Remote sensing data sources and applications

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Key collaborators

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Ressources naturelles

Conclusions / Opportunity

- Multi-source and multi-scale systems to capture change and quantify changes in carbon stocks over time
 - Optically-driven approaches for "activity" data are operational, facilitated by archival data and increasingly common open data policies
- Need an NFI
- Optical data for cover, extent, change (characterize, stratify)
 - temporally dynamic forest extent
- RS change to characterize change locations (and rates)
- RS change to target (as required) in situ measurements (TSP, REDD support, augment NFI)
- Number of plots required to provide certainty?
- Opportunity for airborne lidar to 1. emulate plots; 2. spatially extend ground measures

^{*} Based (loosely) on discussions with M. Hansen and E. Næsset.

Data sources

Plots

Inventories

Remote sensing

Models

Framework notion:

Integrating multi-scale remote sensing and modeling

Establish high spatial resolution, fine scale grid



 Populate grid through remote sensing (e.g., Landsat) and modeling



Update the grid through satellite change detection

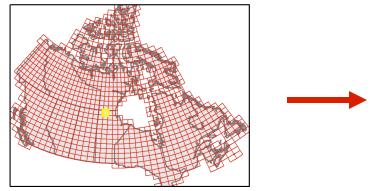


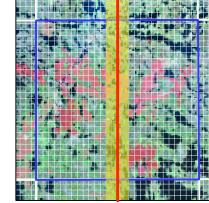
Find change, attribute change (as possible)

 Confirm / adjust modeled outcomes through sample based, higher spatial resolution remote

sensing

Repeat





RS / REDD linkages

- Disturbance
 - Landsat
- Degradation
 - High spatial resolution imagery
- Conversion
 - Landsat + interpretation (land use)
- Recovery
 - Landsat

Saturation

- Modeling: specifics of C characterization, such as dead organic matter (need for modeling)
- Remote sensing, saturation
 - Optical and Radar experience measurement saturation (empirical measures).
 - synoptic
 - Lidar does not saturate (more direct measures)
 - sample
 - Very High Spatial Resolution imagery (allometrics estimates possible)
 - sample

High spatial resolution imagery and analysis

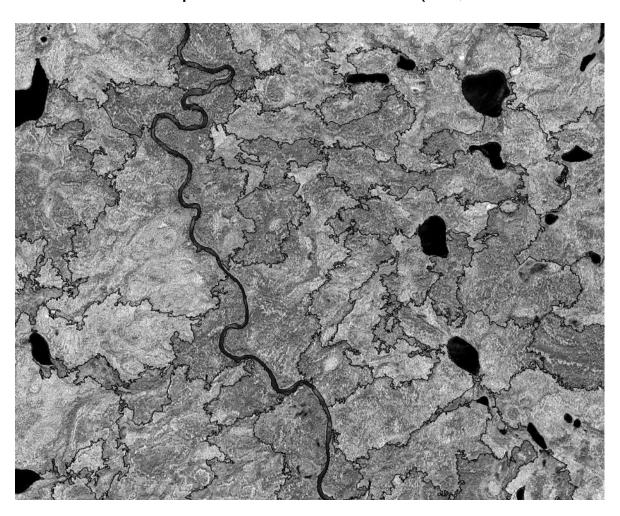
- Segmentation
 - Stands
 - Trees
 - Stand-level characterizations
- Automated attribution
 - Density
 - Height
 - Species



Falkowski, M.J.; Wulder, M.A.; White, J.C.; Gillis, M.D. 2009. Supporting large-area, sample-based forest inventories with very high spatial resolution satellite imagery. Progress in Physical Geography 33(3): 403–423.

