# INVISTIGATION ON CANOPY HEIGHT AND DENSITY DIFFERENTIATIONS IN THE MANAGED AND UNMANAGED FOREST STANDS USING LIDAR DATA (CASE STUDY: SHASTKALATEH FOREST, GORGAN)

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KEY WORDS: Canopy height, Canopy density, Lidar, Managed forest, Unmanaged forest, Dr. Bahramnia forest plan of Gorgan

## **ABSTRACT:**

Forest management plans are interesting to keep the forest stand natural composite and structure after silvicultural and management treatments. In order to investigate on stand differences made by management treatments, comparing of these stands with unmanaged stands as natural forests is necessary. Aerial laser scanners are providing suitable 3D information to map the horizontal and vertical characteristics of forest structures. In this study, different of canopy height and canopy cover variances between managed and unmanaged forest stands as well as in two dominant forest types were investigated using Lidar data in Dr. Bahramnia forest, Northern Iran. The in-situ information was gathered from 308 circular plots by a random systematic sampling designs. The low lidar cloud point data were used to generate accurate DEM and DSM models and plot-based height statistics metrics and canopy cover characteristics. The significant analyses were done by independent T-test between two stands in same dominant forest types. Result showed that there are no significant differences between canopy cover mean in two stands as well as forest types. Result of statistically analysis on height characteristics showed that there are a decreasing the forest height and its variance in the managed forest compared to unmanaged stands. In addition, there is a significant difference between maximum, range, and mean heights of two stands in 99 percent confidence level. However, there is no significant difference between standard deviation and canopy height variance of managed and unmanged stands. These results showd that accomplished management treatments and cuttings could lead to reducing of height variances and converting multi-layers stands to two or single layers. Results are also showed that the canopy cover densities in the managed forest stands are changing from high dense cover to dense cover.

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## 1. Introduction

The Hyrcanian original and natural forests in the northern Iran are almost high dense canopy cover and multi-layers vertical structure with high variance. Forest management plans are certificated approaches for applying forest managing and silviculture treatments programs to optimal forest management and keeping the forest stand natural composite and structure. The forest managers are interesting to aware of forest composite and structure situation before and after silvicultural and management treatments. In order to investigate on stand differences made by management plan, comparing of these stands with unmanaged stands as a nature of forests is necessary.

Traditional way to measure forest stucture and composits variables is field surveying by different methods and tools. These measurements are generally expensive, time-consuming and labour intensive, as well as difficult to perform, especially in mountainous and dense forests. Use of remote sensing tools and sources with different cappabilites and abilities is an alternative way particularly in large areas. Improving satellite data sources and classification methods offer new opportunities for obtaining more accurate forest biophysical maps. Rapid improvements in remote sensing technology have led to various types of sensors, such as multispectral, hyper spectral, ultraviolet, thermal sensors, light detection and ranging (Lidar), radio detection and ranging (radar), and other sensors. Each type of sensor has been designed for specialized purposes, tasks, and different applications. These new potential sources have been shown to be appropriate tools to assess and monitor forest attributes with reasonable accuracy levels.

The optical sensors in visible and infrared wavelengths are producing information based on registering reflectances of objects in diferent radiometric, spectral, and spatial resolutions. The optical aerial/satellite imagery usually presents twodimensional spectral information and reflectance responses of a canopy cover's surface. Launching of many satellites providing data of submetric ground resolution can be usefull for accurate detection forest variables. However, LiDAR data from an airborne laser scanner (ALS) provides semi three-dimensional data set relating to canopy cover and canopy height. Nasset (1998 and 2002) demonstrated that lidar data for plot-based estimation error of maximum and mean canopy height with full canopy closure is less than 0.5 meters. Capacity of LiDAR data to estimate vertical and horizontal canopy structure of stands was studied in the some research such as Coops et al (2007).

In this study, different of canopy cover density and crown height variances between managed and unmanaged forest stands were investigated in Dr. Bahramnia forest management plan in managed forest stands and unmanged forest stands as well as in two dominant forest types using Lidar data.

## 2. Material and Methods:

#### 2.1. Study area

The study area comprises 1100 hectares in the Southeast part of the Golestan province, Eastnorth of Iran, with elevation ranging from 270 to 740 m above mean sea level (Figure 1). The research was done on managed (parcls of 4 to 24 seri1) and unmanged (parcls of 2 to 7 of seri 2 Dr. Bahramnia forestry plan) stands. The previous forest silviculture treatment method in managed stands was shelterwood cutting (1972-1992), however, they are currently treating by single tree selection and cutting (close to nature) method.



