

# New Efficient and Automatic Approach to Extract Dendrometric Features from Terrestrial LiDAR Point Clouds in Forest Inventories

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# 1. Introduction

Some important facts (and one question that needs to be answered):

- ✓ **Forests play a key role in the promotion of ecosystem services and social benefits.** Costanza et al. (1997) estimated the annual value of forest ecosystem services at \$4.7 trillion/year, representing approximately 15% of the World GNP (gross national product).
- ✓ Forest area in Mediterranean countries has been increasing since 1990. Unfortunately, **an increasing forest area tell us nothing about forest degradation and potential capacity to adapt to climate change.** It is needed to take a closer look.

In this sense, **effective monitoring of forest structure turns out to be a key role for adapting to climate change.**

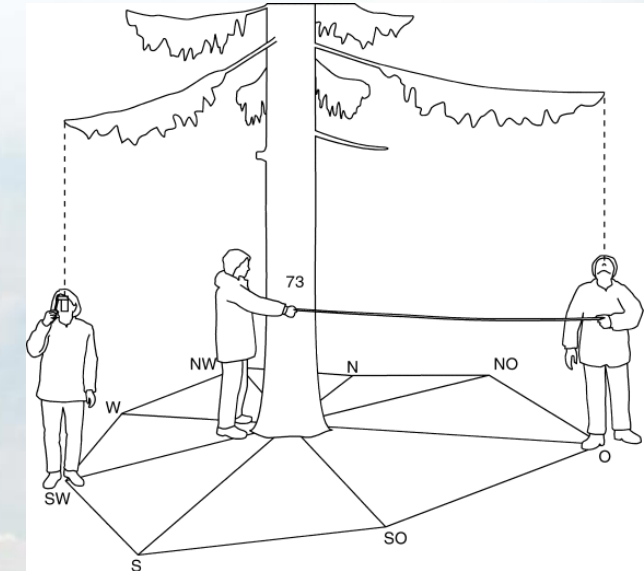
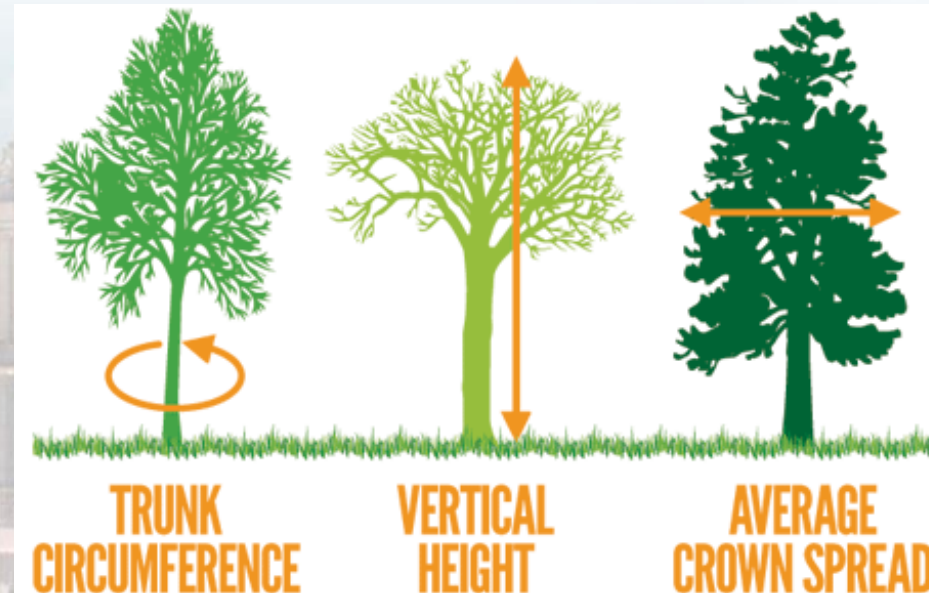
**But, how to deal with it?**



