

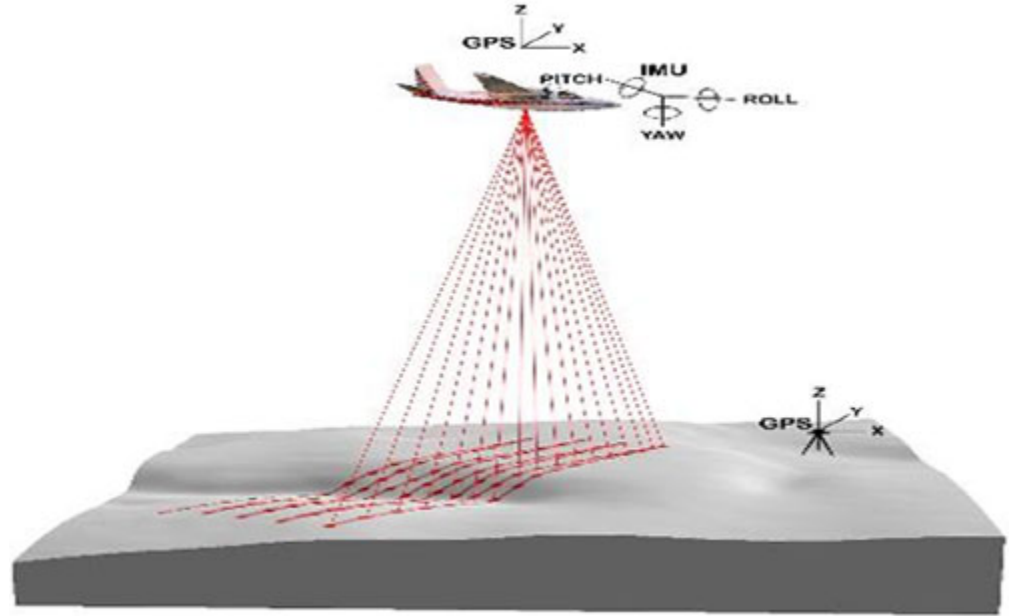
# LIDAR TECHNOLOGY AND APPLICATIONS IN FORESTRY

Vincenzo Giannico, University of Bari, Italy

# LiDAR

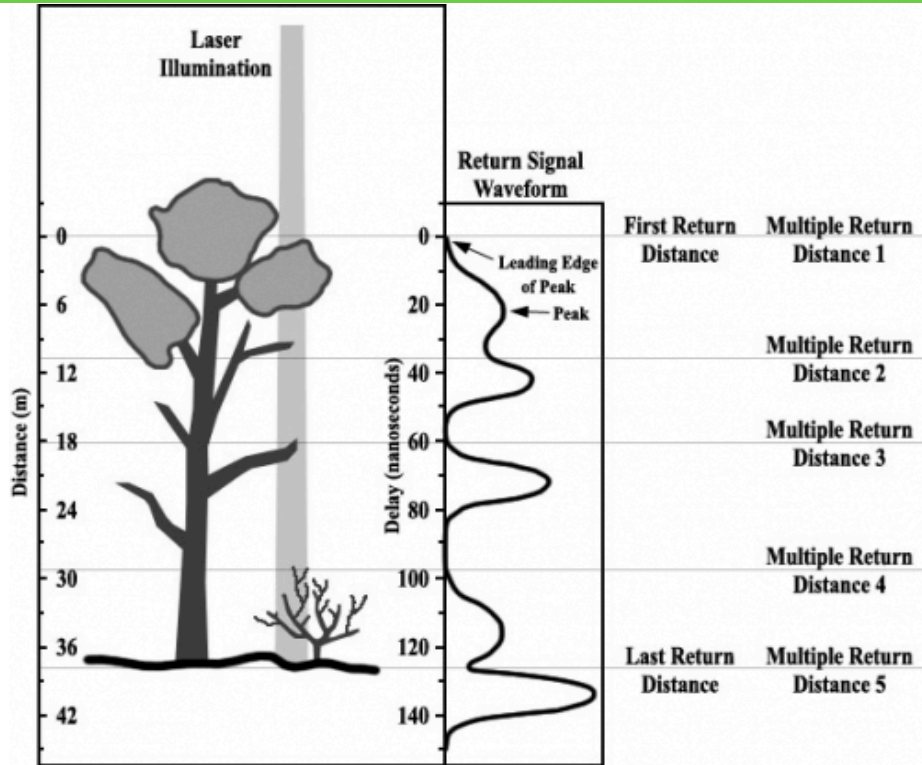
Remote sensing technique that creates 3D models from reality:

- Active sensor capable of generating up to 400.000 laser impulses per second
- At the same time, the sensor detects the reflections (returns), which bounce off the surface to the sensor
- ALS and TLS
- Based on impulse-reflection timing, GPS, and a Gyroscope, the reflecting surface is located in space  $(x,y,x)$



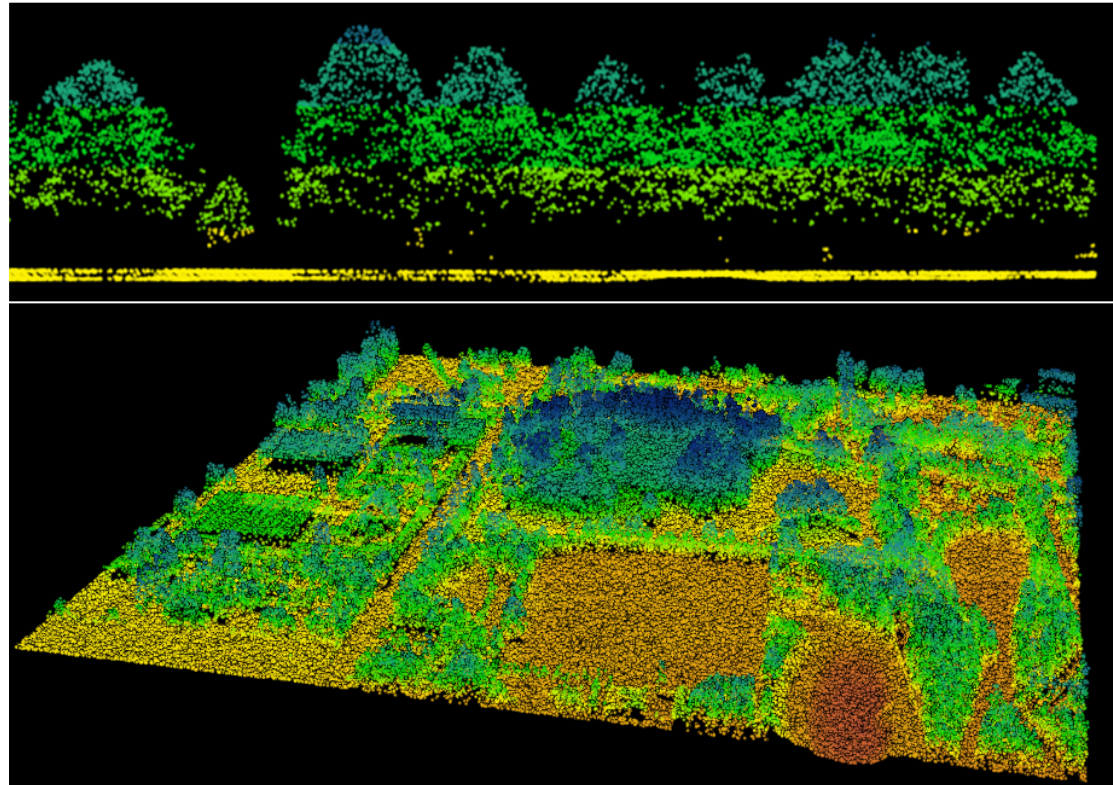
# LiDAR

- A laser impulse becomes larger during its way to the ground, describing a cone (Footprint) (0.2-70m)
- Recent LiDAR sensors can record multiple returns for each impulse and penetrate non-uniform surfaces, such as vegetation
- Each return is discretized in a point
- The result is a highly dense point cloud