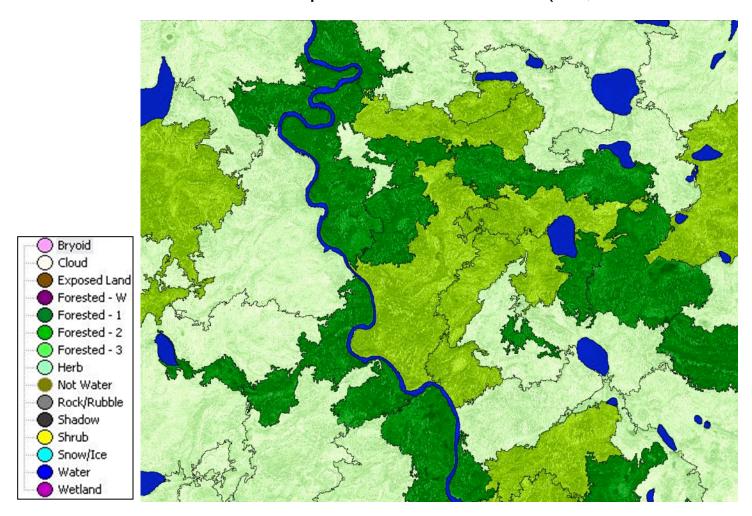
Automated Segmentation:

Goal: partition image into spatially distinct units that are homogenous and mutually distinct in terms of forest composition and structure (i.e., create stand boundaries)



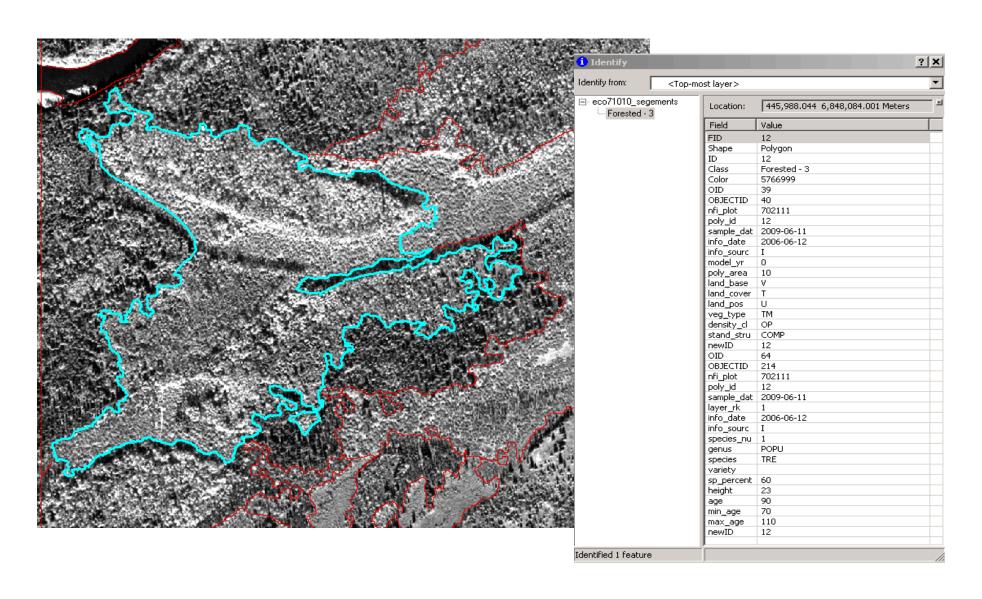
Automated Segmentation:

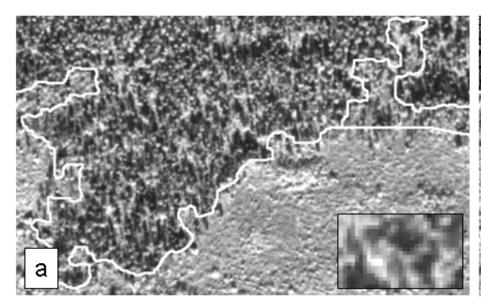
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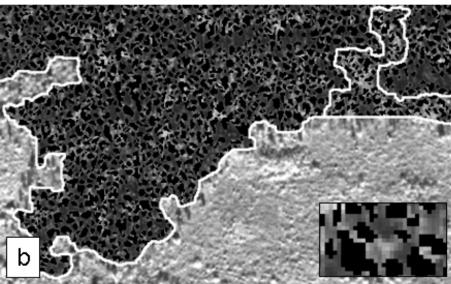