Fig.1: Location of study area in the Golestan province of Iran

#### 2.2. Field data

We applied a systematic random sampling method to collect field data with  $150 \times 200$  meter network (3.33% sampling intensity) and plots were circular with 0.1 ha. Totally, the 308 plots (219 plots in managed and 87 plots in unmanged stands) were measured in study area. In each plot, hieht of all trees with a diameter greater than 12.5 cm at breast height was measured by Vertex VL 402 device. The geographic center of each plot was recorded using a differential GPS (Trimble R3) device.

## 2.3. Lidar data:

Rayan Naqshe Company acquired the laser scanner data for this study under leaf on canopy condition on the 12th of October 2011using Riegl LMS Q5600 laser scanning system. The laser wavelength and mean point density are 1064 nm and about 4  $m^2$ , respectively. In addition, the flight elevation was approximately 1000 m above the ground. For more information please look at the table 1.

Table 1: LMS Q560 Riegl charestrictistics

30	Minimum Range	-
20	Accuracy	
10	Precision	
NIR (1069)	Laser Wavelength	
$\pm 22.5, \pm 30$	Scan Angle Range	
$\leq 0.5 \text{ mrad}$	Laser Beam Divergence	
240≥	Laser puls speed (KRZ)	
4	average measurement	
	density	
<50	Vertical accuracy	
<30	Accuracy horizontal	
IMU-lle	ĪMU	
400>	IMU sampling rate	2.4.

#### sing Lidar data:

Rayan Naqshe Company using RayAnalyze software analyzed the full-wave form data. The delivery consist of point data coordinate together with additional parameters width, amplitude, number of target within laser beam and total number of target within laser beam. In addition, the points cloud Lidar data using RiProcess software classified in first, last and other pulses. Rayan Naqshe Company accomplished initial processing of laser data. After processing, it observed that there are many noises in the last, first and other pulses. Therefore, all outliers with exceeder than 50 meters in last, first and other pulses were removed using Terrascan software.

The last pulses data were used to create to the bare earth surface using Kraus and Pfeiffer and fusion software. A triangulated irregular network (TIN) was used to create a digital terrain model (DTM) using X, Y and Z (height) values of the individual terrain ground points acquired from Kraus and Pfeiffer algorithm (1998) with spatial resolution of 1m. The DTM accuracy was assessed using 90 ground elevation points recorded by DGPS. The results showed that mean elevation

differences between DTM and DGPS points were less than 40cm. In addition, a one-meter resolution digital surface model (DSM) was created using first and last pulses data by TIN algorithm in Fusion software. Then, canopy height model (CHM) was created using DTM and DSM models and applying 2 meters height break in Fusion software. The accuracy of CHM was assessed by height of 90 trees measured by Vertex VL 402 device. The results showed that the mean height difference between CHM and height of ground trees was less than 90cm. Then, according to previous study (Naesset, 2002, 2004) and Hureich and Thoma (2008), height metrics and canopy cover metrics were extracted using Fusion software with considering pixel size of  $31.623 \times 31.623$  m corresponding to plot size (1000 m<sup>2</sup>). Canopy height structure metrics in plot level were extracted by Fusion software (Table 2).

Table 2: Plot based canopy	height structure metrics
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Height metrics	
h <sub>p0</sub> , h <sub>p90</sub>	Height
r · · ·	Percentiles
Maximum Height of points in plots	Hmax
Minimum Height of points in plots	Hmin
Mean Height of points in plots	Hmean
Median Height of points in plots	Hmedian
Coeficient variance of height points in plots	Hcv
Standard deviation of height points in plots	hSD
skewnes of height points in plots	Hskeewness
kurtosis of height points in plots	Hkurtosis
variance of height points in plots	Hvariance

#### Statistical significant analysis:

In order to comparing the hieght change and percent canopy cover status in the managed and unmanaged stands, General statistical measures and significant analyses were done by independent T test between two stands as well as between two same dominant forest types (i.e. Carpinus-Parrotia and Parrotia-Carpinus).

## **Results:**

Table 3 shows the statistics measures of canopy cover metrics extracted on Lidar data in managed stand (seri 1) and unmanged stands (seri2). The statistics measures of canopy cover metric of forest types extracted from Lidar data in unmanaged and managed stands are showed in tables 4 and 5.

Table 3: Statistics measures of canopy cover metric extracted from Lidar data in managed and unmanged stands

Attribute	Seri	Plots	Mean	Min (%)	Max (%)	Range SD (%)
			(%)			(%)
Canopy	Ι	219	98.992	52.811	100	47.189 3.389
cover (%)	Π	87	99.491	97.883	100	2.117 0.444

Table 4: Statistics measures of canopy cover metric of Carpinus-parrotia forest type extracted from Lidar data in managed and unmanged stands

Attribute	Seri	Stand type	Mean	Min (%)	Max (%	)Range	SD (%)
			(%)			(%)	
Canopy	Ι	Carpiuns-	99.03	85.53	100	44.47	2.13
cover (%)	Π	parrotia	99.49	98.37	99.98	1.61	0.37

Table 5: Statistics measures of canopy cover metric of Parrotiacarpinus forest type extracted from Lidar data in managed and

Attribute	Seri	Stand type	Mean	Min (%)	Max (%)	Range	SD (%)
			(%)			(%)	
Canopy	Ι	Parrotia -	99.29	97.06	100	2.94	0.70
cover (%)	Π	Carpiuns	99.36	97.88	99.97	2.08	0.54

unmanged stands

Results of significant different of percent canopy cover (extracted by Lidar data) between managed and unmanged stands using independent T-test showed that there is not significant different between two stands in 95 percent confidence level (table 6) as well as in same forest type (table 7 and 8).