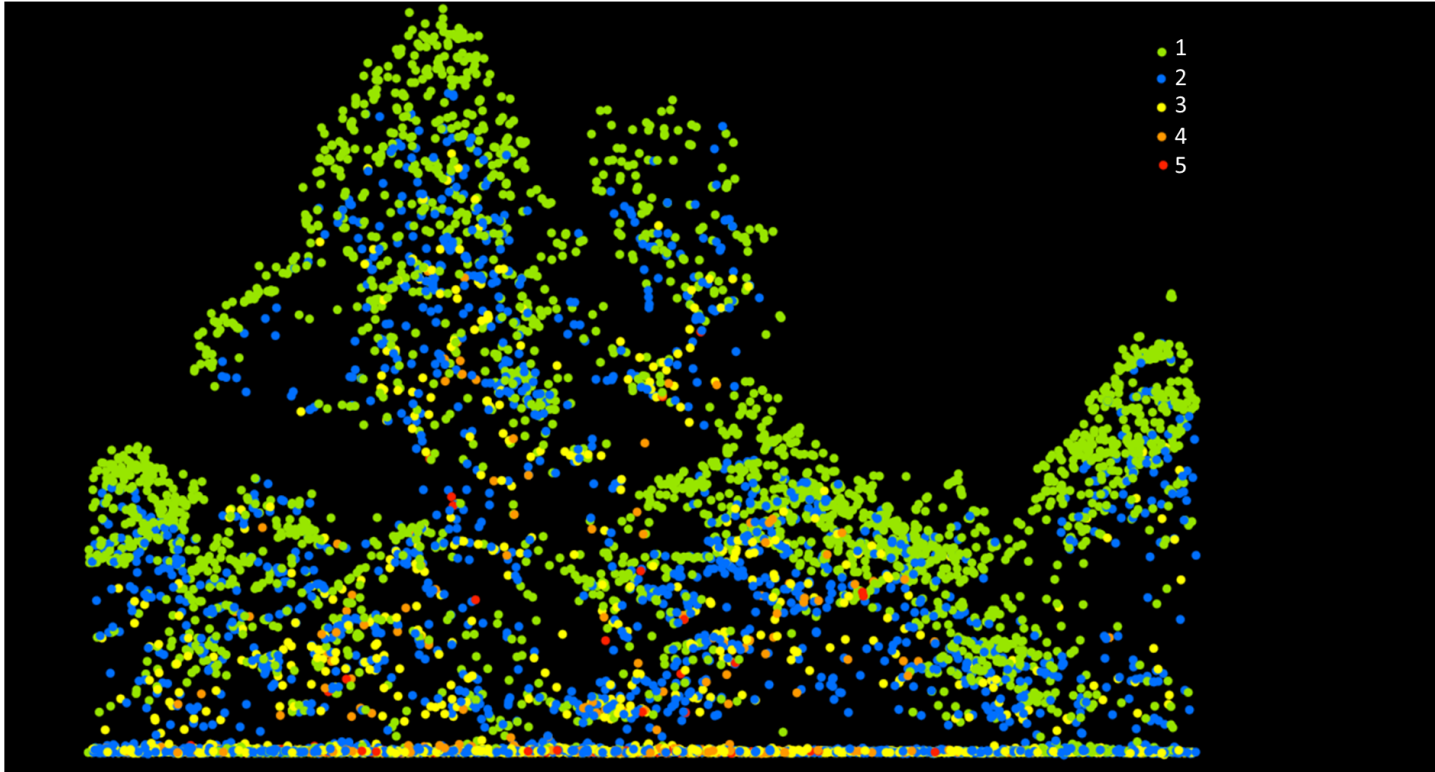
# LiDAR

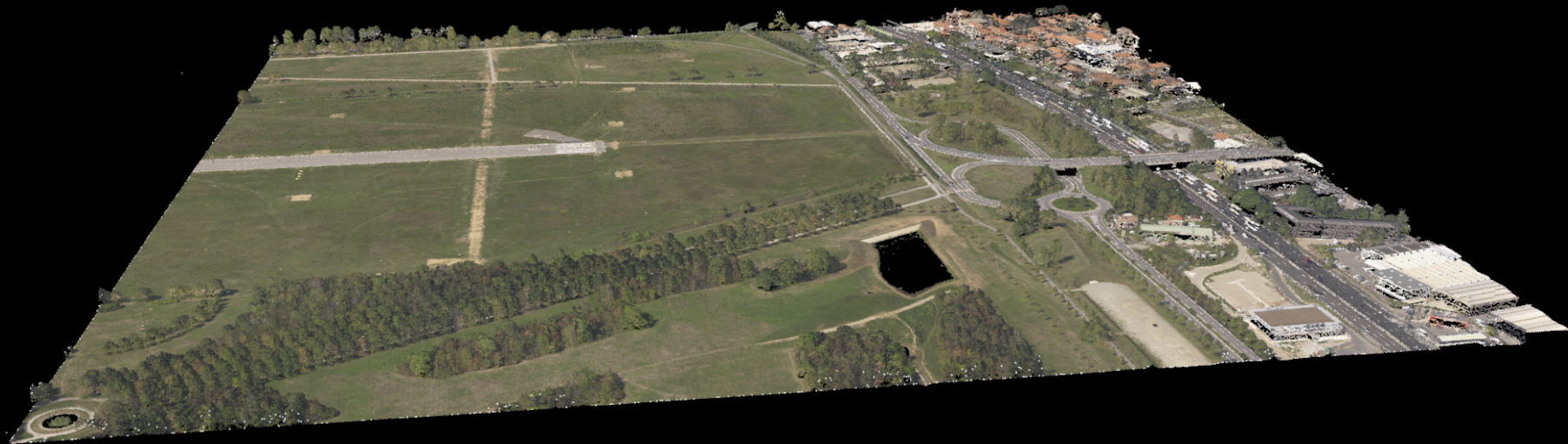
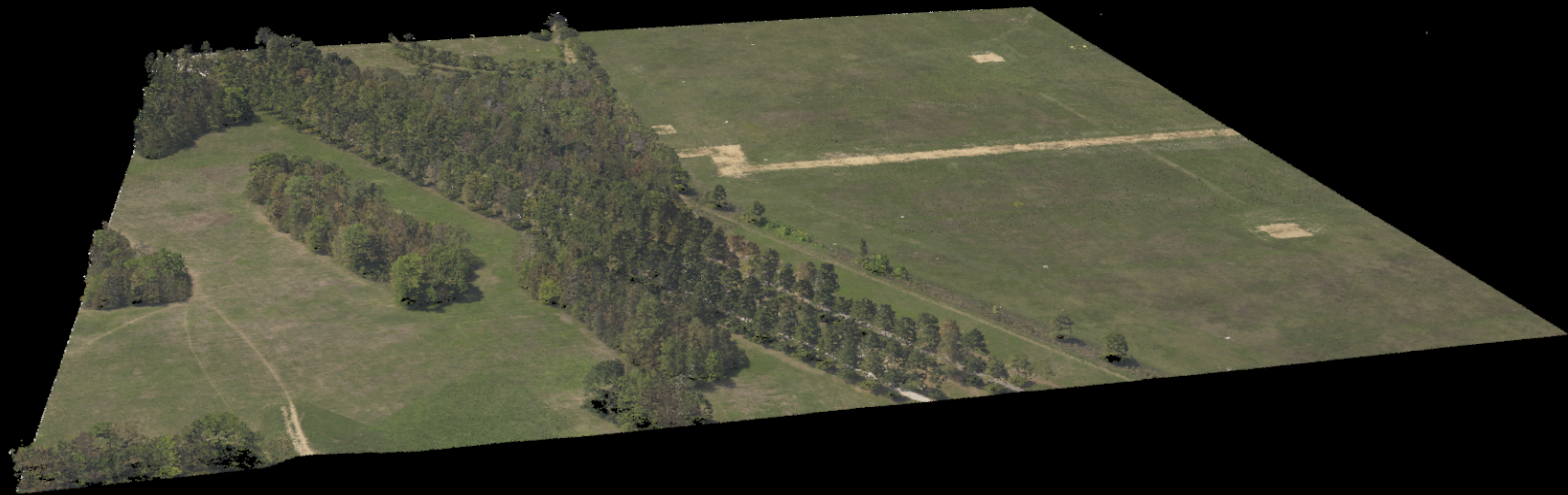
- 3D point cloud representing the portion of the earth surface scanned
- Accurate profile of forested areas
- Point density and number of returns depend on sensor frequency and flight condition
- In LAS files additional features for each point are stored (return number, classification, scan angle, RGB, etc. )