Attribution

Following segmentation, FI attributes are generated for each segment







Height estimation

- Input data
- Stand level metrics based on VHSR image, crown metrics
- Results:
 - significant R2 of 0.53** and an RMSE of 2.84 m

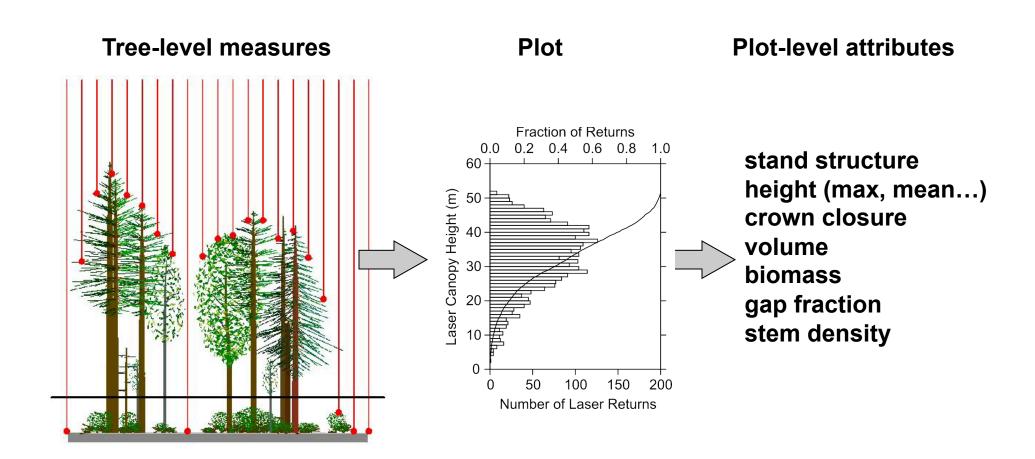
Leading species estimation

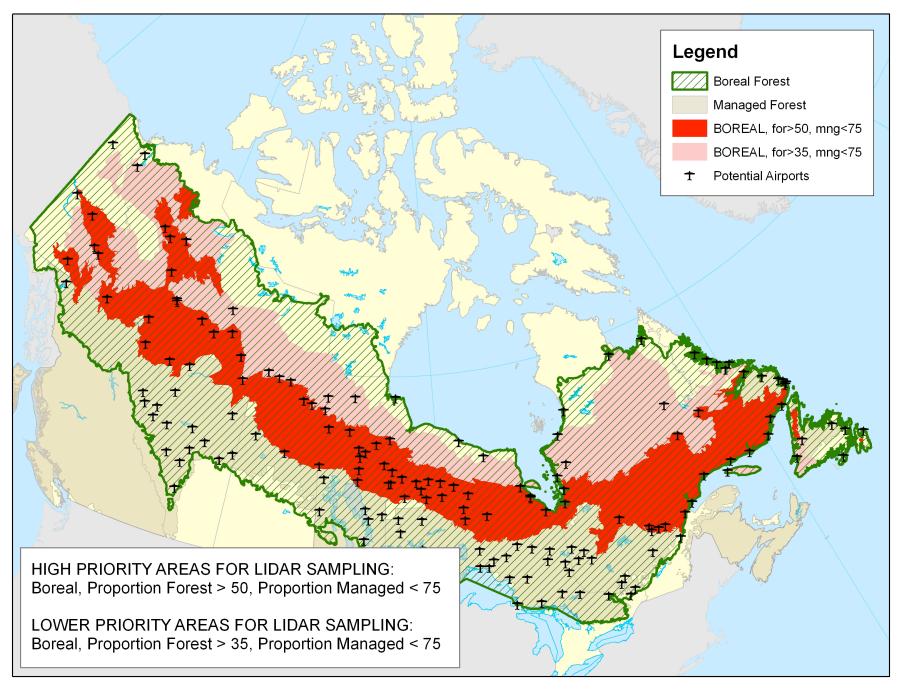
- Input data:
 - Statistics on tree crown shape metrics : Area, Length, Roundness
 - Mean, Variance, 25th, 50th, 75th percentiles
- Results:
 - Most selected metrics: Area variance, Roundness variance, Area 50th percentile
 - Overall accuracy of 74.5%