Table 6: Results of significant analysis of percent canopy cover in managed and unmanged stands

Attribute	T statisticsSigni	ficant level
Percent canopy cover	-1.67	0.173 <sup>ns</sup>
<sup>ns</sup> insignificant in	95 percent confide	ence level

Table 7: Results of significant analysis of percent canopy cover of Parrotia- carpinus forest type in managed and unmanged

	stands	
Attribute	T statisticsSigni	ficant level
Percent canopy cover	-1.176	0.244 <sup>ns</sup>
<sup>ns</sup> insignificant in	95 percent confide	ence level

Table 8: Results of significant analysis of percent canopy cover of Carpinus-parrotia forest type in managed and unmanged

	stands		
Attribute T statisticsSignificant leve			
Percent canopy cover	-1.403	0.164 <sup>ns</sup>	
<sup>ns</sup> insignificant in	95 percent confider	nce level	

Results of significant different of canopy structure metrics extracted by Lidar data between managed and unmanaged stands using independent T-test (table 9) and forest type of both stands (tables 10 and 11) showed that in some canopy hight attributes there is significant different, however in some cases there is not insignificant different.

Table 9: Results of significant analysis of canopy height in managed and unmanged stands

Attribute	Significant	Т
	level	statistics
Minimum canopy height	0.121 ns	-1.555
Maximum canopy height	$0.00000_{**}$	-6.53
Range canopy height	$0.00000_{**}$	-3.576
Mean canopy height	$0.00000^{**}$	-5.652
Standard deviation canopy height	0.078 <sup>ns</sup>	-1.769
Variance canopy height	0.063 <sup>ns</sup>	-1.866

gnificant in 99 percent confidence level . , <sup>ns</sup> insignificant in 95 percent confidence level

Table 10: Results of	significant analysis	s of canopy heig	ght of
Carpinus-parrotia fore	st type in managed	and unmanged	stands

Significant	Т
level	statistics
0.549 <sup>ns</sup>	-0.601
$0.000^{**}$	-4.459
0.021**	-2.347
$0.000^{**}$	-4.329
0.886 <sup>ns</sup>	-0.143
0.726 <sup>ns</sup>	-0.351
	Significant level 0.549 <sup>ns</sup> 0.000** 0.021** 0.000** 0.886 <sup>ns</sup> 0.726 <sup>ns</sup>

gnificant in 99 percent confidence level . , <sup>ns</sup> insignificant in 95 percent confidence level

Table 11: Results of significant analysis of canopy height of Parrotia-carpinus forest type in managed and unmanged stands The International Archives of the Photogrammetry, Remote Sensing and Spatial Information Sciences, Volume XL-1/W5, 2015 International Conference on Sensors & Models in Remote Sensing & Photogrammetry, 23–25 Nov 2015, Kish Island, Iran

Attribute	Significant	Т
	level	statistics
Minimum canopy height	0.254 <sup>ns</sup>	-1.15
Maximum canopy height	0.157 <sup>ns</sup>	-1.432
Range canopy height	0.956 <sup>ns</sup>	-0.056
Mean canopy height	0.068 <sup>ns</sup>	-1.858
Standard deviation canopy height	0.368 <sup>ns</sup>	-0.907
Variance canopy height	0.244 <sup>ns</sup>	-1.176

\*\*significant in 99 percent confidence level., <sup>ns</sup> insignificant in 95 percent confidence level

# Conclusion:

. In this study, different of canopy cover density and crown height variances between managed and unmanaged forest stands were investigated using Lidar data and statistical significant analysis. The significant analyses were done by independent Ttest between two stands in same dominant forest types. Results showed that there are no significant differences between canopy cover mean in two stands as well as forest types. Result of statistically analysis on height characteristics showed that there are a decreasing the forest height and its variance in the managed forest compared to unmanaged stands. In addition, there is a significant difference between maximum, range, and mean heights of two stands in 99 percent confidence level. However, there is no significant difference between standard deviation and canopy height variance of managed and unmanged stands. These results showd that accomplished management treatments and cuttings could lead to reducing of height variances and converting multi-layers stands to two or single layers. Results are also showed that the canopy cover densities in the managed forest stands are changing from high dense cover to dense cover.

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