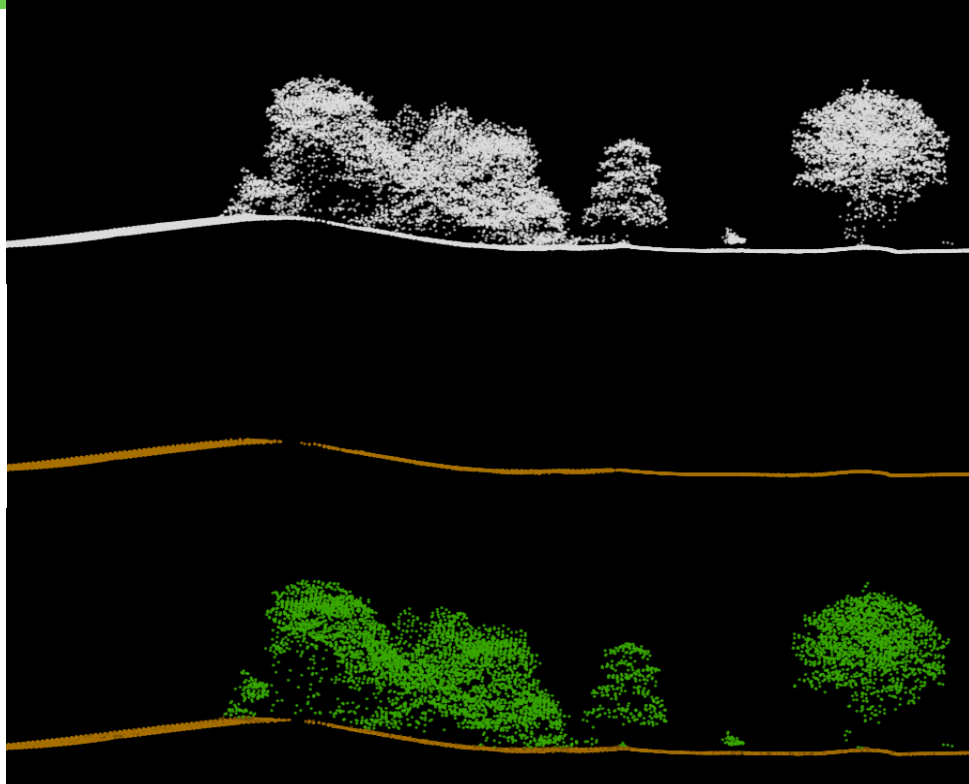


# LiDAR

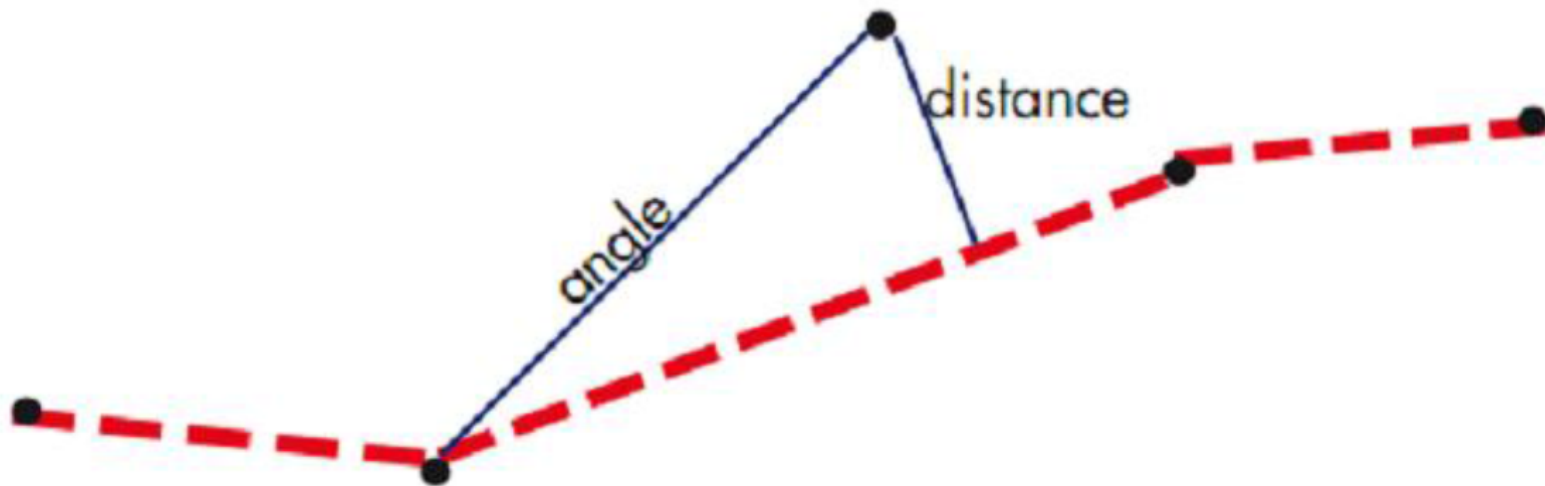




# Preprocessing



# Preprocessing: Ground points



*In ground point iteration user-defined parameters decide if points belongs to ground or not.*