# 1. Introduction

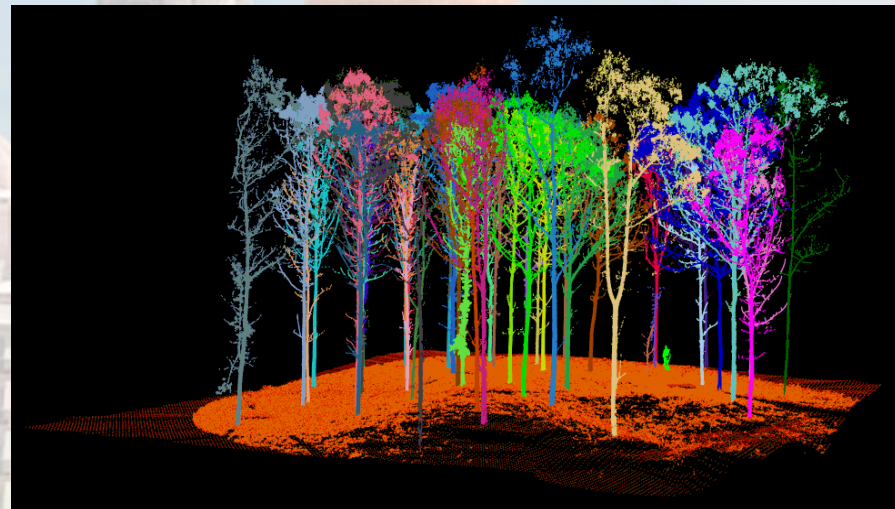
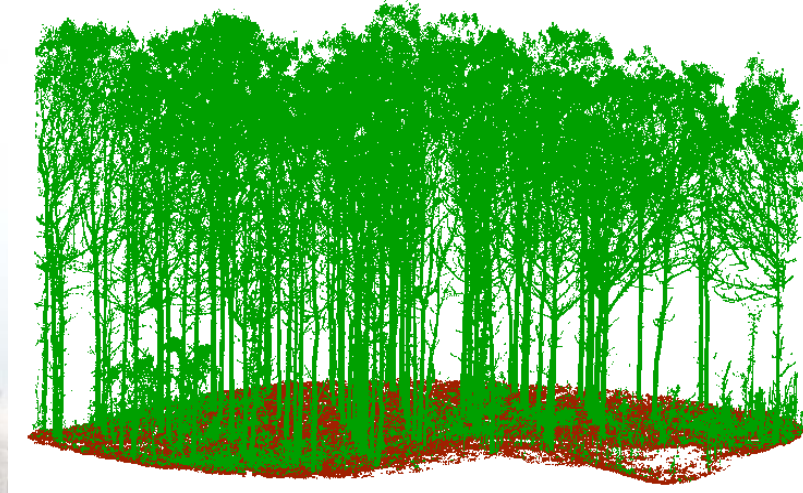
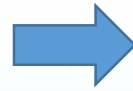


**2/3 of the program costs** within the context of the US Forest Service Forest Inventory were for field work on the reference plots (Gulding, 2000).





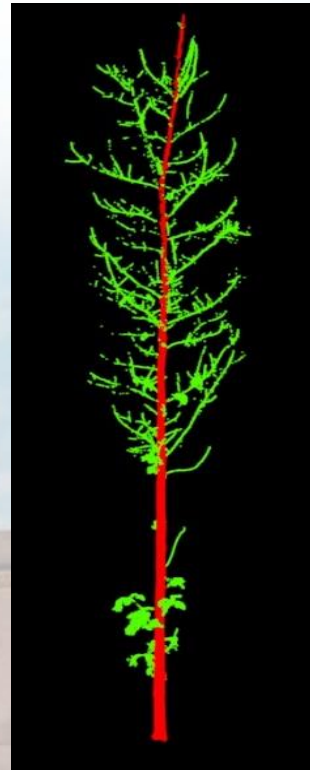
# 1. Introduction



Individual Tree  
Detection (ITD)

Individual Tree  
Segmentation (ITS)

