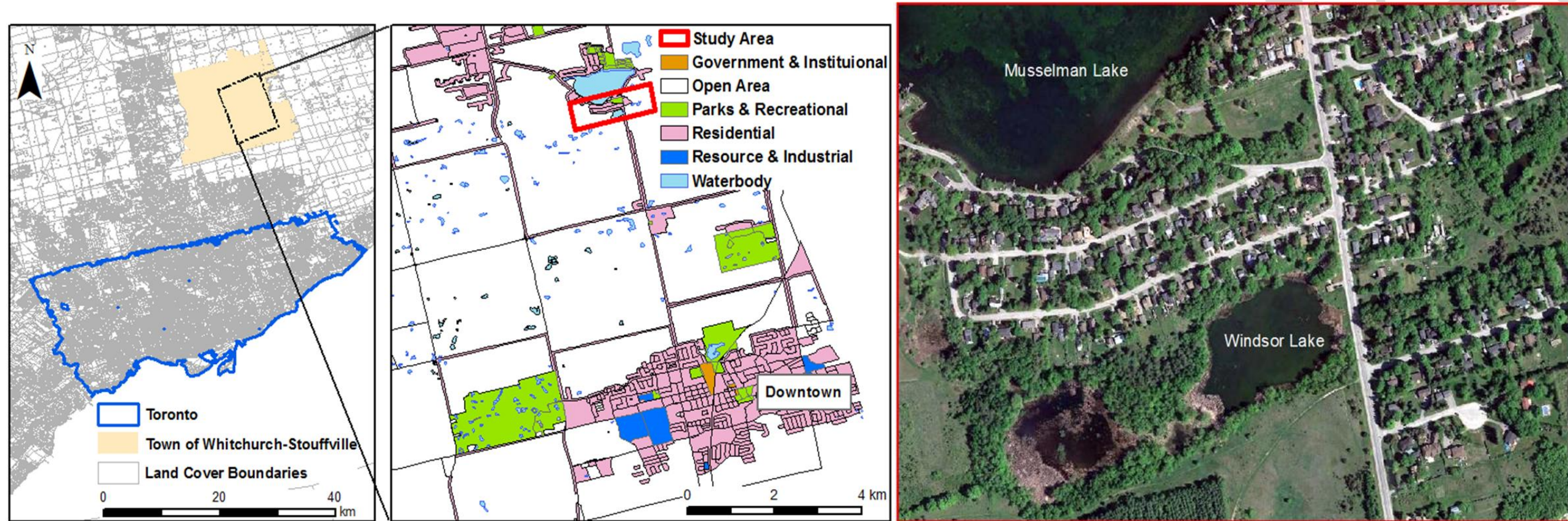
Objectives

To develop a workflow that can create **3D Models of Urban Tree Crown Volumes** using multispectral ALS data.

- to **classify vegetation covers** in urban areas using multispectral ALS ranging and intensity data and the land cover classifier;
- to **derive** dendrometric parameters such as **tree height and crown diameter** from the multispectral ALS data;
- to **establish allometric relationships** between the ALS-derived measurements (tree height and crown width) and the field-measured diameter at breast height (DBH).