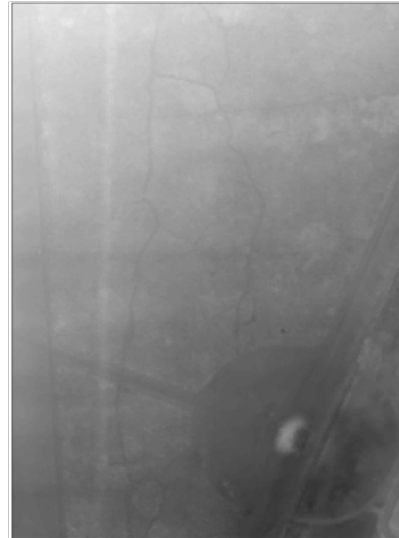
# Processing: surface models

Digital models (raster) :

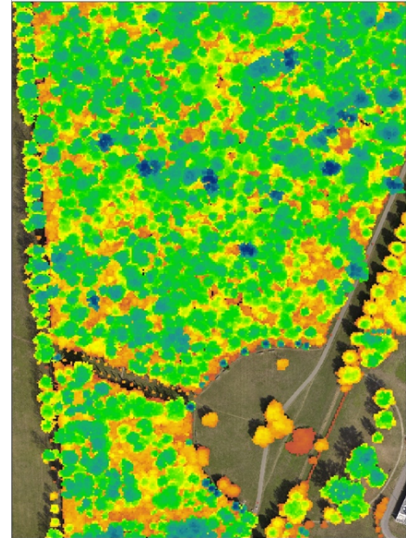
- *Digital Terrain Model*
- *Digital Surface Model*
- *Canopy Height Model*