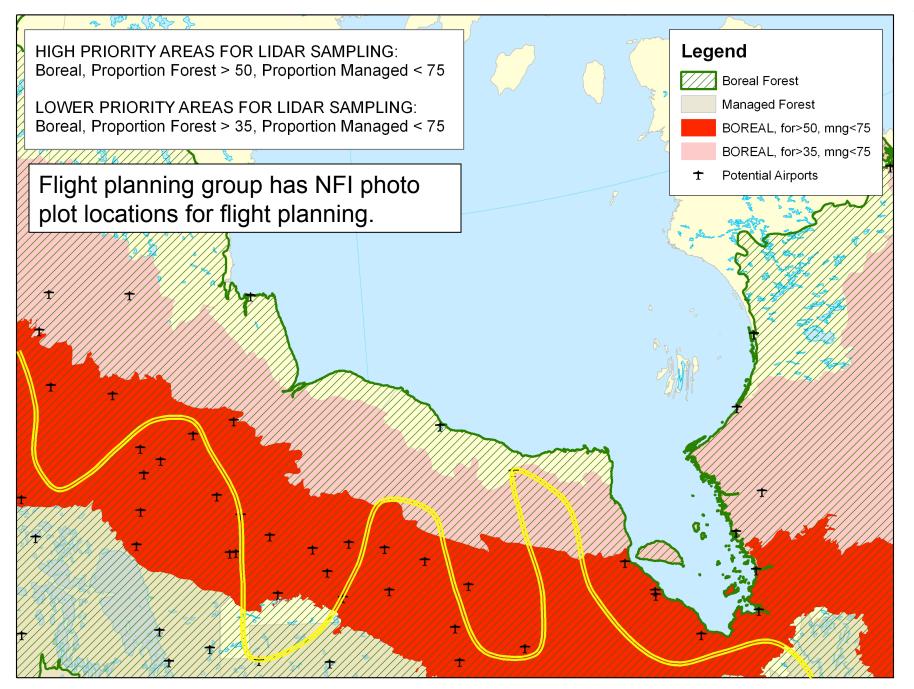
Mora, B., M.A. Wulder, and J.C. White (2010). Segment-constrained regression tree estimation of forest stand height from very high spatial resolution panchromatic imagery over a boreal environment. Remote Sensing of Environment. [DOI forthcoming]

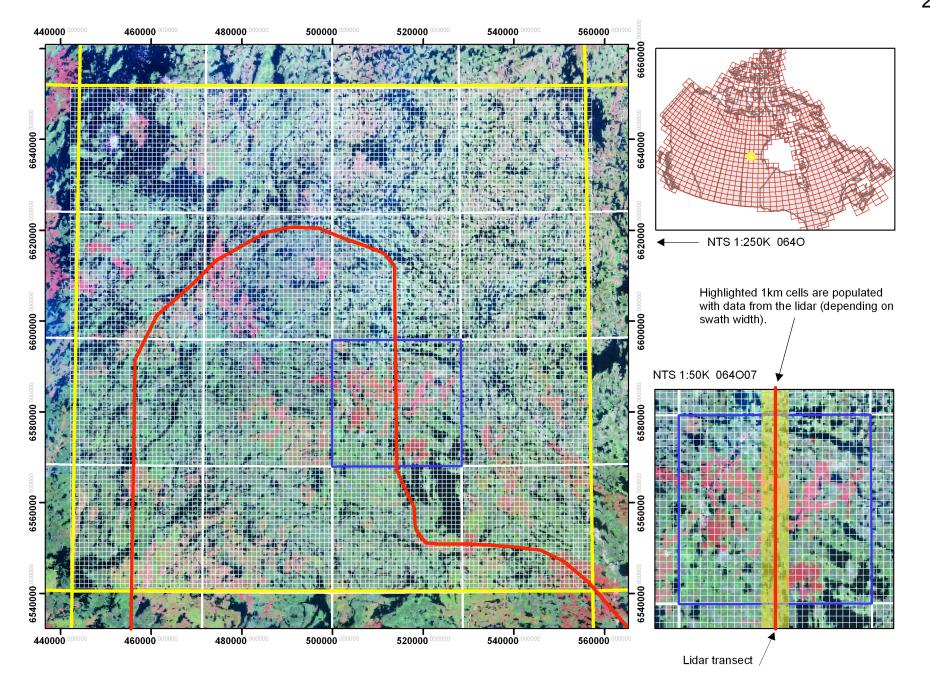
Mora, B., M.A. Wulder, and J.C. White (2010). Identifying leading species using tree crown metrics derived from very high spatial resolution imagery in a boreal forest environment. [Accepted]