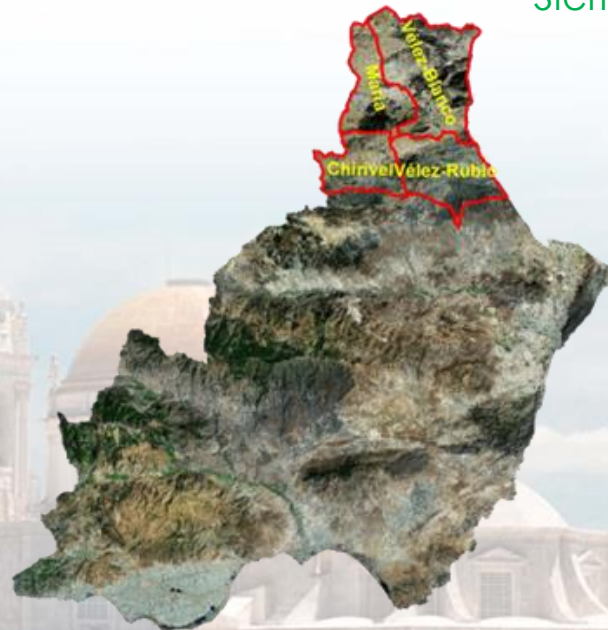
Individual Tree Crown  
Delineation (ITCD)





