

# A Thorough Accuracy Estimation of DTM Produced From Airborne Full-Waveform Laser Scanning Data of Unmanaged Eucalypt Plantations

Gil Gonçalves and Luísa Gomes Pereira

**Abstract**—A digital terrain model (DTM) is needed to extract forest variables. Its accuracy affects that of these variables, and thus it has to be known. Airborne laser scanning (ALS) is being increasingly used to produce DTM. The potential of full-waveform ALS (FWALS), however, for DTM production in forest areas has not been thoroughly evaluated, particularly under adverse conditions of heterogeneous forest stands, such as the typically extensively managed eucalypt plantations of north-central Portugal studied here. In the study area of 900 ha, an exhaustive assessment of FWALS-derived DTM accuracy was carried out based on the field survey of 43 plots. In each of these circular plots of roughly 400 m<sup>2</sup>, the coordinates were measured for all trees and all prominent terrain points using Global Navigation Satellite Systems receivers and total station. In addition, the quality of the FWALS data was assessed by surveying 1 m × 1 m grids of points on three bare areas. The accuracy assessment measures are computed by assuming normal and nonnormal distributions of the differences between the h values of FWALS terrain points and those of control points at the same planimetric locations. While the accuracy of the FWALS data is as expected, around 3 cm, that of the DTM of the forested area exceeds expectations. In fact, the DTM accuracy of 12 cm is higher than that obtained in prior ALS studies carried out in less adverse conditions than those studied here. Thus, the current FWALS-based methodology to derive DTM appears highly suitable to extracting forest variables.

**Index Terms**—Author, please supply index terms/keywords for your paper. To download the IEEE Taxonomy go to [http://www.ieee.org/documents/2009Taxonomy\\_v101.pdf](http://www.ieee.org/documents/2009Taxonomy_v101.pdf).

## I. INTRODUCTION

ONE OF the indispensable conditions for a sustainable development is a detailed and up-to-date knowledge of the available natural resources. In the case of Portugal, an important natural resource is that of its forests. Crucial information for foresters and land-use planners concerns, in particular, tree species, tree position, number of trees, mean height, volume, and biomass per hectare. In relation to forest fires—which have attained dramatic proportions in Portugal in the last few years—the need for additional information on vegetation variables like shrub density and height, as well as on terrain topography must

be stressed. Such information is fundamental for determining the risk of ignition and for predicting fire dynamics.

In Portugal, the technological delay has hampered the development of a suitable methodology for a fast and reliable survey of forestry data, adapted to new technologies and techniques. The majority of the forest inventories are done in situ by specialized teams. This makes them expensive, and their quality is greatly correlated with the experience of the teams. The relief is modeled by the so-called digital terrain models (DTM) produced by photogrammetric means. In forested areas, photogrammetry is rarely adequate for DTM production because the trees hide the terrain underneath. Airborne laser scanning (ALS), better known as light detection and range, has shown itself to be an alternative/complementary technique to photogrammetry for the production of high-resolution DTM, particularly in forested areas. Furthermore, it also allows the estimation of shrub density and height as well as tree variables such as height, localization, and mean tree crown and volume [1]. Hence, the errors in the DTM will propagate to the normalized digital surface model and, therefore, to the derivation of the forest inventory and fuel variables, like tree and shrub heights.

The vertical accuracy of a DTM produced with ALS data is influenced by the errors/inaccuracies originated from the [1]–[4]:

- ALS system (global positioning system (GPS), inertial navigation system, and laser scanner);
- strategy for acquisition of ALS data (point density, first/last pulse, several pulses, flight height, scan angle);
- conversion of ALS data (full-waveform modeling, filtering and interpolation methods);
- characteristics of the target surface, that is, topography and type(s), density, and spatial distribution of vegetation.

While a general understanding of the accuracy of the ALS systems has been achieved, the accuracy of the derived DTM from ALS data in a forest environment has not been thoroughly evaluated [3], [5], mainly in unmanaged eucalypt forests. As far as we know, there are only few studies concerning eucalypt forest, like those of [6], [7], the later using even regular eucalypt plantations, in which the methods for quality assessment are not fully described. Furthermore, while in the recommendations of the article of [1], it is said that the extraction of DTM for forest areas is well established. This conclusion is based on a list of works using other forests than eucalypt forests. Moreover, by digitizing and recording the complete echo waveform, the full-waveform ALS (FWALS) systems have, in comparison

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TABLE I  
PUBLISHED ACCURACY VALUES OF DTM PRODUCED WITH ALS DATA (n.r = not reported)