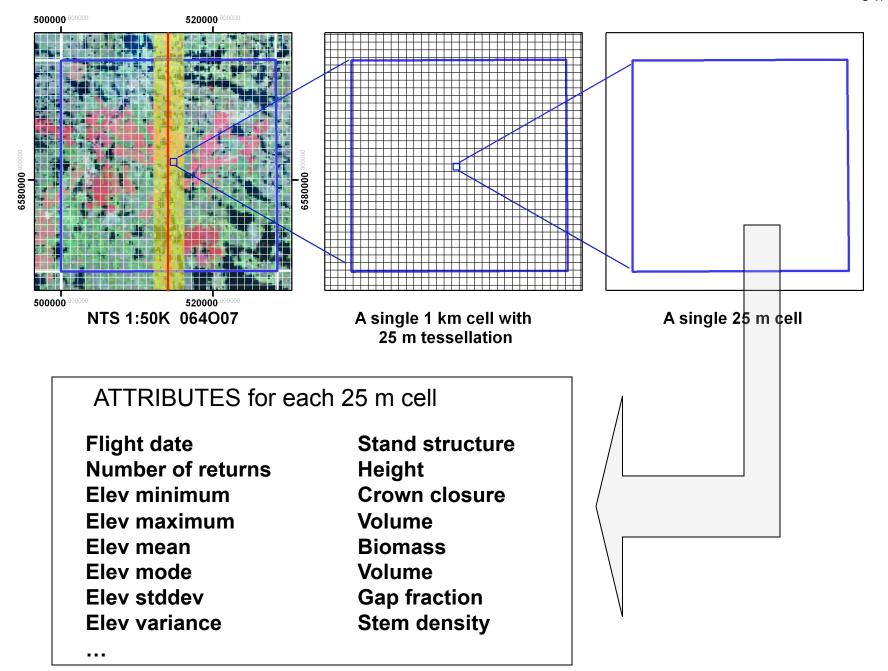
Individual tree measures from the lidar are summarized to produce *critical* plot-level attributes at thousands of plot locations.











Thank you

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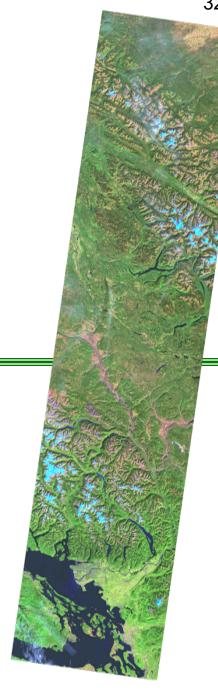


Natural Resources Canada

Canadian Forest

Ressources naturelles Canada

Service canadien des forêts



Recent Applications of Echidna® Lidar in the U.S.

A. H. Strahler¹, C. E. Woodcock¹, C. B. Schaaf¹, R. Myneni¹, J. Liu¹, G. J. Newnham³, D. L. B. Jupp⁴, D. S. Culvenor³, J. L. Lovell⁴, W. Ni-Meister², S. Lee², X. Li¹, F. Zhao¹, X. Yang¹, T. Yao¹, Q. Zhang¹, M. Schull¹, M. Roman III¹, Z. Wang¹, Y. Shuai¹

¹Boston University
²Hunter College of CUNY
³CSIRO Forest Biosciences
⁴CSIRO Marine and Atmospheric Research