Study Area & Data Sources

- **Study area**, Titan multispectral LiDAR datasets, field measurements



Data Providers: DMTI Spatial Inc.; Google & First Base Solutions

Titan Multispectral LiDAR Datasets

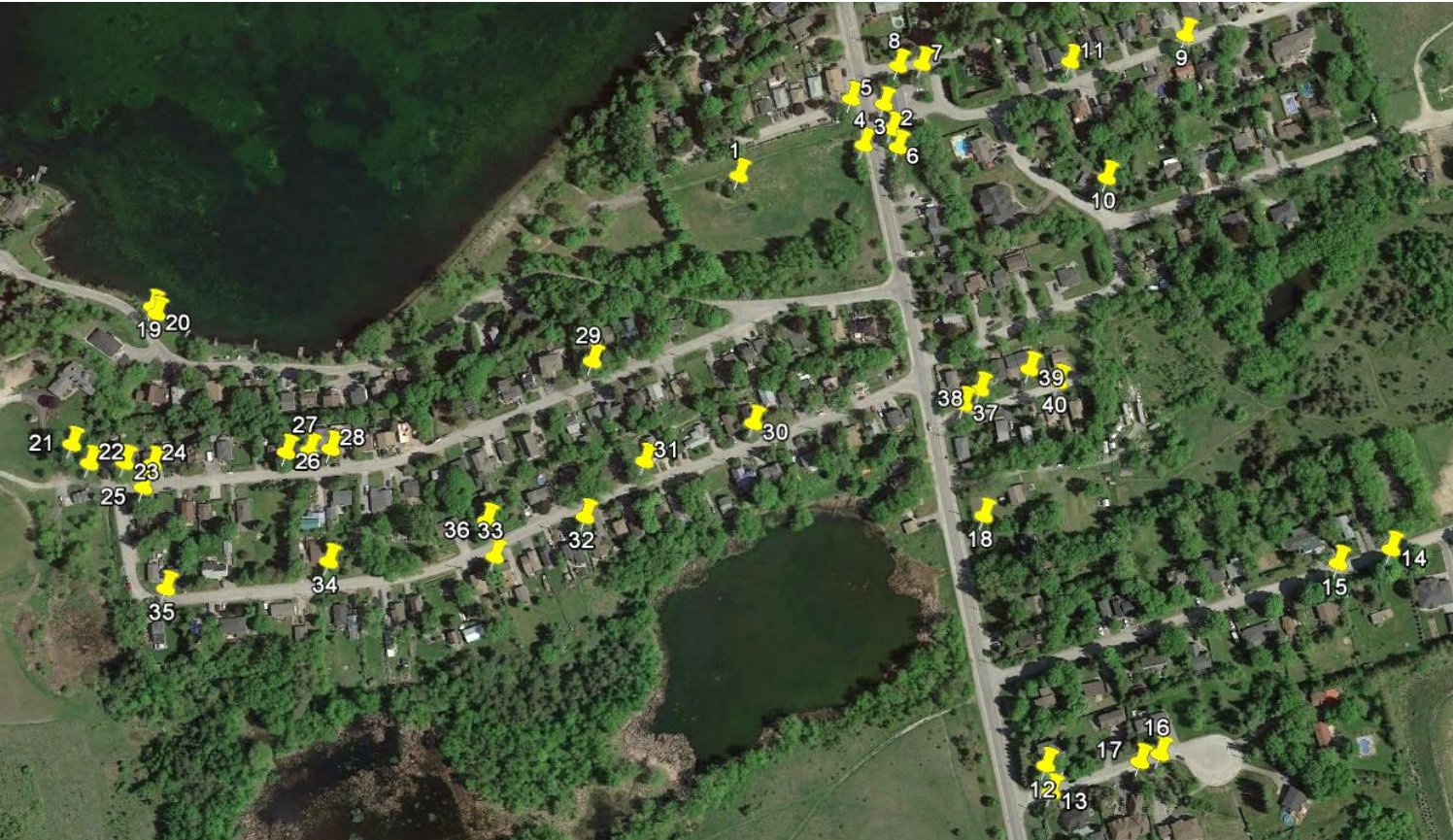


| | |
|----------------------------|---------------------------|
| Number of ALS strips | 2 |
| Laser channels | 1550 nm; 1064 nm; 532 nm |
| Flight of View (FOV) | 30° |
| Pulse repetition frequency | 100 kHz per channel |
| Flight Height | 1030 m; 1043 m |
| Average Point Density | 7.7 points/m ² |
| Average Point Spacing | 0.8 m/point |



Three laser channels at wavelength of 1550 nm (shortwave infrared, SWIR), 1064 nm (near infrared, NIR), 532 nm (green)

Field Measurements



| | Height (m) | Crown Width (m) | DBH (cm) |
|---------|---------------|--------------------|-------------|
| Maximum | 26.90 | 16.57 | 98.0 |
| Minimum | 9.70 | 4.64 | 27.0 |
| Mean | 17.48 | 9.11 | 48.8 |
| Std | 4.57 | 3.07 | 14.6 |

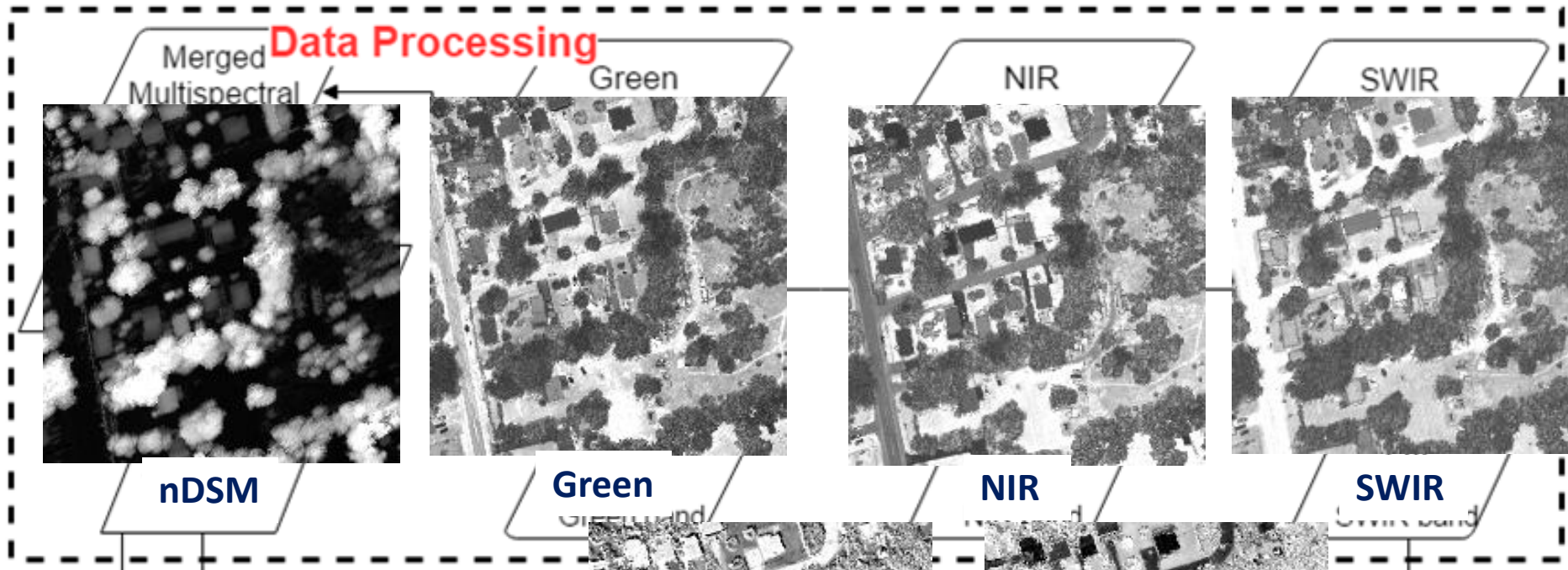
A total of 40 trees are selected from the field Tree heights were measured using a hypsometer. DBH was measured with a diameter.

