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Temporal Analysis of Field, SSURGO, and LiDAR Derived Site Indices in the Southeastern U.S.

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TEMPORAL ANALYSIS OF FIELD, SSURGO, AND LIDAR DERIVEDSITE INDICES IN THE SOUTHEASTERN U.S.

A Thesis Presented to the Graduate School of Clemson University

In Partial Fulfillment of the Requirements for the Degree Master of Science Forest Resources

> by Steven A. Ham May 2013

Accepted by: Dr. Elena Mikhailova, Committee Chair Dr. Lawrence Gering Dr. Christopher Post

ABSTRACT

Sustainable forest management requires accurate information about site index (SI, tree height at a base age). The objectives of this study were to compare site indices from field inventory data (2008-2009), Soil Survey Geographic Database (SSURGO), and Light Detection and Ranging (LiDAR, 2008), and to determine the uncertainty in the site indices from the southern part of the Clemson Experimental Forest. When LiDAR derived analysis were used to compare to SSURGO there were statistical differences for site indices for all of the tree species in this study: loblolly pine (*Pinus taeda*), scarlet oak (*Quercus coccinea*), shortleaf pine (*Pinus echinata*), white oak (*Quercus alba*), and yellow poplar (*Liriodendron tulipifera*). LiDAR has the potential to provide reliable and rapid estimates of site index variability within the soil map units. Loblolly pine and shortleaf pine had the greatest statistical differences with the LiDAR derived site indices being much larger than the SSURGO values. The results of this study indicate that a larger sample size for LiDAR is a better option to decrease variation, and that the map unit level may be the best option.

Keywords and abbreviations: basal area factor (BAF), Clemson Experimental Forest (CEF), climate change, diameter at breast height (DBH), forest inventory, Light Detection and Ranging (LiDAR), loblolly pine, site index (SI), soil inventory, Soil Survey Geographic Database (SSURGO)

DEDICATION

I would like to dedicate this thesis to Steve Ham, Sherry Dalton, Larry Grimes, and Sheila Fernandez. Without their support and help along the way, I would not have been able to get through college, or be the person that I am today.

ACKNOWLEDGMENTS

I want to express my sincere appreciation to my committee: Dr. Elena Mikhailova, Dr. Lawrence Gering, and Dr. Christopher Post for their help and support throughout this study. I wish to thank and acknowledge Benjamin Kendall and Knight Cox for providing the forest inventory cruise data for the Clemson Experimental Forest, the Anderson County Assessor's Office and the South Carolina Department of Natural Resources for the LiDAR datasets, and Dr. William Bridges for the help with the statistics. Financial support for this project was provided by Clemson University. Technical Contribution No. 6069 of the Clemson University Experiment Station.

TABLE OF CONTENTS

Table of Contents (Continued)

Page

LIST OF TABLES

List of Tables (Continued)

List of Tables (Continued)

Table Page

List of Tables (Continued)

LIST OF FIGURES

List of Figures (Continued)

CHAPTER ONE INTRODUCTION

Statement of the Problem

Soil quality is the most important factor in assessing forest productivity and forest management decisions (Hamilton, 2007). Concerns about the effects of global climate change have triggered intensive research over recent decades on the impacts of land use on biomass production in forest ecosystems (Poudel et al., 2011). The temperature increase in higher latitudes associated with global warming may stimulate forest production due to longer growing season and more favorable conditions for photosynthesis (Bergh et al., 2003; Poudel et al., 2011). Several studies predicted higher forest productivity (by 10-30% in the next 100 years), reduced length of rotation periods by 5-10 years, and increased carbon stocks (Pussinen et al., 2002; Kirilenko and Sedjo, 2007; Eggers et al., 2008; Subedi et al., 2009). Current research on climate change over the continental United States using a moisture index (Grundstein, 2009) showed that the southeastern part of the country has been getting wetter, and cooler. The question, then, is how do these combined changes affect forest productivity and site index in the Southeastern U.S.?

Literature Review

Site index (SI) is defined as the total height to which dominant trees of a given species will grow on a given site at some index age, usually 50 years in the Southeast (Hamilton, 2007). Site index is commonly used to evaluate site productivity and is

provided through the SSURGO (Soil Survey Staff, 2013). The site index can be determined by two methods: 1) from the site index curves using accurate age and height measurements or 2) from the tables in the soil survey (Baker and Langdon, 2012). For example, site index for loblolly pine at age 50 can vary from 60 to 120 feet depending on soil and other site characteristics. Site index curves were developed in various time periods ranging from early 1920's to late 1980's (National Register of Site Index, 2013). Site index does not provide the measure of uncertainty due to soil, site, and climate variability.