ORTHOPHOTO

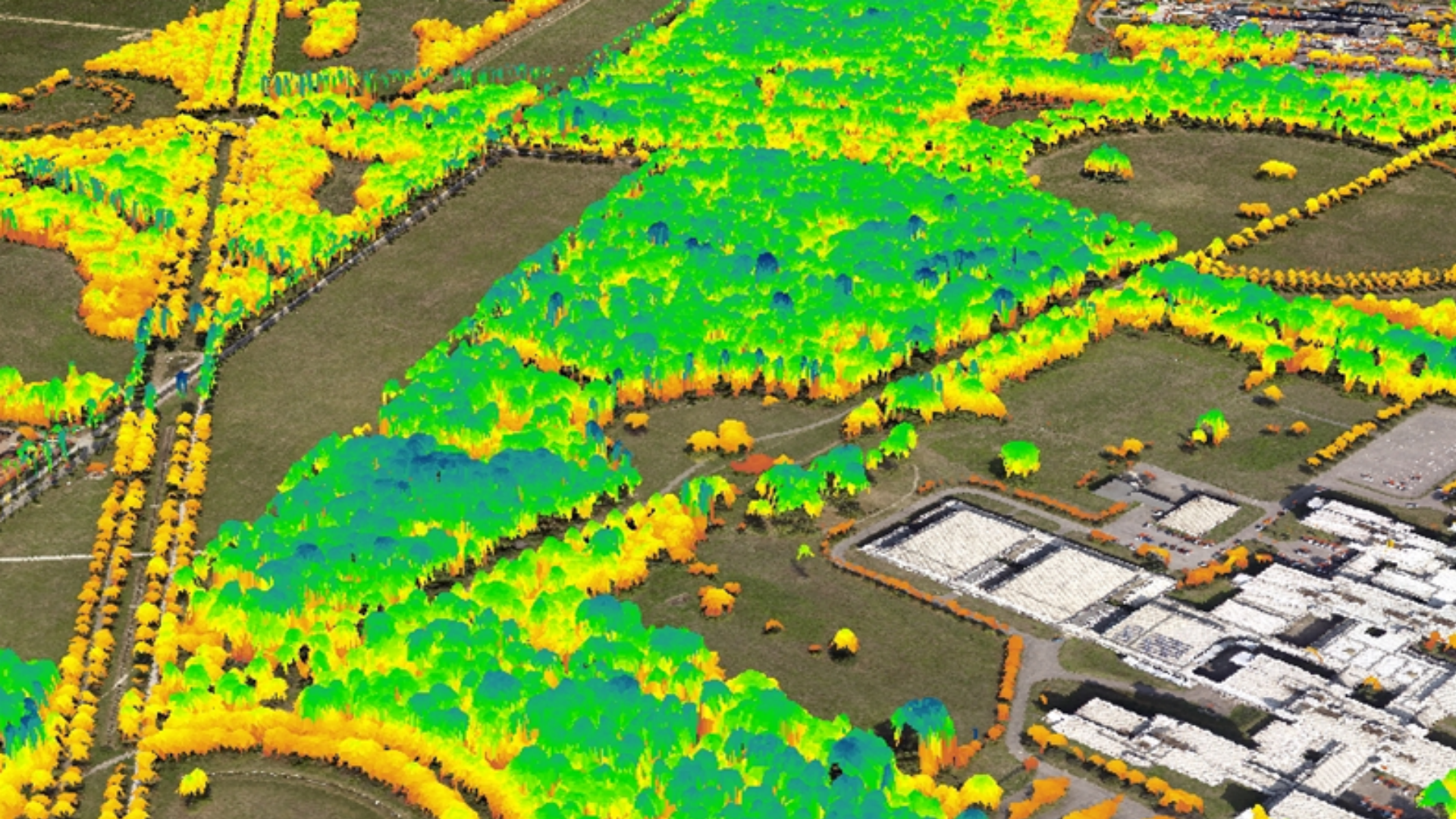


DTM

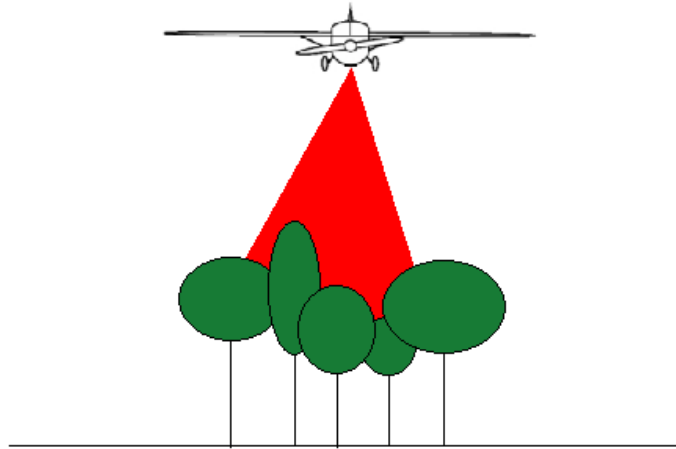


CHM





# LiDAR applications in forestry

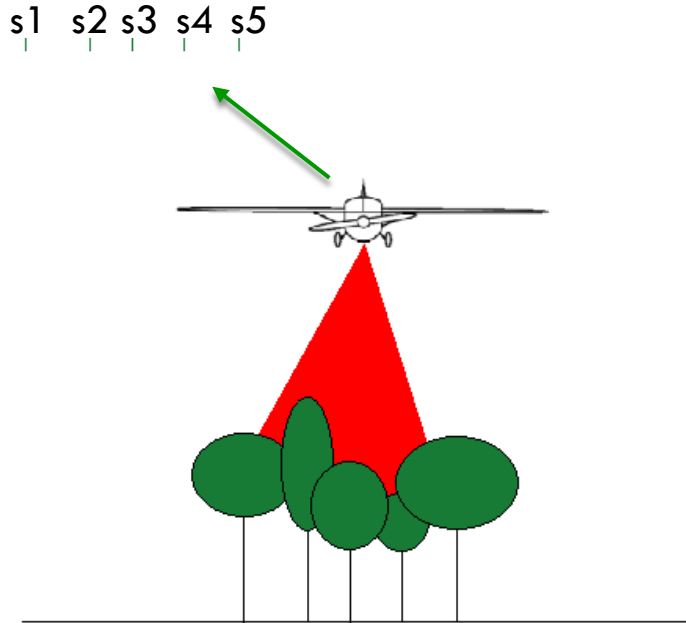




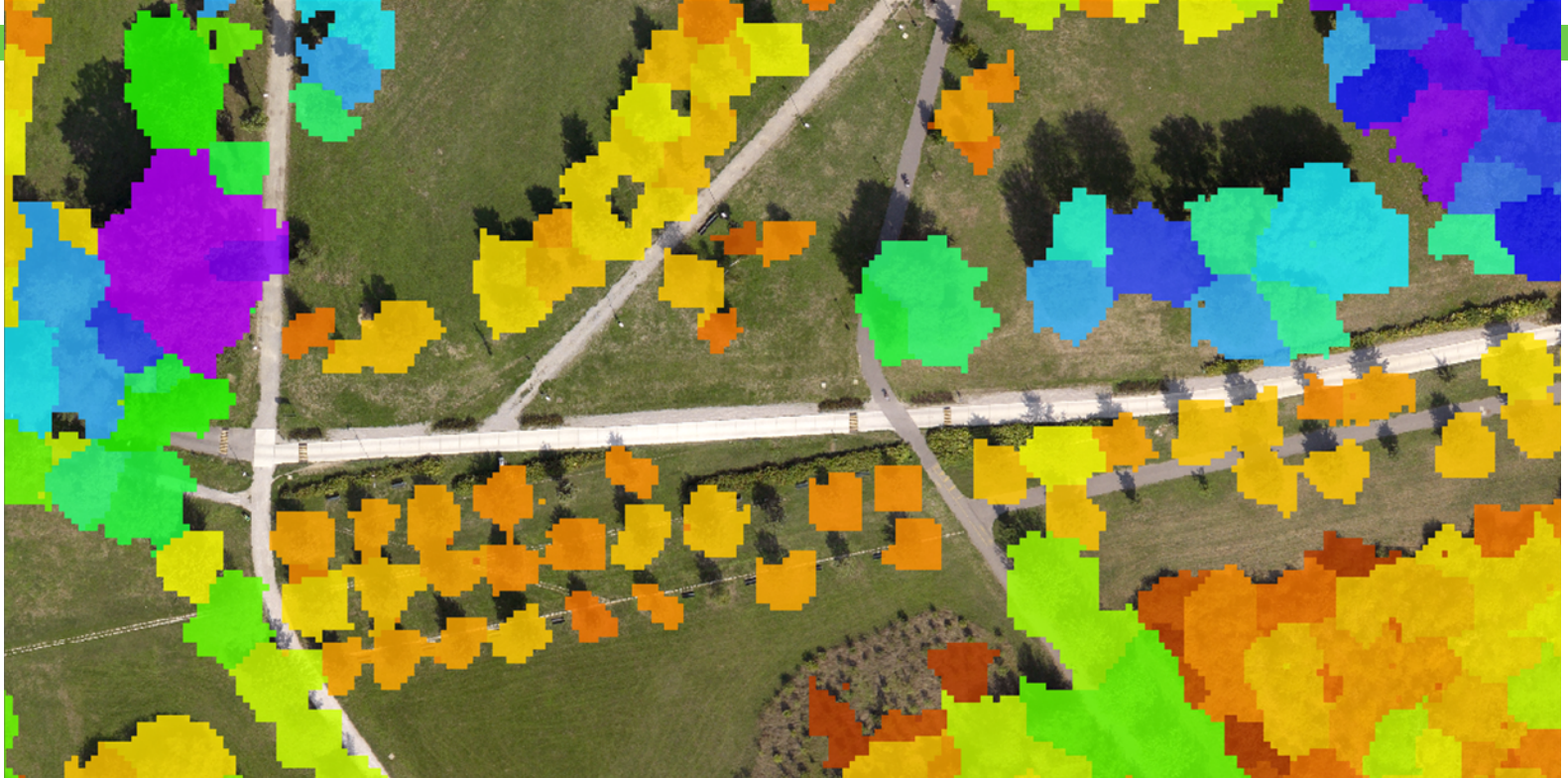
# LiDAR applications in forestry

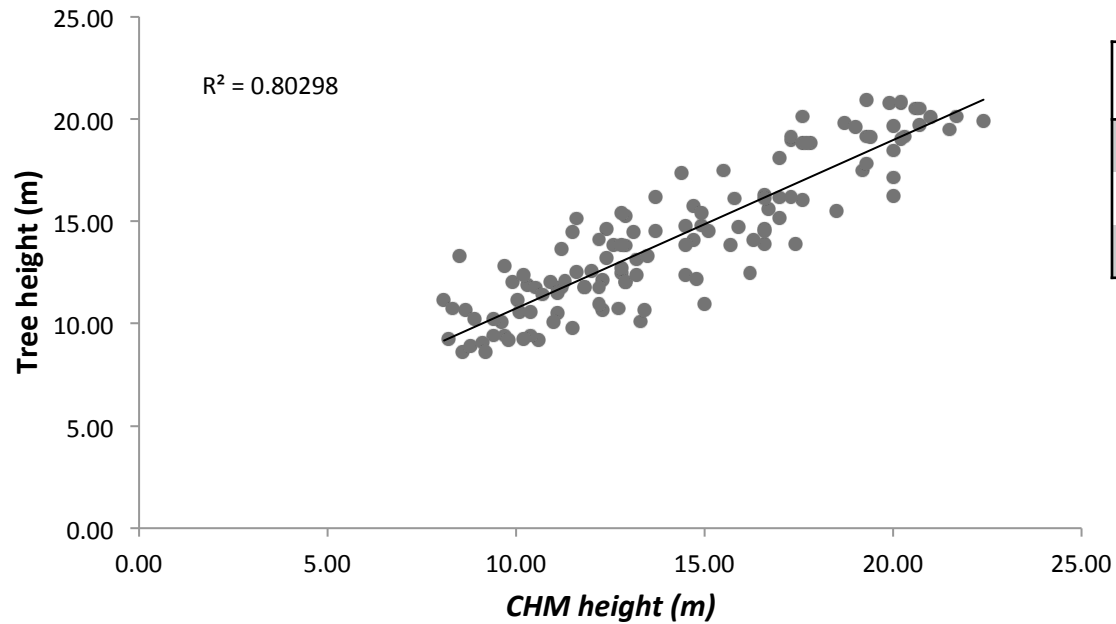
## Single tree approach

- Tree segmentation
- Tree species detection
- Individual tree biophysical characteristics (DBH, H, Crown diameter)
- ...