Methodology

- ALS data processing
- Vegetation isolation
- Dendrometric parameter estimation
- Allometry-based ALS-DBH modeling



Step 1: ALS Data Processing

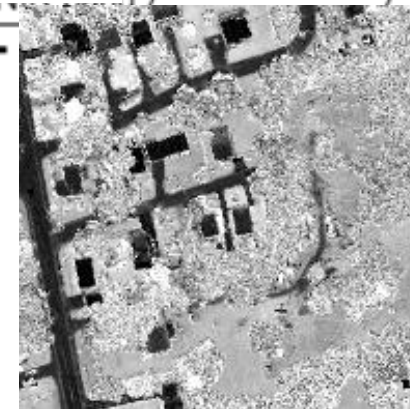


$$pNDWI = \frac{C_{Green} - C_{NIR}}{C_{Green} + C_{NIR}}$$

$$pNDVI = \frac{C_{NIR} - C_{SWIR}}{C_{NIR} + C_{SWIR}}$$

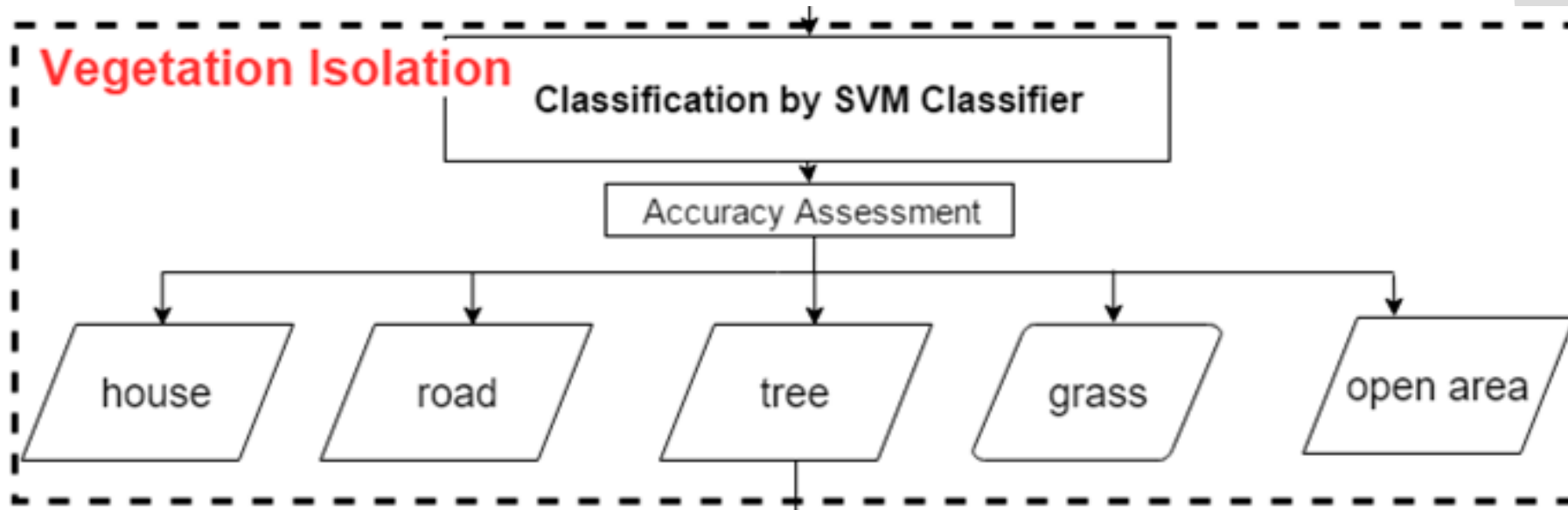


pNDWI



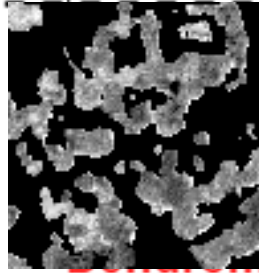
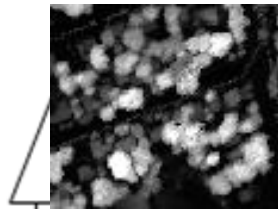
pNDVI