## 2. Materials and Methods

### Study Site 1



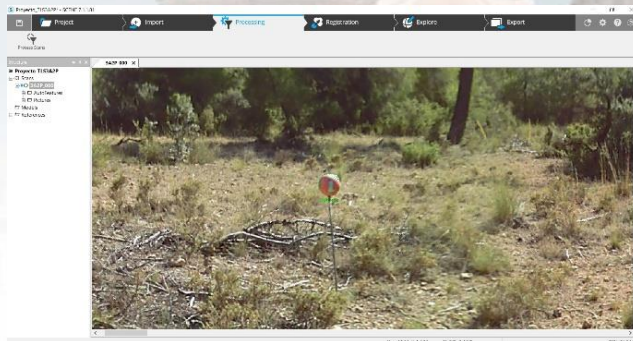
Sierra de Maria-Los Vélez Natural Park





## 2. Materials and Methods

### Study Site 1





## 2. Materials and Methods

### Study Site 2



Province of León:  
1. Toral de los Vados  
2. Villasabariego  
3. Villamañan



## 2. Materials and Methods

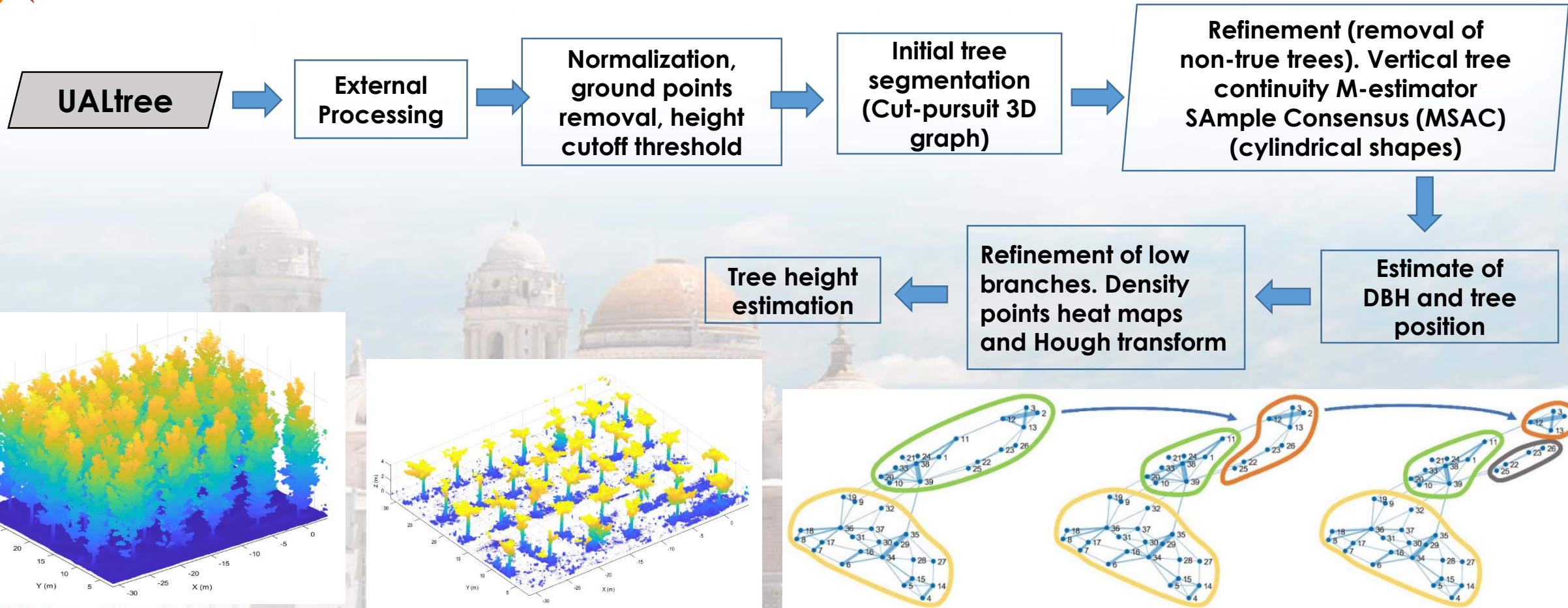
### Study Site 2





## 2. Materials and Methods

