Canopy structure and distribution of vegetation in **Great Smoky Mountains National Park** 

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### Great Smoky Mountains National Park (GSMNP)

- ► The GSMNP is the most visited national park in the U.S., and it hosts a rich ecosystem of plants and wildlife.
- ➤ The Park encompasses 816 sq. miles in Tennessee and North Carolina and ranges in elevation from 876 to 6,643 feet above mean sea level.
- ► GSMNP is a biodiversity hotspot. International Biosphere Reserve and a World Heritage Site





### Structure and composition of vegetatation in GSMNP

Mapping and understanding the vegetation composition and structure is important for:

- ▶ forest health management
- maintaining and tracking changes in plant and wildlife habitats and biodiversity in the park
- ► aid in the forest management planning and decisions





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Cornus florida L. mortality and understory
composition changes in western Great Smoky Mountains
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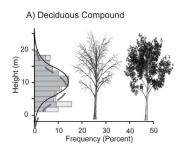
RIDGE Laboratory

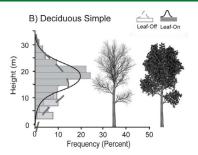
### Objectives:

- 1. Characterize three dimensional structure of the vegetation (whole vegetation canopy and understory) in GSMNP
- 2. Analyze the vegetation distribution across the park across topographic and climate gradient; structural diversity within and across various forest types
- 3. quantify biomass and productivity of GSMNP vegetation
- 4. map changes using repeating analysis over time



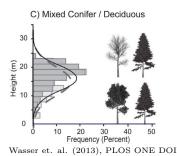
# Vegetation profile using LiDAR

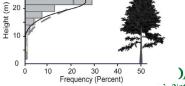




D) Conifer

30







### LiDAR data for GSMNP

#### Tennessee

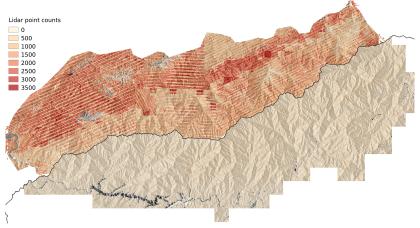
- ▶ LiDAR data for 540 sq. miles of the Tennessee portion of the GSMNP and the Foothills Parkway from 1,658 flight miles were collected during February—April 2011 by the U. of Georgia and Photo Science, Inc.
- ▶ Four multiple discrete returns per pulse were collected at a rate of 20.2 Hz from a nominal flying height of 1,981 m above ground level.
- Overlapping data were split into 724 non-overlapping 1,500 × 1,500 m tiles, which we obtained from the National Park Service.
- ▶ 724 LiDAR tiles (approx. size 98 GB) projected onto a 3.0 m resolution digital elevation model (DEM) derived from the LiDAR point cloud.
- ▶ Projection: UTM Units: meters

#### North Carolina

- LiDAR data for North Carolina was collected by NC Floodplain Mapping Program in 2005.
- $\blacktriangleright$  Overlapping data were split into non-overlapping  $10,000\times10,000$  ft tiles, which we obtained from the NC Floodplain Mapping Program.
- ▶ 184 LiDAR tiles (approx. size 8.9 GB) projected onto a 3.0 m resolution digital elevation model (DEM) derived from the LiDAR point cloud.
- ▶ Projection: NC State Plane Units: ft



# LiDAR quality across GSMNP

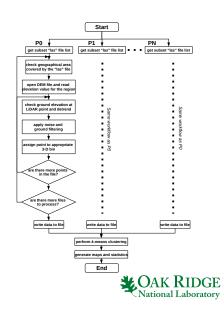


Data from NC vs TN side of park have large disparity in the quality, making integrated analysis across the park a challenge.

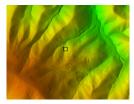


# Computational Workflow and Data Processing

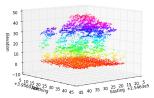
- We employed a process-parallel approach to extract and analyze LiDAR point cloud data using python.
- ➤ To estimate vegetation heights above ground level, elevations from the 3.0 m DEM were subtracted from point cloud data.
- ► The resulting points were grouped into 1 m vertical bins, up to 75 m, at a horizontal resolution of  $30 \times 30$  m.
- Anomalous high points (aerosols, birds) and low points (steep slopes, surface litter) were filtered out.
- ► Corrections were made for low height vegetation (shrubs and grasses) and for many returns at the same elevation.