Step 2: Vegetation Isolation



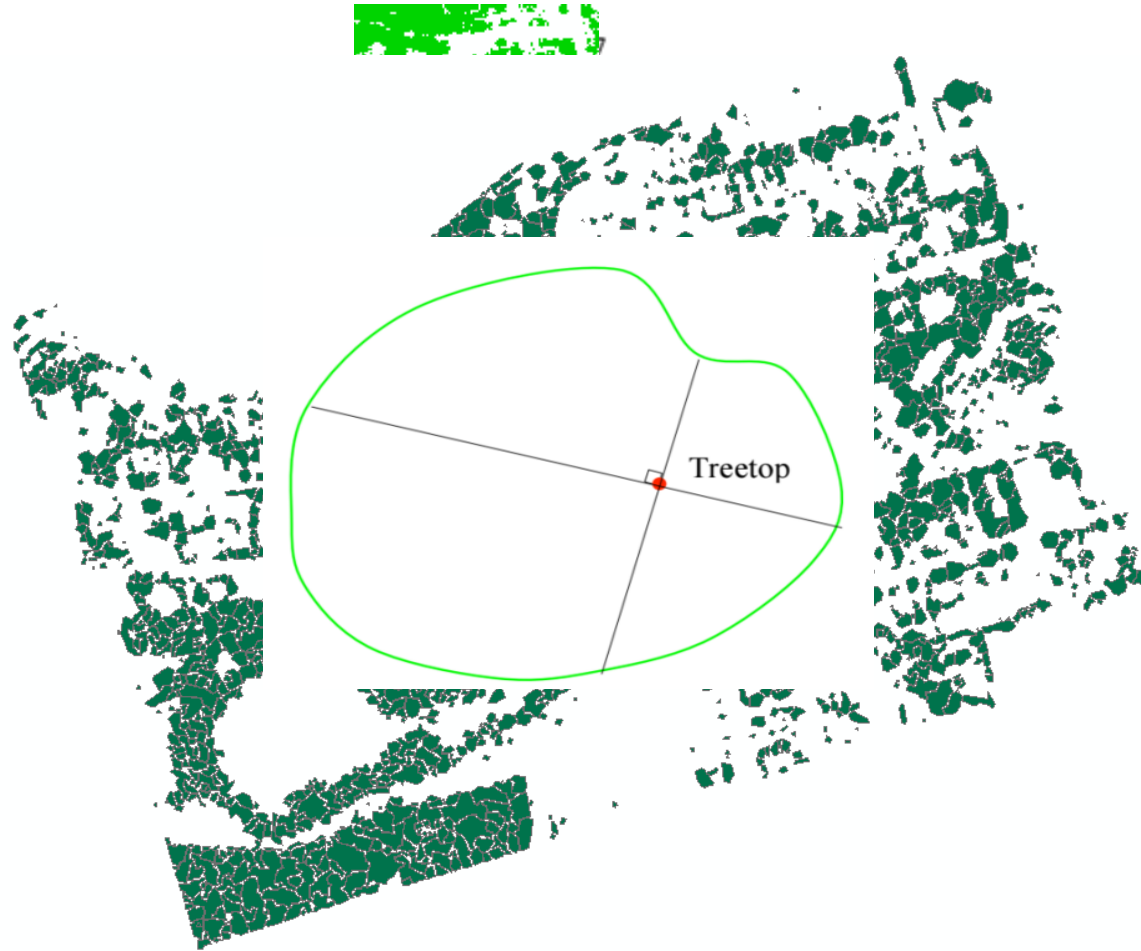
| Combinations of input data for SVM classification | |
|---|---|
| Selection 1 | Green + NIR + SWIR + nDSM + pNDWI + pNDVI |
| Selection 2 | Green + NIR + SWIR + nDSM |
| Selection 3 | NIR + nDSM |

Step 3: Dendrometric Parameter Estimation



Parameter Estimation
with AI

Validation with
Google Earth
photos



Light Model (CHM)



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Step 4: Allometry-based ALS-DBH modeling

- A multiple linear regression model was developed empirically to predict the DBH by the ALS-derived tree height and crown width.