Supported by NASA Grant NNG-0GG192G To Boston University

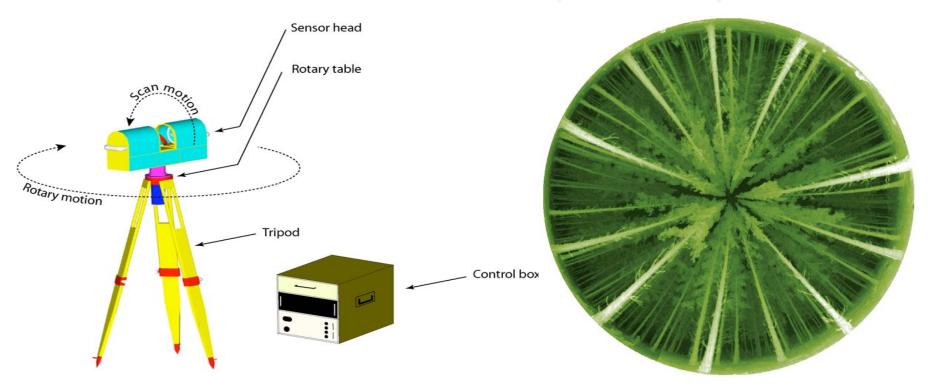




Overview

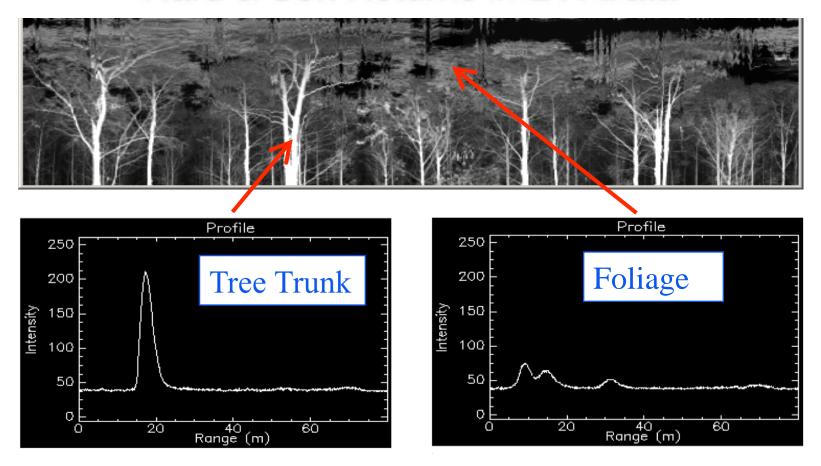
- Echidna® lidar description
- Retrieval of forest structural parameters using the Echidna® Ground-Based Lidar
 - Leaf Area Index and foliage profile
 - Canopy height
 - DBH
 - Stem count density
 - Above-ground biomass
- 3-D reconstruction of Sierra Nevada forest stands using merged point clouds from multiple Echidna® scans
 - Merging scans
 - Measuring trees in the point cloud
 - Comparison with field-measured trees
- A quick look at the new Dual Wavelength Echidna® Lidar

Ground-Based Lidar (Echidna®)



- Echidna® is ground-based lidar technology designed and patented by CSIRO specifically for forest and vegetation assessment
- The Echidna® and the current prototype the Echidna® Validation Instrument (or "EVI") has key differences from scanning rangefinders
 - Digitizes the full waveform
 - Has variable beam divergence
 - Uses full hemispherical scanning and beyond
 Wavelength 1064 nm
 - Linear response and calibration