# LiDAR applications in forestry



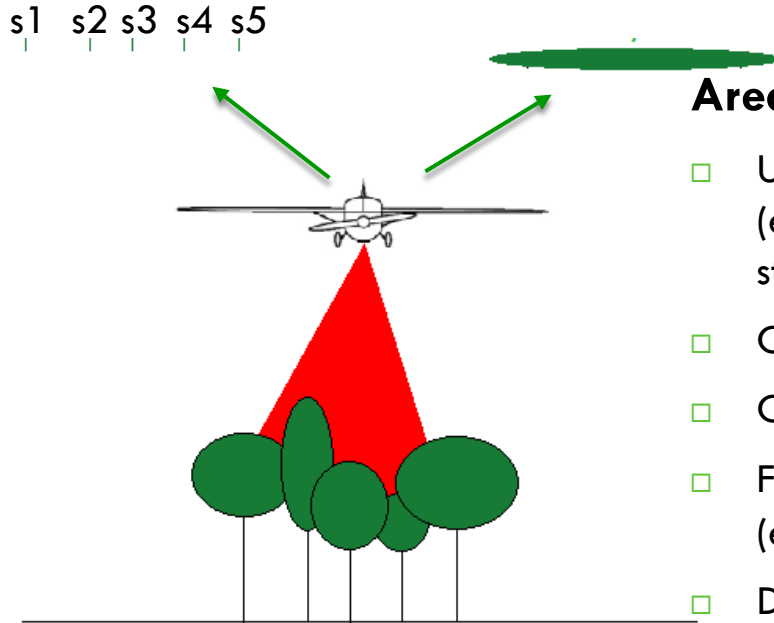


	Mean	Max	Min	$\sigma$
<b>H Field</b>	14.28	22.40	8.09	3.78
<b>H CHM</b>	14.26	20.96	8.59	3.47
<b>Delta</b>	0.02	1.44	-0.50	0.31

# LiDAR applications in forestry

## Single tree approach

- Tree segmentation
- Tree species detection
- Individual tree biophysical characteristics (DBH, H, Crown diameter)
- ...



## Area-based approach

- Use of **LiDAR-derived metrics** (e.g. percentiles, descriptive statistics, indices, etc.)
- Canopy cover
- Calculation of biomass/ha
- Forest stand level classifications (e.g. sp. composition, density)
- Diversity indices detection
- ...

# LiDAR metrics

---

## Examples of LiDAR-derived metrics (descriptive statistics)

---

95th percentile height of first returns (m)

Similar to maximum height but less sensitive to anomalously high points.

---

Skewness of all returns

A measure for the degree of symmetry in the Point-cloud density distribution.

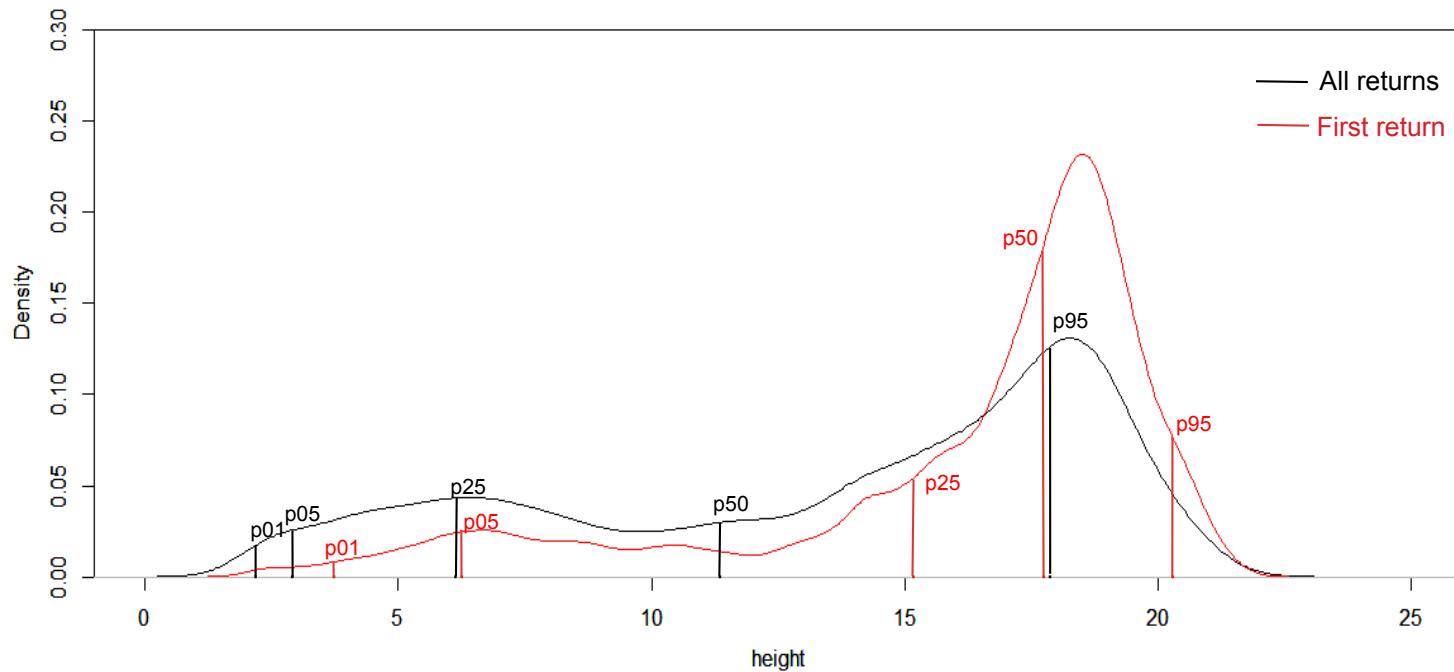
---

Kurtosis of all returns

A measure for the degree of peakedness/flatness in the Point-cloud density distribution.