$$DBH_{Field} = a \cdot CD_{ALS} + b \cdot H_{ALS} + c$$

$$a = \frac{(\sum H_{ALS}^2)(\sum CD_{ALS} \cdot DBH_{Field}) - (\sum CD_{ALS} \cdot H_{ALS})(\sum CD_{ALS} \cdot DBH_{Field})}{(\sum CD_{ALS}^2)(\sum H_{ALS}^2) - \sum CD_{ALS} \cdot H_{ALS}^2}$$

$$b = \frac{(\sum CD_{ALS}^2)(\sum H_{ALS} \cdot DBH_{Field}) - (\sum CD_{ALS} \cdot H_{ALS})(\sum H_{ALS} \cdot DBH_{Field})}{(\sum CD_{ALS}^2)(\sum H_{ALS}^2) - \sum CD_{ALS} \cdot H_{ALS}^2}$$

$$c = \overline{DBH} - a \cdot \overline{CD_{ALS}} - b \cdot \overline{H_{ALS}}$$

Results and Discussion

- Accuracy assessment of the classification results
- Performance of the watershed segmentation
- Validations for the ALS-derived dendrometric parameters
- Validation of the ALS-DBH linear regression modeling



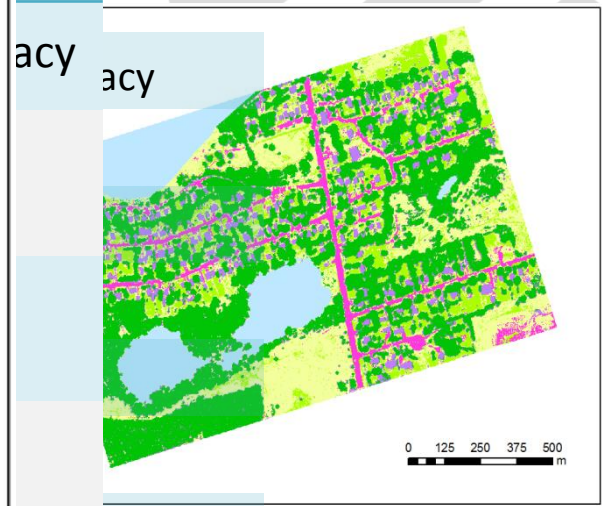
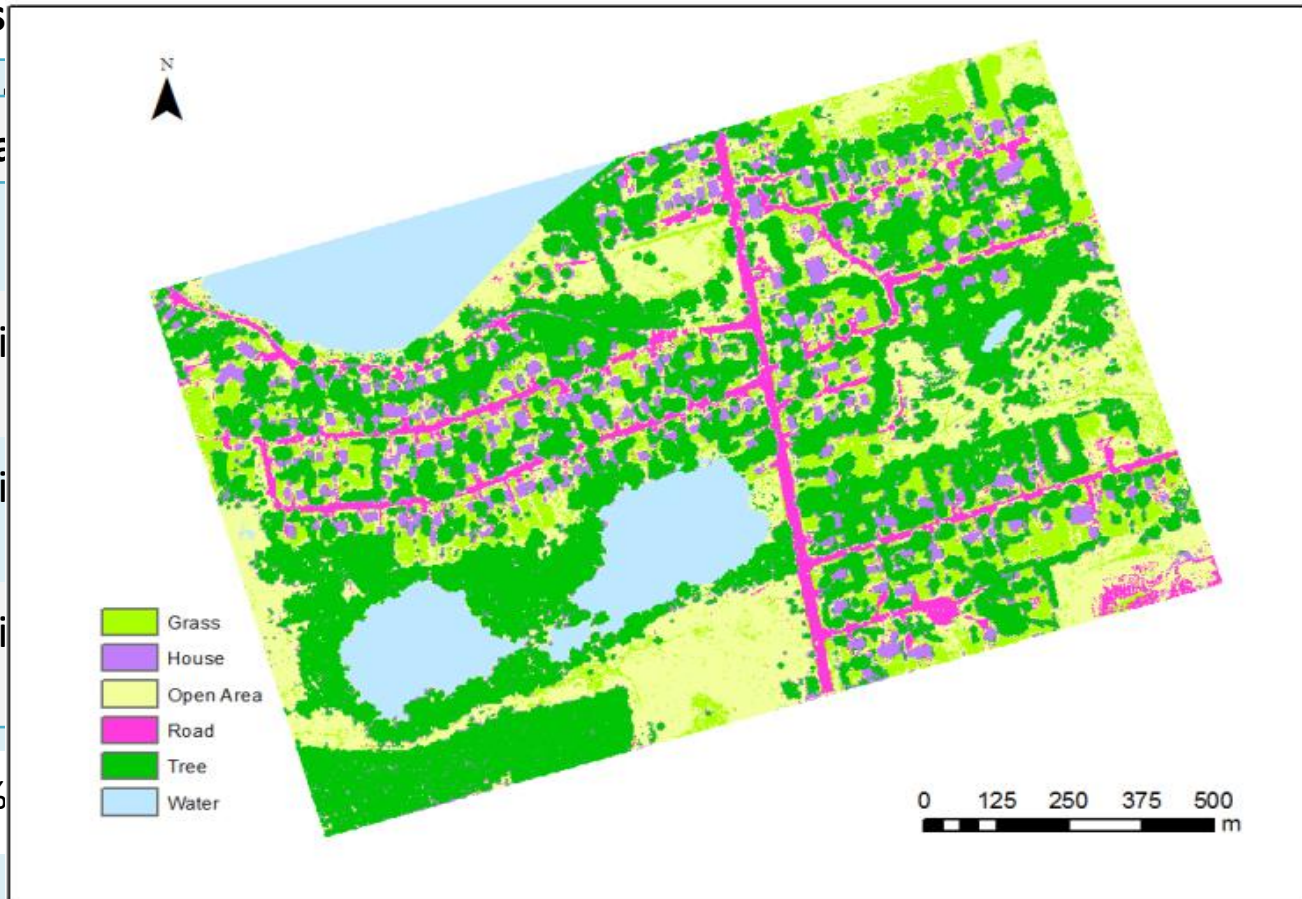
Assessment of Classification Accuracy