### Field Data and Processing Methods



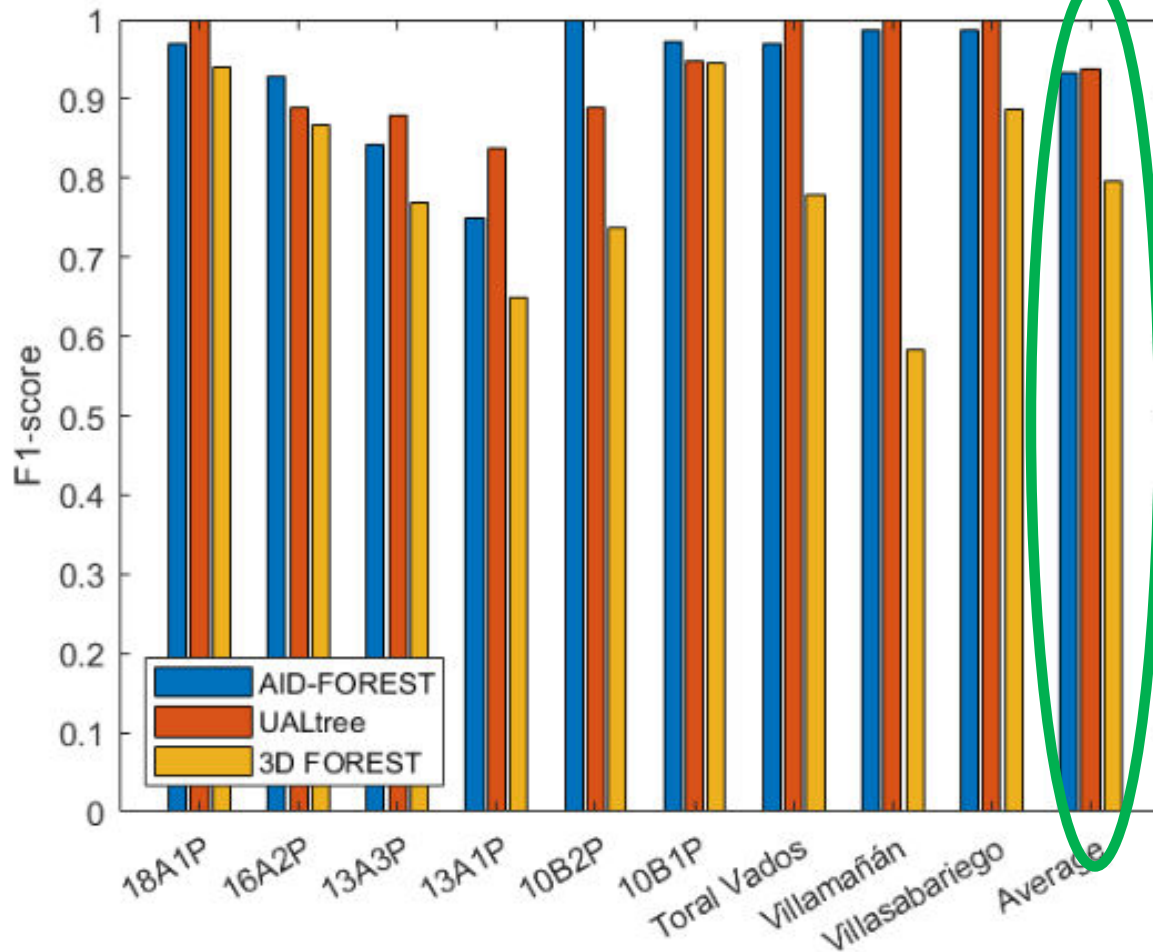


# 3. Results

## Tree detection accuracy

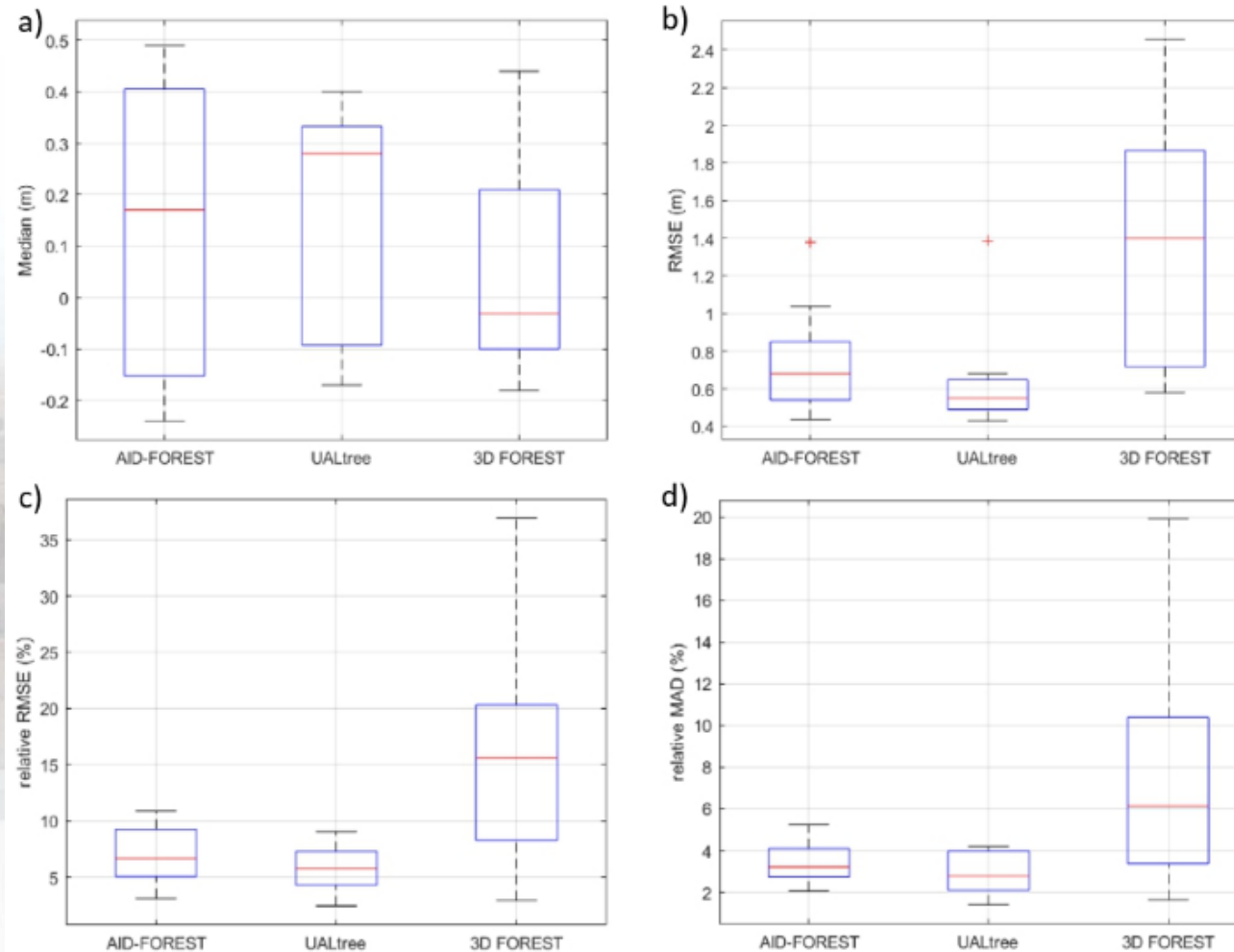
Tree detection assessment in terms of F1-score:

- AID-FOREST
- UALtree
- 3D FOREST





# 3. Results Accuracy of tree height estimation





# 3. Results

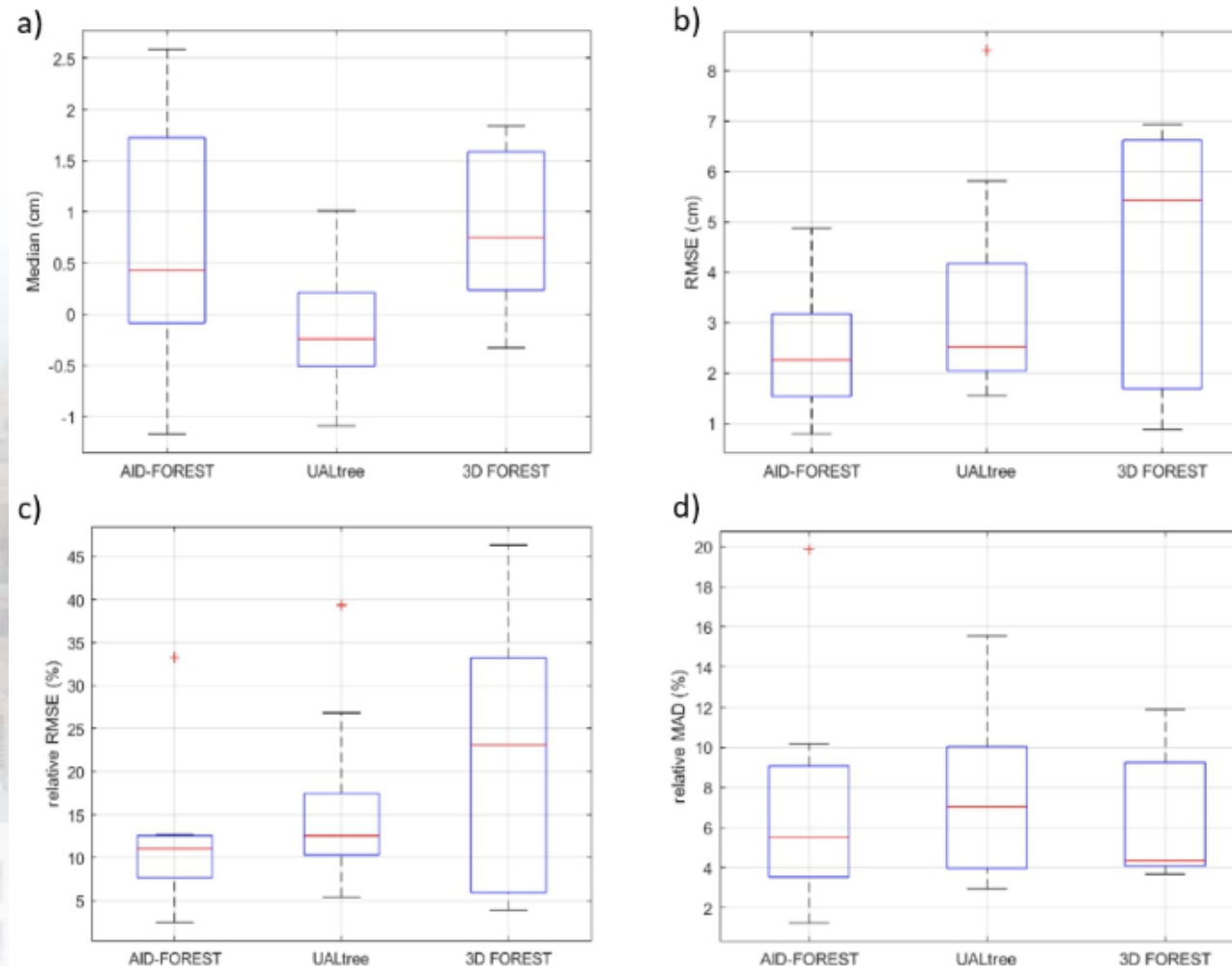
## Tree detection accuracy

- ▲ **Table 2.** Wilcoxon (W-test) applied to tree height (H) extracted using AID-FOREST and UALtree. N = sample size (matched trees in each plot). The average results  $\pm$  standard deviation are reported. (\*) Significant differences ( $p < 0.05$ ) between estimated and observed measurements at tree level.

Plot	AID-FOREST				UALtree			
	N	$H_{estimated} (m)$ Average	$H_{observed} (m)$ Average	W-test p-value	N	$H_{estimated} (m)$ Average	$H_{observed} (m)$ Average	W-test p-value
18A1P	32	7.09 $\pm$ 1.03	7.09 $\pm$ 1.03	0.083	34	<b>7.04 <math>\pm</math> 1.15</b>	<b>6.95 <math>\pm</math> 1.17</b>	<b>0.010*</b>
16A2P	26	<b>6.84 <math>\pm</math> 1.10</b>	<b>6.41 <math>\pm</math> 1.13</b>	<b>0.0001*</b>	24	<b>6.82 <math>\pm</math> 1.00</b>	<b>6.45 <math>\pm</math> 1.08</b>	<b>0.0003*</b>
13A3P	16	6.55 $\pm$ 1.07	6.34 $\pm$ 1.04	0.255	18	<b>6.51 <math>\pm</math> 1.23</b>	<b>6.19 <math>\pm</math> 1.15</b>	<b>0.003*</b>
13A1P	15	<b>5.52 <math>\pm</math> 1.04</b>	<b>5.09 <math>\pm</math> 1.09</b>	<b>0.003*</b>	18	<b>5.58 <math>\pm</math> 1.17</b>	<b>5.21 <math>\pm</math> 1.14</b>	<b>0.0001*</b>
10B2P	8	11.60 $\pm$ 1.23	11.83 $\pm$ 1.18	0.779	8	11.90 $\pm$ 1.14	11.83 $\pm$ 1.18	0.575
10B1P	18	10.51 $\pm$ 1.06	10.40 $\pm$ 1.03	0.371	18	10.62 $\pm$ 0.90	10.40 $\pm$ 1.03	0.077
Toral Vados	33	22.35 $\pm$ 1.25	22.54 $\pm$ 1.42	0.075	33	<b>22.39 <math>\pm</math> 1.10</b>	<b>22.54 <math>\pm</math> 1.42</b>	<b>0.028*</b>
Villamañán	34	<b>23.85 <math>\pm</math> 1.15</b>	<b>24.43 <math>\pm</math> 1.73</b>	<b>0.037*</b>	34	23.90 $\pm$ 1.12	24.43 $\pm$ 1.73	0.092
Villasabariego	34	19.16 $\pm$ 3.20	19.53 $\pm$ 2.80	0.069	34	19.36 $\pm$ 2.72	19.53 $\pm$ 2.80	0.205

# 3. Results

Accuracy of DBH estimation





### 3. Results Accuracy of DBH estimation

**Table 3.** Wilcoxon (W-test) applied to Diameter at Breast Height (DBH) extracted using AID-FOREST and UALtree. N = sample size (matched trees in each plot). The average results  $\pm$  standard deviation are reported. (\*) Significant differences ( $p < 0.05$ ) between estimated and observed measurements at tree level.