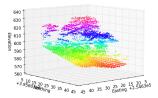
# LiDAR Point Cloud Example: 30m pixel



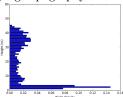
a) 3-D LiDAR point cloud extent at  $30 \times 30$  m (black square) shown in a typical GSMNP cove forest.



c) LiDAR point cloud after topographic detrending and filtering (3,936 points).



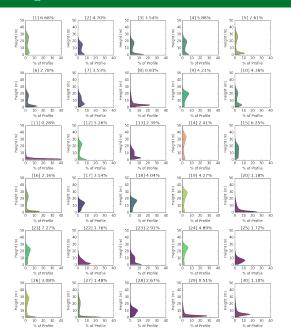
b) Raw LiDAR point cloud (3,985 points), showing imprints of underlying topography.



d) Vertical distribution of LiDAR point density in a cove forest dominated by tall trees and a dense understory.

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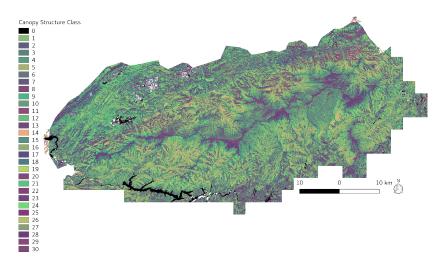
### 30 vegetation structure classes



We use a similarity color scheme based in Principal Component Analysis: Low (< 20m) canopy (RED); Medium (30 - 36 m) canopy (BLUE); Tall (> 46m) canopy (GREEN)

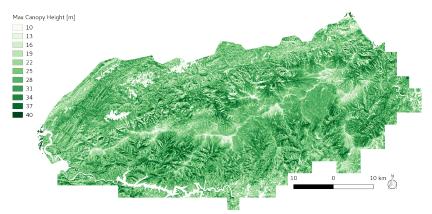


### Vegetation of 30 vegetation structure classes across $\operatorname{GSMNP}$





# Maximum canopy heights across GSMNP



95th percentile height of vegetation canopy

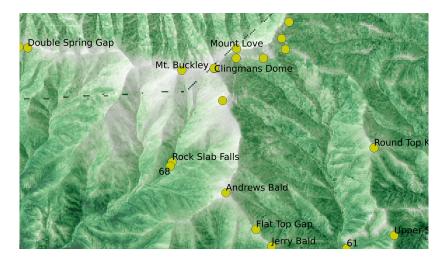


# Clingman's Dome: Google Earth



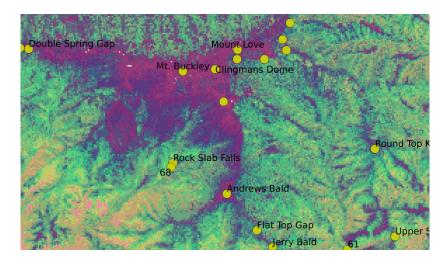


### Clingman's Dome: Max. canopy height





### Clingman's Dome: Vegetation structure





### Cades Cove: Google Earth



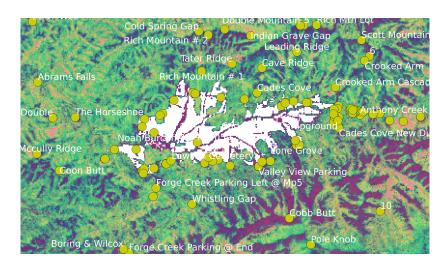


### Cades Cove: Max. canopy height





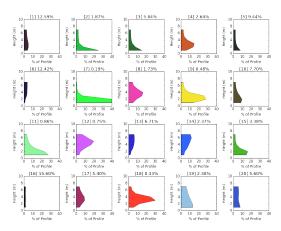
### Cades Cove: Vegetation structure





### What does the understory vegetation look like?

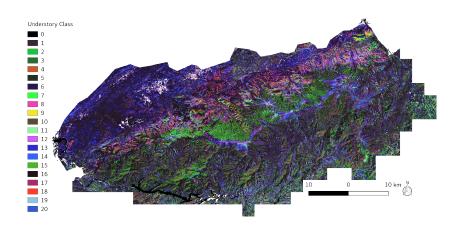
- ► Vegetation structure is however fairly dominated by the taller forest canopies.
- ► To better characterize and map the understory vegetation we performed a second set of analysis by using the LiDAR point cloud data set only up to 8mheight from the ground.