Confusion Matrix for the Classification

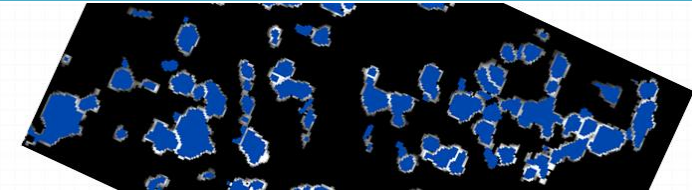
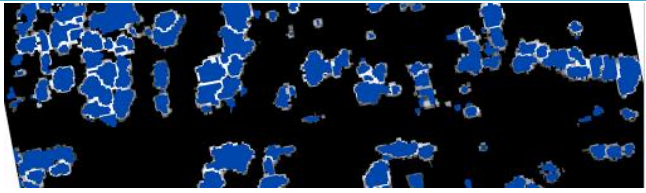
| | Actual |
|---------------------|--------|
| Classified Pixels | 1000 |
| Grass | 250 |
| Tree | 250 |
| House | 250 |
| Road | 250 |
| Open Area | 250 |
| Total | 1000 |
| Producer's Accuracy | 83.19% |

Overall Accuracy = 90.23%



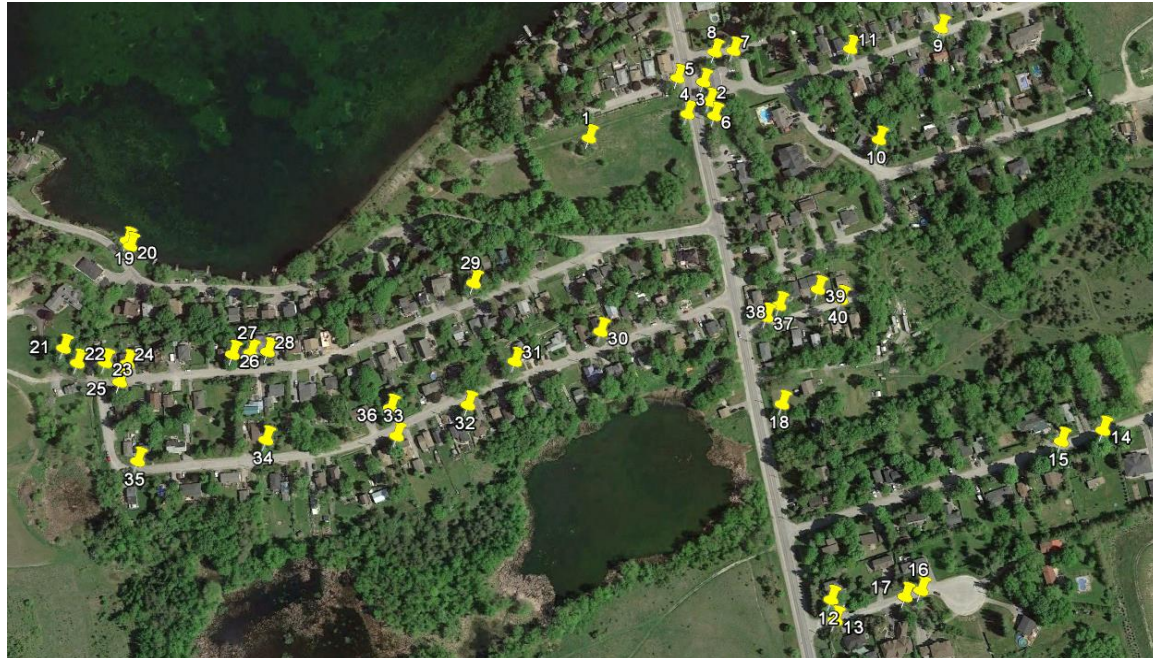
Performance of Watershed Segmentation



| Evaluations of the segmentation result | | |
|--|--|---|
| | Site1 | Site2 |
| Reference |  |  |
| Segmentation Results |  |  |
| $N_{1,1}$ | 49 | 80 |
| Total | 65 | 114 |
| Accuracy | 75.4% | 70.2% |

$$\text{absolute accuracy}_{\text{tree isolation}} = \frac{n_{1,1}}{n_{\text{total}}}$$