South Carolina has diverse soil coverage. Seven of twelve soil orders occur in South Carolina: Histosols, Entisols, Inceptisols, Mollisols, Alfisols, Spodosols, and Ultisols. Lynchburg is a state soil of South Carolina (it belongs to the order of Ultisols) because it is a prime farmland and has a high site index for loblolly pine. Approximately four hundred soil series (of 21,000 soil series) occur in South Carolina. In the Southeastern U.S., soils in the Piedmont physiographic province are severely eroded and contain low quantities of soil C as a result of agricultural activity prior to 1940 (Dunn and Holladay, 1977; Richter et al., 1999). Conversion of degraded croplands to forests likely has increased terrestrial carbon stocks (Richter et al., 1999), although the magnitude of this increase and the impact of forest management on it are unclear because of the lack of long-term data and the high variability in existing soil and plant data. Impacts of land use and climate on site index are poorly understood. A previous study in the Pacific Northwest indicates LiDAR derived site indices correlate well with field site indices (Gatziolis, 2007).

2

Research Questions

- How reliable is the site index provided by SSURGO?
- Are site indices developed in the 1920's valid today?
- What are the ways to incorporate variability and uncertainty in the site index?
- Does climate change affect the site index in the SE of the United States?
- Can advanced techniques (e.g. SSURGO etc.) be used to evaluate and update site index in the SSURGO as needed?

Research Objectives

- Compare site indices from SSURGO to the most recent forest inventory cruise data;
- Compare site indices from SSURGO to LiDAR derived site indices;
- Conduct statistical analysis for any trends and to determine the uncertainty in the site indices.

CHAPTER TWO MATERIALS AND METHODS

Study Area

The location for this study is the southern part, Anderson County (Figure 1) of the Clemson University Experimental Forest (CEF) in Clemson, SC (N 34° 41' 55.7", W 82° 52' 45.7") (Griffith et al., 2002). The CEF lies in the upper Piedmont region at the base of the foothills. The climate is generally mild and has four distinct seasons with annual precipitation averaging 54.01 inches with annual temperatures ranging from an average minimum temperature of 29.80°F in January to an average maximum annual temperature of 90.6°F in July (IDcide 2013). The CEF is divided into 15 divisions, which are subdivided by compartment and then broken down to stand level. In total, the forest has more than 2,000 stands and 41 of them were used in this study (Tables 15-19). The entire forest is an intensively managed multi-use forest with research areas, natural areas, timber production areas, and wildlife managed areas. The cover type ranges from hardwood coves, upland hardwoods, mixed hardwood and pine, natural pines, to pine plantations (Forest inventory 2008-2009). The data for the study area were limited to the forest area in Anderson County. Attention was given to points that had field forest inventory (2008-2009), SSURGO, and LiDAR data (Table 1).

History of Ownership and Management

The CEF was previously eroded farm land before Clemson College began supervision of the lands in 1939 under an agreement with the federal government (Dunn and Holladay, 1977). In 1946 a forester, Norbert Goebel, was hired to manage the forestlands (Dunn and Holladay, 1977). Silvicultural practices (planting, thinning and harvests) to improve the timber production, wildlife habitat and water quality were initiated. In 1954, the project was deeded to Clemson College, due to the efforts of U.S. Senators Charlie Daniel, Strom Thurmond, State Senator Edgar Brown and Dr. George H. Aull (Dunn and Holladay, 1977).

Forest Inventory

The cruise data were gathered from the forest inventory in 2008-2009 according to field forest specification for the CEF (Table 2). It was inventoried as point-level data measured in cubic feet. The point location was determined by a resource grade GPS and previous point locations on the forest. Each point was measured on a tree level basis. At each point trees were tallied using a BAF (basal area factor) = 10 prism. For each tally tree species, diameter at breast height (DBH) (1 inch class), type of product, merchantable height, total height, the percent defect, and any comments about the tree were recorded. The trees were measured with a diameter tape and mechanical clinometer. Height measurements were obtained to the nearest foot and diameter DBH to the nearest 1-inch class (Table 2). The contractor provided the tree data for volume determinations. The points are numbered and correspond to a GPS location. Stand information was also

available for the entire Clemson Experimental Forest as GIS and Excel spreadsheets. The amount of volume harvested for each stand and the date the stand was cut was also available. This allowed for any points to be nullified if stands were harvested after the 2008 LiDAR data were flown, or before the 2011 LiDAR data were flown.