We use a similarity color scheme based in Principal
Component Analysis: Low (< 3m) canopy
(GREEN); Medium (3 - 5 m) canopy (RED); Tall
(5 - 8m) canopy (BLUE)

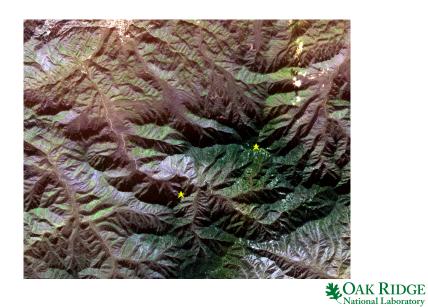
Validational Laboratory

### Understory vegetation map

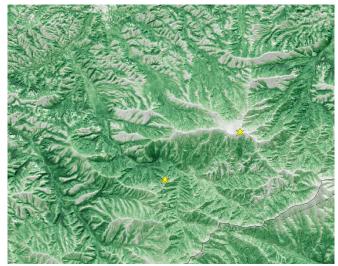




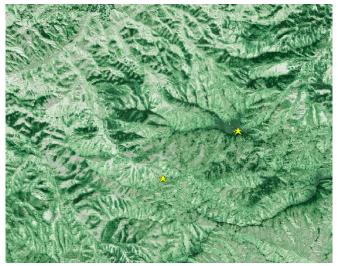
### Mt. Leconte / Chimney Tops: Rapideye 12/29/2015



### Mt. Leconte / Chimney Tops: Max. canopy height



### Mt. Leconte / Chimney Tops: Amount of understory vegetation



Map shows the fraction of total tree canopy in the understory. High elevation areas in the park have low height trees with a dense suggest RIDGE vegetation (dominated by Rhododendron and Mountain Laturational Laboratory

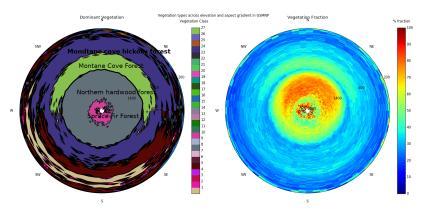
### Understanding the patterns across the park

Putting things in landscape scale perspective to understand the vegetation across the park.

Virtual Mountain: We represent the variability across the topography using a Virtual Mountain plot that captures the distribution of data across elevation and aspect gradient for the entire park. Radial direction represents elevation with highest elevation at the center to lower elvations in radially outward direction. Azimuth direction represent the topographic aspect on the landscape.



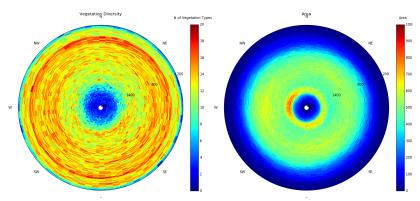
### Vegetation across complex topography within GSMNP



Climate and topographic gradients across the park supports diverse forest types, which show preferential distribution and dominance along these gradients.



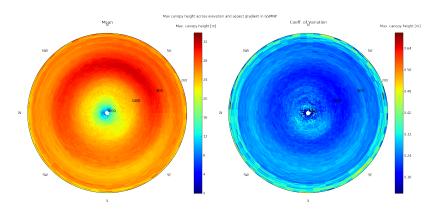
### Vegetation diversity within the park



Higher vegetation diversity at low to mid elevations, while lower diversity at high elevations



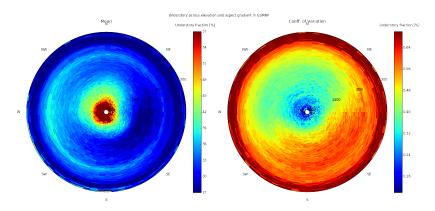
### Patterns of tree heights



Forest canopies are taller at low to mid elevations, and shorter vegetation at higher elevations