Validations of ALS-derived Dendrometric Parameters



$$\text{Bias} = \frac{\sum_{i=1}^n X_{ALS,i} - X_{field,i}}{n}$$

$$\text{RMSE} = \sqrt{\frac{\sum_{i=1}^n (X_{field,i} - X_{ALS,i})^2}{n}}$$

$$\text{RMSE\%} = \frac{\text{RMSE}}{\bar{X}_{ALS}}$$

Validation statistics for the ALS-derived dendrometric parameters

| Parameter | RMSE | RMSE% | Bias | Bias% |
|-----------------|------|-------|-------|-------|
| Height (m) | 1.21 | 6.8% | -0.20 | -0.1% |
| Crown Width (m) | 1.47 | 16.4% | 0.18 | 2% |

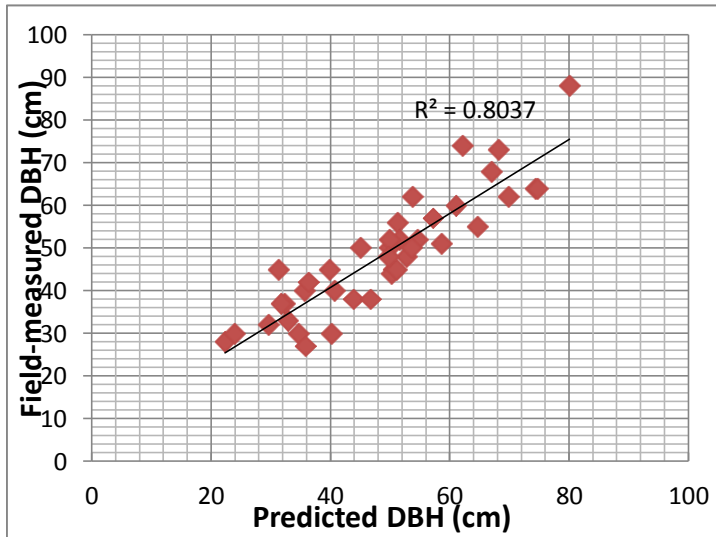


ALS-DBH Regression Models

| Models for predicting | | Results of model fitting and model validation | | | | Model Equation | |
|-----------------------|-----------------------|---|--------------------------|---------------------|---------------------------|----------------|--|
| Model | # of training samples | Model | Model Fit R ² | Model Fit RMSE (cm) | Validation R ² | | Validation RMSE (cm) |
| 1 | 20 | 1 | 0.83 | 5.35 | 0.80 | 6.82 | $= 4.12 - 0.03 \times CD + 2.51 \times H$ |
| 2 | 20 | 2 | 0.86 | 6.60 | 0.76 | 5.60 | $= -11.28 + (-0.30) \times CD + 3.26 \times H$ |
| 3 | 20 | 3 | 0.86 | 3.86 | 0.71 | 8.25 | $= 4.37 + (-0.38) \times CD + 2.59 \times H$ |
| 4 | 20 | 4 | 0.75 | 7.89 | 0.77 | 4.82 | $= -3.52 + 0.62 \times CD + 2.70 \times H$ |
| 5 | 20 | 5 | 0.78 | 5.59 | 0.85 | 6.55 | $= 7.28 + (-0.16) \times CD + 2.41 \times H$ |
| 6 | 20 | 6 | 0.81 | 7.00 | 0.83 | 5.20 | $= -13.15 + 0.05 \times CD + 3.43 \times H$ |



Validation of ALS-estimated DBH



| Accuracy of ALS-derived vs. field-measured results | | | | |
|--|------|-------|-------|-------|
| Parameter | RMSE | RMSE% | Bias | Bias% |
| DBH (cm) | 6.39 | 13.1% | -0.44 | -0.1% |
| | | | | |

Concluding Remarks

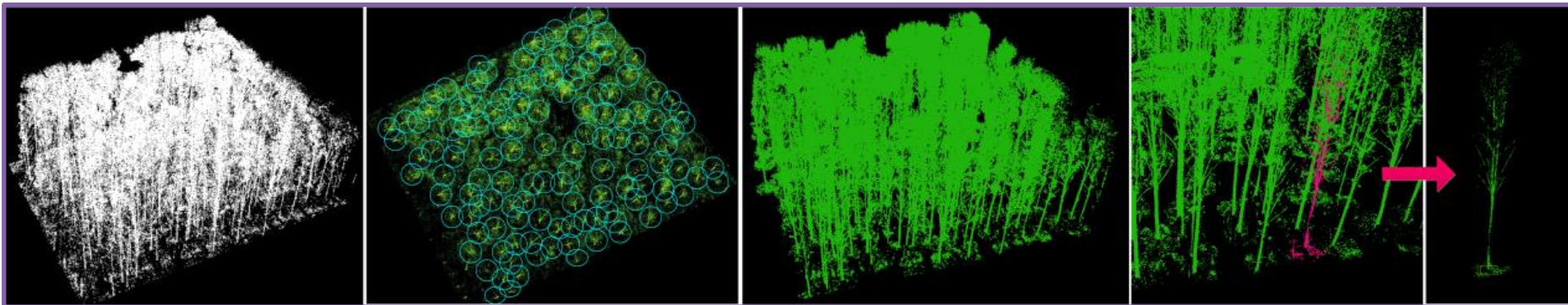
- A workflow for mapping a tree-covered urban area at the spatial resolution of 1 m was presented;
- A 90% land cover classification accuracy was achieved using multispectral ALS data;
- It was shown that the detection of treetops can be improved by the use of spectral and geometric properties of the multispectral ALS data;
- It was demonstrated that the DBH can be estimated using multispectral ALS data.
- Due to the scanning angle of ALS, the DBH cannot be directly measured using ALS compared with TLS or PLS.