Soil Inventory and Site Index from SSURGO

Soil Survey Geographic Database (SSURGO) tabular and spatial complete data were downloaded from http://soils.usda.gov/survey/geography/ssurgo/ for Anderson County, South Carolina. Site index is listed for various tree species at base age 25 or 50. The field inventory data were collected at base age 50. Therefore, this base age was used throughout the analysis. The forest inventory GIS data were used to spatially compare the SSURGO data to the study area (Figure 2). First, all the stands in the study area within Anderson County were selected using a county outline from ArcGIS and exported the file into NAD83 UTM Zone 17N meters. A "select by location" using the county outline as the source layer and selected the stands that are within the source layer was performed. The selection was then exported as a new layer. The spatial SSURGO map unit data layer was joined to the component key table by the map unit key in both attribute tables. Then the soil map unit table was joined to the forest productivity component table by the component key. The ACCESS database associated with the SSURGO data were then used and a report for site index was created and exported as an excel file. The Excel file was then edited to create one row for each map unit. The excel file was then exported to the project geodatabase as a single table and joined by the map unit symbol to the spatial

6

SSURGO data. Each Map unit symbol contained its own site indices for various tree species and was compared to the inventory data and LiDAR derived measurements.

Point to Plot Conversion

In order to be able to compare LiDAR to field data the point data (collected with a basal area factor (BAF) = 10 ft²/ac prism) had to be converted to fixed radius plot data. The first step was to calculate the basal area (BA) for each inventoried tree. This was found using equation (1):

$$
BAeach tree = 0.005454 * (DBHclass)2
$$
 (1)

where $BA_{\text{each tree}}$ is calculated in (ft²/tree) and represents the per tree basal area for the point; and DBH_{class} (in) is the 1-inch diameter at base height (DBH) class for each tree on the point. Once $BA_{each tree}$ was calculated, the tree factor for each individual tree could be calculated. The equation for this calculation is:

$$
F_t = \frac{F}{BA_{\text{each tree}}} \tag{2}
$$

where \mathbf{F}_{t} is calculated in (Number of trees/ac) and represents the tree factor for each tree, (ft²/ac) is the BAF represented by the prism, and $BA_{\text{each tree}}$ is measured in (ft²/tree). Once \bf{F} was calculated for each tree, the total number of trees per acre for the entire point was calculated. This was found using the equation:

$$
\frac{\text{Trees}}{\text{acre}} = \sum F_t \text{ each tree on the point}
$$
 (3)

where $\frac{Tree}{area}$ is calculated in (trees/ac) and represents the total number of trees per acre on a particular point, and \mathbf{F}_{t} is measured in (trees/ac). The next step was to calculate the total basal area per acre for the point. This was found calculated using the equation:

$$
\frac{BA}{\text{acre}} = F * \text{Number trees tallied on the point} \tag{4}
$$

where $\frac{BA}{2}$ represents the basal area per acre for the entire plot in (number of trees*ft²/ac), and F (ft²/ac) is the BAF of the prism used. With the $\frac{BA}{\epsilon}$ calculated, the mean basal area for the mean tree on the plot was then calculated using the equation below:

$$
BA_{\text{mean}} = \frac{\frac{BA}{\text{area}}}{\frac{\text{area}}{\text{area}}}
$$
 (5)

where BA_{mean} represents the basal area for the mean tree of a particular point in (ft²/tree), and $\frac{BA}{2}$ is measured in (number of trees*ft²/ac), and $\frac{Tree}{2}$ is measured in (number of trees/ac). After the basal area for the mean tree, calculate the DBH for the mean tree in the equation below:

$$
DBH_{\text{mean}} = \sqrt{\frac{BA_{\text{mean}}}{0.005454}}
$$
 (6)

where DBH_{mean} represents the DBH for the mean tree of the point in (in), and BA_{mean} is measured in (ft^2 /tree). The mean DBH was then used to calculate the radius for the plot in the equation below:

$$
R_{\text{mean}} = \frac{DBH_{\text{mean}}}{\left[12 \times \left(\frac{1}{ss}\right)\right]}
$$
 (7)