## Pattern of understory vegetation



Dense understory and low height vegetations are dominant at higher elevations in the park



# Summary and Conclusions

- ▶ We developed an approach, parallel software tools, and workflow for analyzing large volumes of LiDAR point cloud data in a scalable fashion.
- ▶ Multivariate Spatiotemporal Clustering (MSTC) provides a valuable quantitative framework for stratifying vegetation canopy structure data derived from LiDAR point clouds.
- ► We applied these tools to LiDAR data from the GSMNP to identify vegetation classes based on overstory/understory distributions.
- ► Early validation efforts for various areas in the park are promising. We are also comparing with the available plot observations from the park.
- ► These tools and the resulting maps will inform resource management and conservation planning by forest and wildlife managers, who were not previously able to use large, complex LiDAR data sets.

### Free and Open Source Software for Open Science

#### Data Source:

USGS, National Park Service, US Forest Service

#### LiDAR Data Processing:

Custom developed Python software using laspy module for processing las files. multiprocessing module was used to implement shared memory parallelism.

Unsupervised classification software was developed in C using MPI for distributed memory parallelism.

### Geospatial analysis and visualization

All geospatial analysis and visualization was done using GDAL, GRASS GIS, QGIS, and Python



#### Computational resources

All data analysis were conducted on compute clusters at ORNL running Fedora/CentOS Linux.

#### Document preparation

All documents/presentations/posters were prepared using LATEX and beamer.

#### Publications, data and software products

All publications, data and software from this research are available publicly.



#### Characterization and classification of vegetation canopy structure and distribution within the Great Smoky Mountains National Park using LiDAR

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Abstract-Vegetation canopy structure is a critically important habitat characteristic for many threatened and endangered birds and other animal species, and it is key information needed by forest and wildlife managers for monitoring and managing forest resources, conservation planning and fostering biodiversity Advances in Light Detection and Ranging (LiDAR) technologies have enabled remote sensing-based studies of veretation canonies by capturing three-dimensional structures, yielding information not available in two-dimensional images of the landscape provided by traditional multi-spectral remote sensing platforms. However, the large volume data sets produced by airborne LiDAR instruments pose a significant computational challenge, requiring algorithms to identify and analyze natterns of interest buried within LiDAR point clouds in a computationally efficient manner, utilizing state-of-art computing infrastructure. We developed and applied a computationally efficient approach to analyze a large volume of LiDAR data and characterized the veretation canony structures for 139,859 bectares (540 sq. miles) in the Great Smoky Mountains National Park. This study helps improve our understanding of the distribution of vegetation and animal habitats in this extremely diverse ecosystem.

#### I. INTRODUCTION

Forcet coxystems are a complex monaix of diverse plan and tree species, the Racion and distribution of which are distributed or which are representation regimes, to propagable re-re-presentation regimes, to propagable re-re-presentation regimes, to propagable re-re-presentation regimes, to propagable re-re-presentation regimes, and a resultant respective resultant respective respective resultant resultant respective resultant result

Remote sensing has been widely used to monitor regional to

global forest ecosystems and for mapping of vegetation types.

However, traditional remote sensing methods for vegetation classification often use light reflectance from the top layer of vegetation. Advances in Light Detection and Ranging LL-Disk (schadingles have enabled ermore sensing-board under not vegetation canopies by providing a three-distrainable set of vegetation canopies by providing a three-distrainable with the contract of the contract o

The objective of this study is to develop methods to realize the potentials of rich LIDAR data set to map and characterize the three-dimensional structure and distribution of vegetation canopies. We develop and apply data analytic techniques to identify the cologically important and understandable structural types by mining the large and complex volumes of LIDAR data.

II. MATERIALS

#### Study area

The goographic area for this study was the Great Ensoly Mouriants. Niclosed Javid (SOMNF), which is part covers the Great Soulcy Mountains and the Bline Edge Mountains of the Great Soulce of Soulc

Kumar, J., J. Weiner, W.W. Hargrove, S.P. Norman, F.M. Hoffman, and D. Newcomb. 2015 LiDAR-derived Vegetation Canopy Structure, Great Smoky Mountains National Park, 2011. ORNL DAAC, Oak Ridge, Tennessee, USA. http://dx.doi.org/10.3334/ORNLDAAC/1286



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