Plot	AID-FOREST				UALtree			
	N	<u>DBH<sub>estimated</sub></u> (cm) Average	<u>DBH<sub>observed</sub></u> (cm) Average	W-test p-value	N	<u>DBH<sub>estimated</sub></u> (cm) Average	<u>DBH<sub>observed</sub></u> (cm) Average	W-test p-value
18A1P	32	20.00 $\pm$ 3.75	19.77 $\pm$ 4.21	0.389	34	19.36 $\pm$ 4.68	19.41 $\pm$ 4.46	0.063
16A2P	26	17.12 $\pm$ 2.03	16.73 $\pm$ 2.74	0.533	24	17.04 $\pm$ 3.60	16.81 $\pm$ 2.82	0.977
13A3P	16	19.00 $\pm$ 3.62	18.07 $\pm$ 4.46	0.062	18	<b>18.32 <math>\pm</math> 5.35</b>	<b>17.48 <math>\pm</math> 4.89</b>	<b>0.047*</b>
13A1P	15	<b>17.63 <math>\pm</math> 3.34</b>	<b>14.63 <math>\pm</math> 4.43</b>	<b>0.010*</b>	18	16.47 $\pm$ 6.82	14.76 $\pm$ 4.56	0.285
10B2P	8	<b>30.81 <math>\pm</math> 5.09</b>	<b>31.35 <math>\pm</math> 5.22</b>	<b>0.049*</b>	8	28.48 $\pm$ 9.25	31.35 $\pm$ 5.22	0.779
10B1P	18	<b>26.67 <math>\pm</math> 5.97</b>	<b>28.69 <math>\pm</math> 6.18</b>	<b>0.0005*</b>	18	28.55 $\pm$ 6.24	28.69 $\pm$ 6.18	0.285
<u>Toral Vados</u>	33	<b>26.23 <math>\pm</math> 2.52</b>	<b>23.77 <math>\pm</math> 2.41</b>	<b>0.0000*</b>	33	23.31 $\pm$ 3.71	23.77 $\pm$ 2.41	0.126
<u>Villamañán</u>	34	<b>30.46 <math>\pm</math> 2.75</b>	<b>29.10 <math>\pm</math> 2.39</b>	<b>0.002*</b>	34	<b>28.22 <math>\pm</math> 3.24</b>	<b>29.10 <math>\pm</math> 2.39</b>	<b>0.041*</b>
<u>Villasabariego</u>	34	<b>21.28 <math>\pm</math> 4.12</b>	<b>20.69 <math>\pm</math> 4.12</b>	<b>0.001*</b>	34	<b>19.36 <math>\pm</math> 4.47</b>	<b>20.69 <math>\pm</math> 4.12</b>	<b>0.0000*</b>

# 3. Results

## Processing time

**Table 4.** Processing time expressed in seconds per million points processed. All calculations were executed with Intel® Core™ i7-8565U 4 X 1.99 GHz, 16 GB RAM, and NVIDIA Quadro P520.

<i>Method</i>	<i>18A1P</i>	<i>16A2P</i>	<i>13A3P</i>	<i>13A1P</i>	<i>10B2P</i>	<i>10B1P</i>	<i>Toral Vados</i>	<i>Villamañán</i>	<i>Villasabariego</i>	<i>Average</i>
AID-FOREST	32.69	85.08	93.50	119.73	94.58	46.80	24.93	52.81	75.68	69.53
UALtree	12.07	18.96	29.09	39.68	28.82	23.96	17.67	12.11	19.75	22.46
3D FOREST	156.29	135.50	134.30	132.38	119.34	179.82	370.32	301.10	397.54	214.07



## 4. Conclusions

- ❑ UALtree proved to perform similarly to AID-FOREST in terms of F1-score (tree detection). Both algorithms yielded much better accuracy rates than 3D FOREST.
- ❑ AID-FOREST and UALtree provided similar figures regarding tree height and DBH estimation. Both better than 3D FOREST.
- ❑ Wilcoxon non-parametric test evidenced that UALtree was able to estimate the observed DBH distribution slightly better than AID-FOREST.
- ❑ UALtree worked faster.

Thank you very much for your kind attention