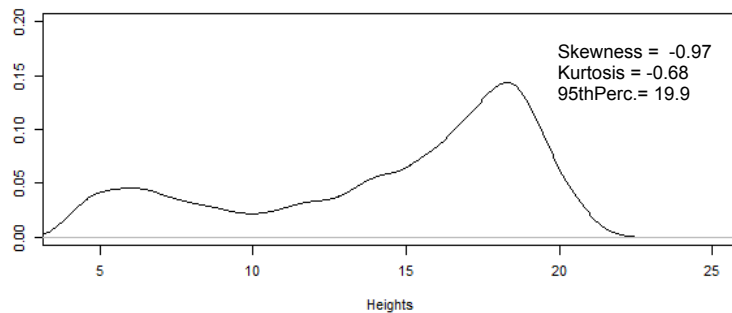
---

# LiDAR metrics

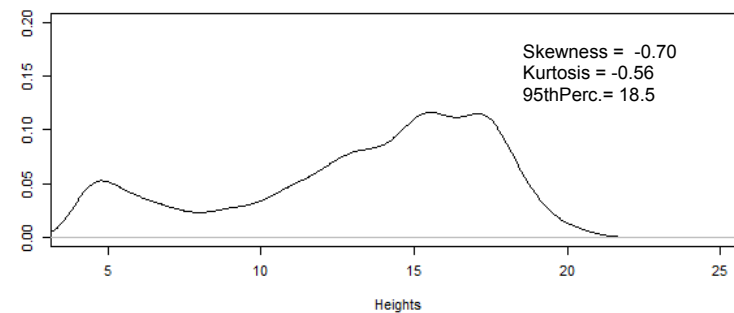


“Mature” stands → negatively skewed

1A

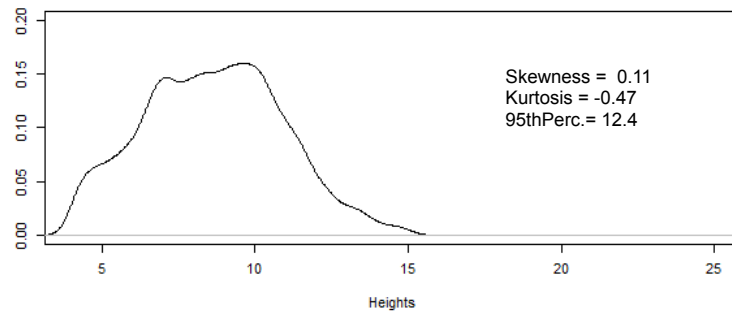


2A

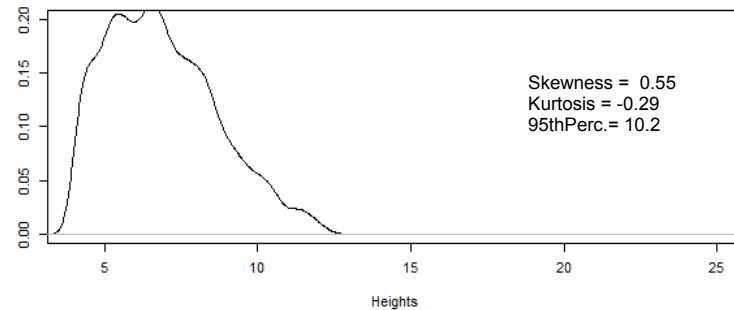


“Young” stands → positively skewed

25A



28A





# LiDAR metrics

---

## Examples LiDAR-derived metrics (indices )

---

Canopy density index

The canopy density is computed as the number of all points above the cover cutoff divided by the number of all returns.

---

Canopy cover index

The number of first returns above the cover cutoff divided by the number of all first returns.

---

Vegetation permeability

Proportion of first returns for which there is a second (either vegetation or ground) return; higher proportion of these dual returns indicates more light energy penetrating the canopy.

---

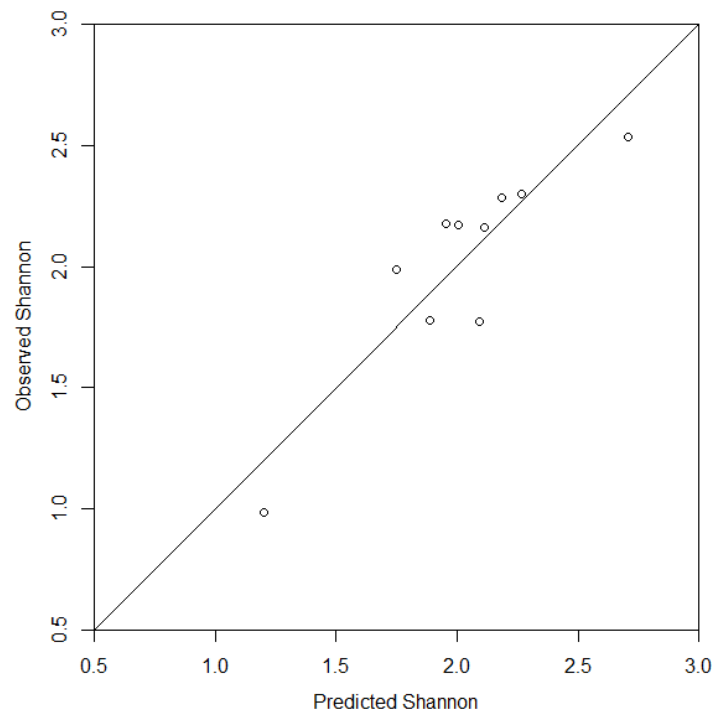
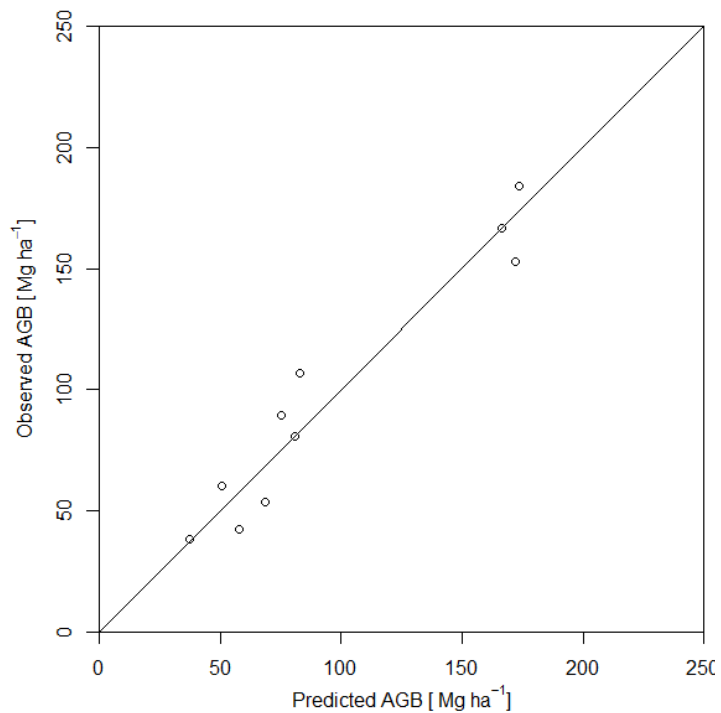
# MODELS: BIOPHISICAL PARAMETERS