Tree Scanning Using A TLS



Four forest scenes covered by
point clouds data I, II, III, and IV.
82.65 million points with a data
size of **2.61 GB**.
427 individual trees manually
counted as ground truth.



(a)

(b)

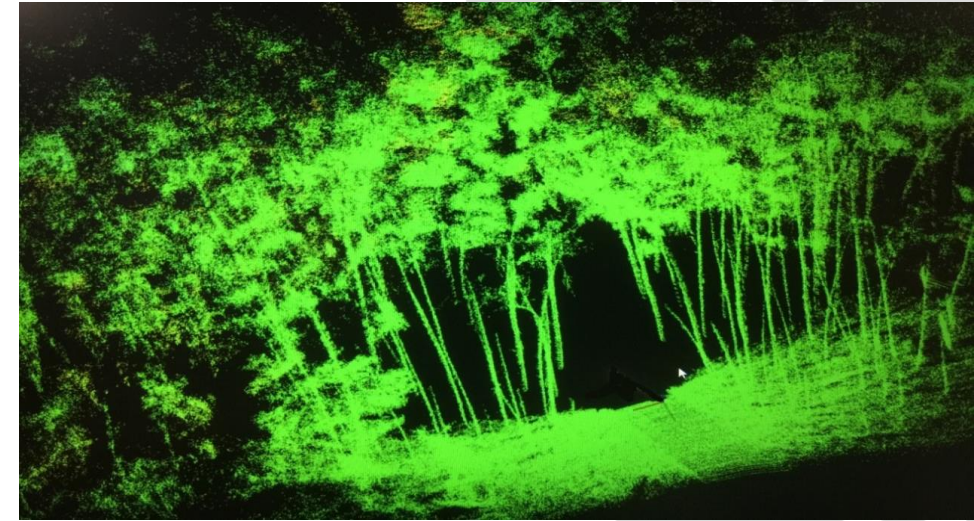
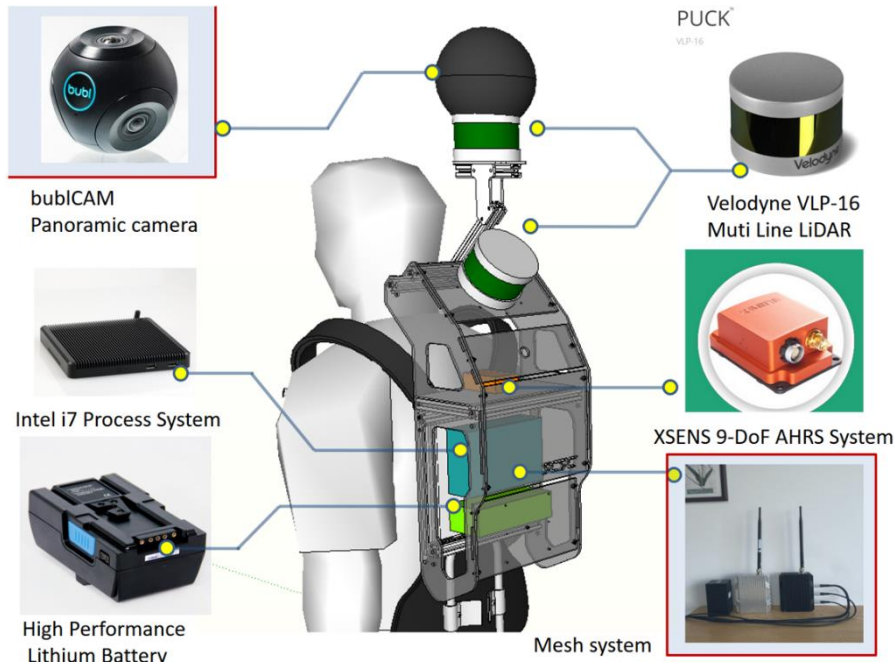
(c)

(d)

Results of individual tree extraction: (a) original point clouds; (b)-(c) tree detection result in top view and side view; (d) individual tree extraction result.



Tree Scanning Using A Backpack PLS



Thank you for listening! Any questions?