Where R_{mean} represents the mean radius of the plot equivalent for each individual point centered at the location of the variable radius plot in (ft), and DBH_{mean} is measured in inches. The R_{mean} (ft) is then used to get the area represented by each plot in the equation below:

$$
A_{\text{mean}} = \pi R_{\text{mean}}^2 \tag{8}
$$

where A_{mean} represents the average area of the plot for the average tree on the point in (ft²), and \bf{R}_{mean} is measured in (ft). All of the point-to-plot conversions were done in Excel.

Remote Sensing Data

Two LiDAR datasets flown over Anderson County, SC in 2007 and 2011 were used in this study and were collected by the South Carolina LiDAR Consortium. The first dataset (2007) was obtained from the South Carolina Department of Natural Resources, and is single return data with nominal point spacing of approximately 1.4m and was processed using Terrascan software. The second LiDAR dataset (2011) was obtained from Anderson County, SC Assessors office and had a nominal point spacing of 1.4m. This data were collected by Towill Inc. using an Optech Orion M-200 LiDAR system. The data were originally collected as waveform data and discrete pulses were extracted and provided in the dataset as up to 4 discrete returns per LiDAR pulse. Data were processed with Terrascan as well as the Terrasolid software suite.

Site Index Inventory by LiDAR

Newly flown, multi-return, LiDAR data are available for Pickens (5/2012), Anderson (8/2012) and Oconee counties (8/2012).

Anderson County LiDAR data from 2008 are already available (Figure 3). Tree locations, heights and crown diameters were extracted from the LiDAR data using the TIFFS software package version 8.0 beta (Globalidar). This software uses the raw LAS files and tiles the raw data and filters out ground and non-ground returns. The filtered data are then used to derive a digital elevation model and canopy height model, which it subsequently uses to identify tree locations using watershed analysis techniques within Matlab based on individual tree crown structures (Chen, 2007).

The point-to-plot conversion radii were put into an Excel file and exported as a dbase file and added to ArcMap. The forest inventory spatial data were also added to ArcMap. Stands in Anderson County were added if they contained relevant information for determining LiDAR site index. The CEF points file are points corresponding with the stand data of the stands that they are in. The attribute table was joined to the Excel file based on the recognized number for the points in the CEF data and the point number in the dbase file with only matching records being kept. The new attribute table was exported to create a new shapefile. A buffer was then applied to create a polygon file with the plot radii from the point shapefile's radius (m) as the buffering distance.

LiDAR tree locations (points) were then added to ArcMap and extracted where only those trees contained inside the plots were kept. The tree locations are points that contain information on individual tree heights. Once the trees were extracted the next

10

thing was to create an excel spreadsheet summarizing the LiDAR heights in each point and corresponding the age of the trees on the point to plot conversion and the SI species with the stand data. The summary sheet was then used to choose the proper SI species curves (Figures 4-8) to derive site indices for each point. Upland oaks (Figure 6) site index curves were used for scarlet oak.

The tallest tree on each point and the stand age were used to derive an SI from the SI Curves (interpolated/extrapolated between two curves) and added to the Excel sheet. The assumption of tree selection is that the tallest tree(s) will be the same specie as the SI specie of the stand from the CEF data. The mean and standard deviation on each LiDAR (2008) point were summarized by map unit and SI species. The procedures were repeated for the LiDAR (2011) dataset with the stand ages adjusted to the proper age. An observation was defined as a single point. The total number of points for a particular specie that were grouped together by matching criteria by map unit, stand number, or a combination of the two defined the sample size for the particular criteria.

After the plot data were summarized they were then summarized for entire map units in a similar manner. Stands were selected by their SI species from the point summary Excel file and exported as a new file. SSURGO soils map units were extracted by SI species from the point summary excel file. Each map unit within a stand was exported as an individual shapefile. Both LiDAR (2008) and LiDAR (2011) tree data were extracted by each individual map unit layer within a stand. Any stand that had trees from multiple files were selected and then merged together and exported as one file to properly select the correct amount of trees from each file. The sample from each stand

11

was calculated as the tallest 5% of all trees, or the tallest 20, whichever was greatest. The stand age was then gathered in the same manner by the corresponding stand identification number. The height was then converted to a SI the same way as the plot data. This data was then added to its own excel file summarized by mean SI and standard deviation for each species in three ways: each stand number as its own occurrence in a map unit, each map unit within a stand, and by each map unit.