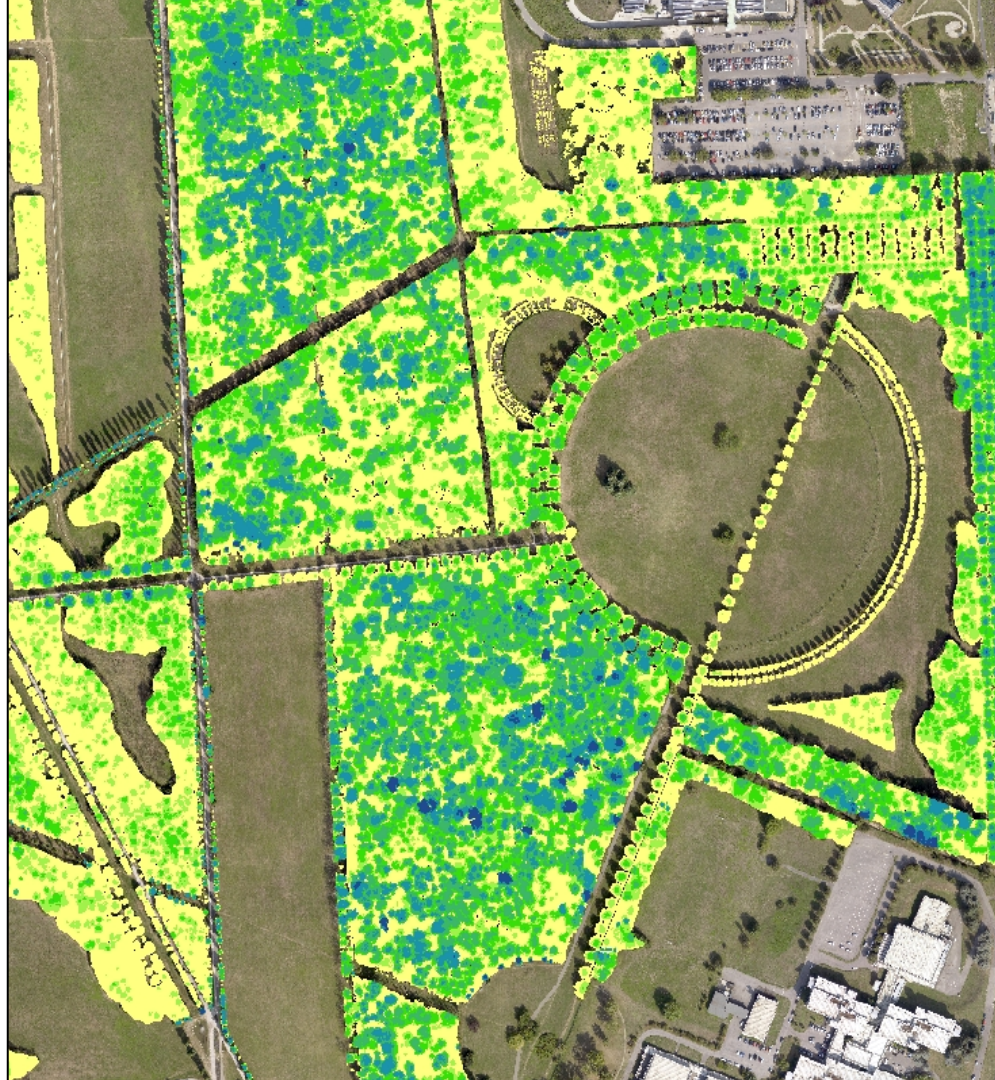
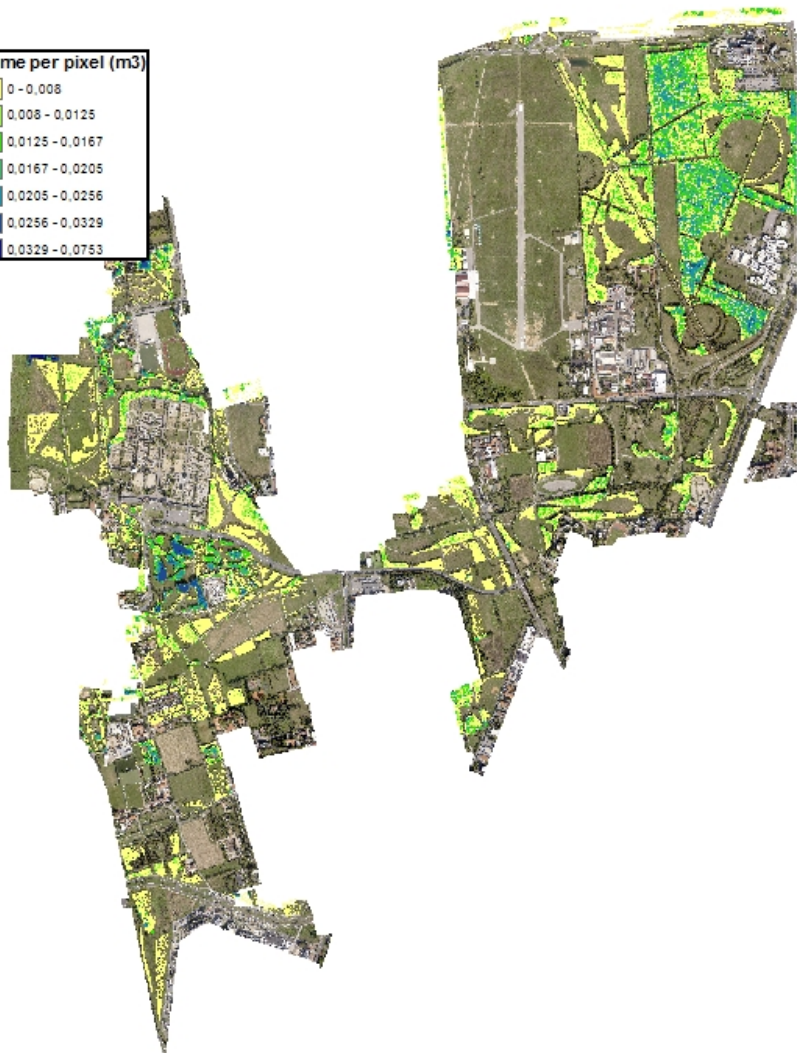
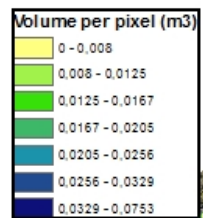
<i>Models (p&lt;0.005)</i>	$\beta_0$	$\beta_1$	$\beta_2$	$\beta_3$
$\ln(AGB) = \beta_0 + \beta_1 \ln(p95\_all) + \beta_2 \ln(p01\_first) + \beta_3 \ln(p25\_first)$	6.3746	1.4907	-0.7169	0.7969
$Shannon = \beta_0 + \beta_1 \ln(kurtosis\_first) + \beta_2 \ln(p01\_first) + \beta_3 \ln(p25\_first)$	0.85199	0.44057	0.50977	-0.17491

<i>Models (p&lt;0.005)</i>	<i>RMSE</i>	<i>RMSEcv</i>	<i>normRMSE</i>	<i>normRMSEcv</i>	<i>adjR<sup>2</sup></i>
<i>AGB</i>	13.5 [Mg h <sup>-1</sup> ]	26.6 [Mg h <sup>-1</sup> ]	0.0925	0.183	0.84
<i>Shannon Index</i>	0.18	0.39	0.12	0.25	0.7

# MODEL ACCURACY





# LiDAR: summary

## Advantages

- High accuracy in estimation of biophysical parameters correlated with structural characteristics of forest stand
- Less expensive compared to field surveys (only limited field surveys are needed)
- Great potential in combination with other type of data (e.g. Multi- or Hyperspectral data)

## Limitations

- Low accuracy in estimating understory vegetation's biophysical characteristics
- Limited by  $< 50\%$  veg. cover
- Very low biochemical information about the scanned surface (as opposed to optical passive sensors)

This presentation was made available through collaborations in the Landscape Ecology and Ecosystem Science Lab (LEES) at Michigan State University with University of Bari, Italy.

For more materials and resources, or to learn how you can collaborate with us, visit the LEES website at:  
<http://lees.geo.msu.edu/>