Hard & Soft Returns in EVI Data



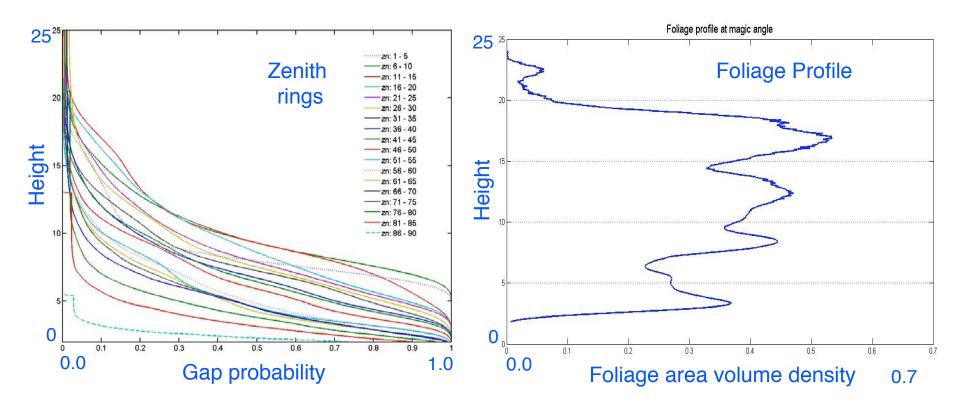
Pulse characteristics

- Length: 2.4 m (FWHM), strongly peaked
- Beam divergence: 5 milliradians (standard operation)
- Digitized every 7.5 cm on return to 140 m range

Styles of Product and Processing

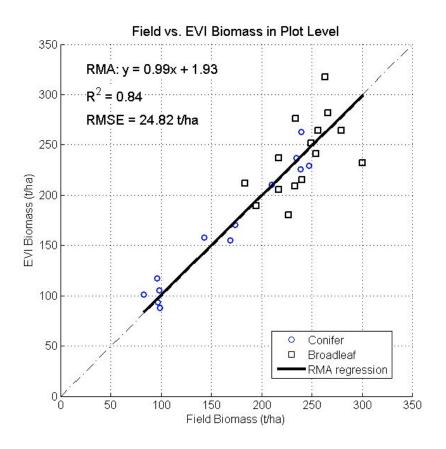
- The EVI lidar provides three types of related information about forest structure:
 - Canopy: Leaf area index, foliage profile, gap probability with height as a function of direction
 - Trunk: Trunk diameters, stem counts, and stand heights
 - Bole: Stem characteristics such as bole length, form factor (taper), curvature (sweep), and defects (CSIRO)
- These parameters allow calculation of
 - Mean DBH (diameter at breast height)
 - Stem count density (stems/ha, stems/m²)
 - Basal area (m²/ha)
 - Above-ground biomass (kg/ha) using allometric equations
 - Foliage area volume density profile (m² leaf area/m³ canopy volume)
 - Bole measurements for dimension lumber (CSIRO)
- Each one can be made easier by projecting and reformatting the data in different ways

Leaf Area Index (LAI) and Foliage Profile



- LAI and foliage profile are derived from gap probability as a function of height
- Data are averaged within zenith rings at 5° increments
- Profile is derived at "hinge angle"—zenith ring 55–60° where LAI is least sensitive to leaf angle

Biomass Estimation



- With EVI-derived mean diameter and stem count density, we can estimate biomass using allometric equations
- Since EVI can't identify species, we used a pooled allometric equation for the leading one or two dominant tree species in the plot
- $R^2 = 0.840$ at the plot level
- $R^2 = 0.975$ at the site level

Conclusions

- Gap Probability-Derived Parameters
 - Pgap with Height
 - Retrieved for 5° zenith rings
 - Foliage area volume density profile is derivative of decrease of Pgap with height
 - LAI and Foliage Profile
 - Competing methods: "Hinge angle" (57.5°) and regression
 - LAIs are somewhat different for the two methods
 - Regression method uses all zenith rings (except 0–5°) so should be more accurate
 - Foliage profiles meet expectations and knowledge of stands
 - Validation of LAI
 - Good agreement with LAI-2000, hemispherical photos, literature values
 - BU LAI-2000 and hemispherical photo retrievals not always reliable
 - Stand Height
 - Retrieved from foliage profile; matches LVIS heights very well

Conclusions, Cont.

- "Find trunks" algorithm
 - Tree diameter (DBH)
 - Retrieves individual measurements well, but with variance
 - Error depends on size and distance from EVI to tree
 - Mean DBH retrieved very well using error-weighted mean
 - Stem count density (trees/m² or trees/ha)
 - Retrieved very well
 - Requires correction for occlusion of far stems by near stems
 - Basal area
 - Retrieved well, but with slightly more error as product of two variables (mean DBH and density) measured with error
 - Biomass
 - Uses allometric equations from leading dominant species
 - Retrieved very well using mean DBH weighted by allometric exponent