Statistical Analysis

The statistical analysis proceeded in three stages. The first stage was to calculate means and standard deviations of site indices of each species broken down by soil map unit for both plot level and soil map unit level and further break the map unit level data down by stand. The second stage was to compare the means to the SSURGO standard using a series of t-tests. The third stage was to convert R^2 -values to t-values and compare using a series of t-tests. The reason for using the t-test is that it provided a direct test of the hypothesis that the mean value was different from the SSURGO in whichever direction that occurs. All calculations were performed in Excel.

CHAPTER THREE RESULTS AND DISCUSSION

Soil Evaluation

Sixteen soil map units (soil order: Ultisols) were identified (Table 3). Map units CdB and CdC had the highest SI for loblolly pine (83 ft, base age 50). The lowest SI for loblolly pine was HwC2 (71 ft, base age 50). CdD had the highest SI for scarlet oak (81 ft, base age 50). The lowest SI for scarlet oak were map units MaC and MaE both had the lowest SI for scarlet oak (75 ft, base age 50). Map unit PaE had the highest site index for shortleaf pine (70 ft, base age 50). The lowest SI for shortleaf pine was map unit CcC2 (55 ft, base age 50). Map units MaD and MaE had the same SI for white oak (75 ft, base age 50). Map unit MaE had the highest SI for yellow poplar (96 ft, base age 50). The lowest SI for yellow poplar was map unit PaE (90 ft, base age 50).

Forest Inventory vs. SSURGO

Loblolly pine and shortleaf pine species were the only SI species at the forest inventory level that significantly differed from the SSURGO site indices. Ten out of the thirty one map unit and specie combinations were statistically different than the SSURGO site indices with 8 being higher. All the loblolly pine map units that were different were higher, and shortleaf pine was split with 2 being higher and 2 being lower (Tables 4A and 4B).

LiDAR Derived Plot Level vs. SSURGO

Almost half of the plot level LiDAR derived site indices were statistically different than the SSURGO SI (28 out of 57 total observations). The majority of the 2008 LiDAR plot map unit derived site indices were statistically different than the SSURGO site indices (16 out of 31 total observations) and occurred in each soil series. Ten out of the sixteen map unit symbols' species combinations LiDAR derived site indices were statistically lower than the SSURGO site indices occurring in all five species. The six map unit symbols that were higher occurred in loblolly pine (5 map unit symbols) and shortleaf pine (1 map unit symbol).

The 2011 LiDAR derived plot site indices were statistically different than the SSURGO site indices for all five species. The majority of the derived site indices that were statistically higher occurred mainly in loblolly pine species. Two shortleaf pine map unit symbol site indices were statistically lower with none being higher (Tables 4A and 4B) which was different than 2008. The high variability associated with using plot data can be seen in Figure 9. The variability could be caused from the small sample sizes in the plot data. Another contributing factor on the graph is all the species and all the map units are represented on the graph together.

LiDAR Derived Map Unit Level vs SSURGO

The majority of the standard deviations of the map unit level data were lower than the plot level data (Tables 4A and 4B). With the higher variability and small sample sizes in the plot level data it was necessary to take another approach and to use map unit level data which had greater sample sizes. The majority of the map unit level LiDAR derived mean site indices for the map unit symbols (2008, 2011, and 2008 and 2011 combined) were statistically higher than the SSURGO site indices and occurred in all five species across each soil series (Tables 4A and 4B). When looking at Tables 4A and 4B there were 6 combinations where LiDAR derived mean site indices were statistically lower than the SSURGO site indices. These occurred in loblolly pine (CbC and HaB), scarlet oak (MaE), and yellow poplar (CdC).

The site indices by map units were further broken down by stand within a map unit for further analysis. When loblolly SI was broken down by stand within a map unit the majority of the derived mean site indices are statistically higher than the SSURGO site indices except for map units CbC, CdC, HaB, and MaC (Tables 5A and 5B). This was different than what was summarized from Tables 4A and 4B, that CbC and HaB were the only two map units where the LiDAR derived mean site indices were statistically lower than SSURGO site indices.

Scarlet oak did not pick up any more map unit symbols that had observations indicating a statistically lower mean SI than when the map units were broken down into stands (Tables 6A and 6B). On the summary tables (Tables 4A and 4B) shortleaf pine did not have any map unit symbols with statistically lower LiDAR derived mean site indices than SSURGO site indices, but when broken down by stand within a map unit 4 map unit symbol LiDAR derived site indices (CbB, CbC, CdC, and CdD) were significantly lower than SSURGO site indices (Tables 7A and 7B).

15

Tables 4A, 4B, 8A, and 8B show white oak did not have any map unit symbols indicating that LiDAR derived mean site indices were statistically lower than SSURGO site indices. Yellow poplar map unit symbols broke down into stands indicated that one other LiDAR derived map unit symbol's SI was statistically lower than the SSURGO (CdD). Tables 10-14 (A and B for each) are intermediate tables showing where the numbers for Tables 4A and 4B came from. Each table is a single species.

Figure 10 shows the high variability in the relationship between SSURGO and forest inventory site indices. The variability is likely present due to the graph showing all species and all map units combined where site index is based on one tree species for one map unit where tree species have different growth patterns. Another possible contributing factor is that both are reported as one single number without a standard deviation.

There are a couple trends present. One trend is that LiDAR derived maximum heights (2008) are significantly higher than the forest inventory measured maximum heights. The model equation shows that in 2008 the LiDAR and forest inventory heights correlate well (Figure 11). Another trend is LiDAR derived maximum heights (2011) are significantly higher than the LiDAR derived maximum heights (2008) indicating growth has occurred (Figure 12). Tables 5B, 7B, and 9B show that some stands had negative growth and indicate a stand has had one of the following or combination of the following occur: the stand has been harvested, trees were lost to some sort of event, an overestimation of LiDAR derived tree heights in 2008, or an underestimation of LiDAR derived tree heights in 2011 has occurred. A contributing factor to the variability is more than likely the different growth patterns of each species (Figures 9-12).

CHAPTER FOUR

CONCLUSIONS

The objectives of this research were to compare site indices from SSURGO to the most recent forest inventory cruise data, compare site indices from SSURGO to LiDAR derived site indices, and conduct statistical analysis for any trends and to determine the uncertainty in the site indices. Site index is supposed to be static and not change. Tables 4A and 4B show the only two species that had a difference were loblolly and shortleaf pine. However, when LiDAR derived analysis was used to compare to SSURGO there were differences. There were statistical differences for site indices for all of the tree species in this study: loblolly pine, scarlet oak, shortleaf pine, white oak, and yellow poplar. This raises the question what methods can produce reliable and rapid estimates of site index?

LiDAR has the potential to provide reliable and rapid estimates of site index variability within the soil map unit (Tables 4A and 4B) unlike the field forest inventory data which provides limited and expensive data. Newly available free LiDAR data sets provide an opportunity to evaluate variability across thousands of soil map units and tree species combinations. Through the LiDAR analysis all the tree species had statistically different site indices than that recorded in SSURGO. The only way to accurately and cheaply answer the question if environmental and anthropogenic changes affect the site index in the SE of the United States is to determine tree heights and the associated

17

variability across large spatial areas (with numerous soil map units and forest characteristics).

The only way to update and evaluate the site index is to have sufficient forest data which is possible through LiDAR analysis. Advanced genetics, different silvicultural practices, recent droughts, warmer temperatures, and other anthropogenic and environmental changes may be having an effect on SI. Even though SI is static, this study indicates there are statistical differences in SI from the LiDAR data which means that through environmental and anthropogenic changes, SI may be changing and more research is needed to understand the effects of these changes. The results of this study indicate that a larger sample size for LiDAR is a better option to decrease variation, and that the map unit level may be the best option.

There are some errors that arise using LiDAR. Ground sampling errors can occur due to dense closed canopy forests. Underestimation of tree height can also occur due to tree lean or when LiDAR returns capture the wrong tree apex. These errors are usually small (Gatziolis, 2007).

Figure 1. Map of the part of the Clemson Experimental Forest.

Figure 2. Process of extracting soil data for the southern part (Anderson County) of the Clemson Experimental Forest.

Figure 3. Clemson Experimental Forest, South Forest, Anderson County LiDAR tree returns(2008).

LOBLOLLY PINE

SOURCE: VOLUME, YIELD & STAND TABLES FOR SECOND-GROWTH SOUTHERN PINES, U.S.D.A. MISC PUBL. NO. 50 (1929). REVISED BY COILE & SCHUMACHER, JOURNAL OF FORESTRY, JUNE 1953.

Figure 4. Loblolly pine Site Index classes for coastal and piedmont areas of the Southeastern U.S. (adapted from Coile & Schumacher, 1953).

SHORTLEAF PINE

Figure 5. Shortleaf pine Site Index classes in natural range (adapted from

Coile & Schumacher, 1953).
UPLAND OAKS

D.J. OLSON, JR., S.E. F.E.S. RESEARCH NOTES NO. 125 APRIL 1959.

Olson, 1959).

WHITE OAKS

SOURCE: BASED ON DATA FROM OLSON, D.J. S.E.F.E.S. RES. NOTES NO. 125, 1959

Figure 7. White oaks Site Index lasses in the Southeastern U.S. (Olson 1959).

YELLOW POPLAR

McCARTHY, E.F., U.S.D.A. TECH. BULL. NO. 356, 1933.

Figure 8. Yellow polar Site Index classes for coastal and Piedmont areas, in

natural range outside mountain areas (Beck, 1962).

Table 1. Spatial data sources and descriptions used in this study.

⁾All data layers projected to Universal Transverse Mercator Zone 17 North (UTM Zone 17 N), North American Datum (NAD) 1983.

Table 2. Field forest inventory specifications for Clemson Experimental Forest.

Basal Factor 10 prism points to be taken in approximately 273 stands on the Clemson Experimental Forest for a total of approximately 2000 points.

Stand Maps and x, y coordinate locations for point locations in each stand to be sampled will be provided. The inventory contractor will be required to navigate to the x, y coordinate location to within 30 feet of the given position, place a flag with the point number at the point center and collect inventory data as described below.

__

Data required on each point will include:

- 1. **DATE AND CREW**
- 2. **STAND ID**.
- 3. **POINT NUMBER**
- 4. **LATITUDE AND LONGITUDE of POINT**
- 5. **COVERTYPE (general covertype represented by point, from list)**
- 6. **TOPOGRAPHIC POSITION of POINT (bottom, cove, lower middle, upper, ridge)**

7. **REPRODUCTION** on point. (# stems < 0.6 inch DBH. on .001 acre (3.7 ft. radius) circular plot by species group (pine, oak, yellow poplar, other).

The following information must be recorded for each tree determined to be in on the BAF 10 sampling point.

8. **TREE NUMBER** for each tree tallied on point.

9. **SPECIES** of each tree tallied. A species list with corresponding species code will be provided.

10. **DBH** of all sample trees 2 inch and greater DBH in point sample. Borderline trees must be measured for distance and diameter to determine if they are in the sample. Tree diameters must be recorded within +/- 1 inch of dbh.

11. **PRODUCT**, the highest value produce for each tree. (pulpwood, CNS, pole, sawtimber, veneer)

12. **MERCHANTABLE HEIGHT** in feet, of each merchantable size tree tallied to reflect the product assigned. (top minimum diameter by product table will be provided)

13. **TOTAL HEIGHT** in feet of each tree.

14. **PERCENT DEFECT**, for adjustment of merchantable volume estimated for each tree using a table provided.

Trees shall be tallied beginning from North and proceeding clockwise around the point. Each tally tree must be marked with a spot of paint, ink or durable crayon facing the point center.

The data collected must be provided at a minimum in DBF 4 (.dbf) database IV format, and delivered as stands are completed.

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 $\frac{1}{2}$

 \mathbb{R}^2

 $\frac{1}{2}$

 $\hat{\boldsymbol{\beta}}$

 $\hat{\boldsymbol{\beta}}$

Table 4B. Comparison of Site Index data from SSURGO database for study area by map unit level ($n = 20$ or 5%, whichever is greatest, for each map unit within a stand for LiDAR 2008 and 2011 data).

 $\frac{1}{2}$

 $\frac{1}{2}$

34

 $\hat{\mathcal{A}}$

 \sim \sim

 $\frac{1}{2}$

 $\sim 10^6$

 $*$ Indicates significantly different than SSURGO at $\alpha = 0.01$ level.

 $\hat{\boldsymbol{\gamma}}$

 $132.9(2.91)$ **

 $133.3(2.75)$ **

 3.0

110817

Table 5B. Comparison of Site Index soil data from SSURGO database by SSURGO Map Unit Symbol within a stand with

* Indicates signineantly different than SSURGO at $\alpha = 0.05$ level.
**Indicates significantly different than SSURGO at $\alpha = 0.01$ level.
Note: A negative number in growth indicates that a stand has been harvested, trees w

**Indicates significantly different than SSURGO at $\alpha = 0.01$ level.

 $\hat{\mathcal{A}}$

Table 6B. Comparison of Site Index soil data from SSURGO database by SSURGO Map Unit Symbol within a stand with
contat oak as Site Index species for the study area (2008 and 2011)

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 41

 $\label{eq:2.1} \frac{1}{\sqrt{2}}\int_{0}^{\pi} \frac{d\mu}{\lambda} \left(\frac{d\mu}{\lambda} \right) \left(\frac{d\mu}{\lambda} \right) \, d\mu$

 $42₁$

 $\label{eq:2.1} \frac{1}{2} \int_{\mathbb{R}^3} \frac{1}{\sqrt{2}} \, \frac{1}{\sqrt{2}} \,$

 $\frac{1}{2} \left(\frac{1}{2} \right)$

 $\hat{\boldsymbol{\gamma}}$

Note: A negative number in growth indicates that a stand has been harvested, trees were lost to some sort of damage, an overestimation in 2008, or an underestimation in 2011 has occurred.

 $\mathcal{L}^{\text{max}}_{\text{max}}$

 $\hat{\mathcal{E}}$

Table 8B. Comparison of Site Index soil data from SSURGO database by SSURGO Map Unit Symbol within a stand with white oak as Site Index species for the study area (2008) Table 9A. Comparison of Site Index soil data from SSURGO database by SSURGO Map Unit Symbol within a stand with yellow poplar as Site Index species for the study area (2008).

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Table 9B. Comparison of Site Index soil data from SSURGO database by SSURGO Map Unit Symbol within a stand with $(1100 \text{ rad} \cdot 2000 \text{ s})$ farthe study $\frac{1}{2}$ $C_{\pm\alpha}$ Indox $\frac{1}{2}$ \cdot \cdot \cdot \cdot \cdot

overestimation in 2008, or an underestimation in 2011 has occurred.

Table 10A. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with loblolly pine as Site Index species for the study area (2008).

** Indicates significantly different than SSURGO at $\alpha = 0.01$ level.

 $\frac{1}{2}$

 $\mathcal{A}^{\text{max}}_{\text{max}}$

 \overline{a}

Table 11A. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with scarlet oak as Site Index species for the study area (2008).

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Table 11B. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with scarlet oak as Site Index species for the study area (2008 and 2011).

Table 12A. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with shortleaf pine as Site Index species for the study area (2008).

 $\hat{\mathcal{A}}$

in an momentum furnitum	**Indicates significantly different than SSURGO at $\alpha = 0.01$ level.

Table 12B. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with shortleaf

Indicates significantly different than SSURGO at $\alpha = 0.01$ level.

 $\frac{1}{2}$

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 $\mathcal{L}_{\mathcal{A}}$

58

 $\frac{1}{2}$

Table 13A. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with white oak as Site Index species for the study area (2008).

 $\hat{\boldsymbol{\gamma}}$

 $\overline{}$

 $\hat{\mathcal{L}}$

Table 13B. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with white oak as Site Index species for the study area (2008 and 2011)

Table 14A. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with yellow poplar as Site Index species for the study area.

 $\hat{\boldsymbol{\gamma}}$

Table 14B. Comparison of Site Index soil data from SSURGO database for SSURGO Map Unit Symbol with yellow

Figure 9. A plot of the LiDAR derived site index (2008) against SSURGO site index (various dates) of 302 plots.

Figure 10. A plot of the SSURGO site index (2008) against Inventory measured site index (2008) of 302 plots.

Figure 11. A plot of the LiDAR derived maximum height (2008) against Inventory (2007-2008) measured maximum height of 302 plots.

Figure 12. A plot of the LiDAR derived maximum height (2011) against LiDAR derived maximum height (2008) of 258 plots.

Table 15. Summary statistics for Figures 11 and 12.

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