Authors	Area extent (ha)	ALS density (pts/m <sup>2</sup> )	No. of check pts.	RMSE ; bias (cm)	Land cover
Kraus and Pfeifer (1998), [9]	9 100	0.1	466	57; 20	Beech.
Hyypya et al. (2000), [10]	1.4	08-10	740	22 (Std)	Boreal forest. Norway spruce and Scots pine.
Reutebuch et al. (2003), [5]	500	4	347	32; 22	Coniferous forest (Douglas-fir in western Washington region).
Wack et al. (2003) [7]	0.48	5	602	19 (Std); 14	Eucalypt forest (Portugal)
Hodgson and Bresnahan (2004), [3]	200 000	0.25	654	21; n.r	Pavement, low grass, high grass, brush/low trees, evergreen and deciduous forests.
Takahashi et al. (2005), [11]	2 areas: 250 000 and 62 500	11.2	283	39; 14	Sugi plantations ( <i>Cryptomeria japonica</i> D. Don)
Yu et al. (2005) [13]	8 areas of 100	10	1 474	9 (Std) 30 (Std)	Boreal forest (Kalkkinen) Norway spruce and Scots pine.
Gonçalves-Seco et al. (2006) [6]	281	7.8	2207	0.19; 0.16	Eucalypt forest (Spain)
Hollaus et al. (2006) [14]	12 800	1.8	2 200	10 <sup>1</sup> ; n.r 50 <sup>2</sup> ; n.r	ALS data along road sides.
Su and Bork (2006) [4]	2 700	0.75	256	59; 2	Riparian Meadows, Upland Grasslands, Shrublands and Aspen Forest.
Evans and Hudak (2007) [15]	2 areas of 85 000	0.3	165 39	70; -70 62; 60	Coniferous Forest

<sup>1</sup>For flat areas

<sup>2</sup>For sloped areas (>60°)

89 to the conventional real-time processing multi-echo systems,  
90 the advantage of permitting echo detection in postprocessing.  
91 This makes the ranging determination process more robust.  
92 While this is expected to lead to higher accuracy of the derived  
93 distances and thus to a more accurate DTM [8], it remains to be  
94 proved in areas with the characteristics of the one here studied.  
95 Table I lists some published accuracy values of DTM cre-  
96 ated by using ALS data of several forests/plantations ([3]–[7],  
97 [9]–[15]). A limitation of the studies is that they do not allow  
98 one to predict the DTM accuracy because they do not address  
99 all the four accuracy-influencing factors above listed.

100 In the context of a research project for estimating forest  
101 inventory parameters and fuel variables under eucalypt stands  
102 in Mediterranean climates, the vertical precision of the DTM  
103 obtained by automatic filtering of FWALS data had to be  
104 evaluated. In this paper, it is thus assessed the vertical accuracy  
105 of a DTM obtained by automatic filtering of FWALS data of  
106 a eucalypt forest. In addition to the peculiar characteristics of  
107 the study area, which are often avoided in related studies, the  
108 accuracy estimation is carried out in a novel way. By novel way,  
109 it meant an exhaustive, well-planned collection of reliable con-  
110 trol data under a eucalypt forest comprising regular as well as  
111 irregular spacing plantations. The understory is heterogeneous  
112 and is mainly composed of whin, fern, heath, and gorse. Prior  
113 to the vertical accuracy assessment of the DTM, the quality of  
114 the delivered laser data was also evaluated. To this end, mea-  
115 surements on horizontal and inclined bare surfaces were used.

## II. STUDY AREA AND DATA ACQUISITION

116

### A. Study Area

117

The study area was selected nearby the city of Águeda, in the  
118 district of Aveiro, situated in the Northern part of Portugal. The  
119 selected area measures 900 ha [Fig. 1(a)]. Its topography varies  
120 from gentle (2.5%) to steep slopes (34.2%), with altitudes  
121 varying from 27 to 162 m [Fig. 1(b)]. The area is dominated by  
122 eucalypt plantations, but it also includes some pine stands and  
123 few built-up areas. The mean tree density is around 1600 trees  
124 per hectare. The forest stands in the area have regular as well as  
125 irregular spaced plantations, both even and uneven-aged stands,  
126 and stands with and without extensive undergrowth [Fig. 1(c)].  
127

### B. Acquisition of the FWALS and Image Data

128

The FWALS data were acquired on the 14th of July 2008. The  
129 laser system utilized was the Litmapper 5600. This system has,  
130 as main hardware components the high-resolution laser scanner  
131 from RIEGL, the LMS-Q560 with full-waveform processing, the  
132 AEROcontrol GPS/IMU system and the DigiCAM, a medium-  
133 format Airborne Digital Camera System operated simultane-  
134 ously with the FWALS system. The AEROcontrol system is  
135 utilized for the precise determination of position and attitude  
136 with an inertial measurement unit (IMU) of 256 Hz raw data rate  
137 and a differential GPS. The main system software relates to  
138 the AEROoffice—the GPS and the IMU data postprocessing  
139

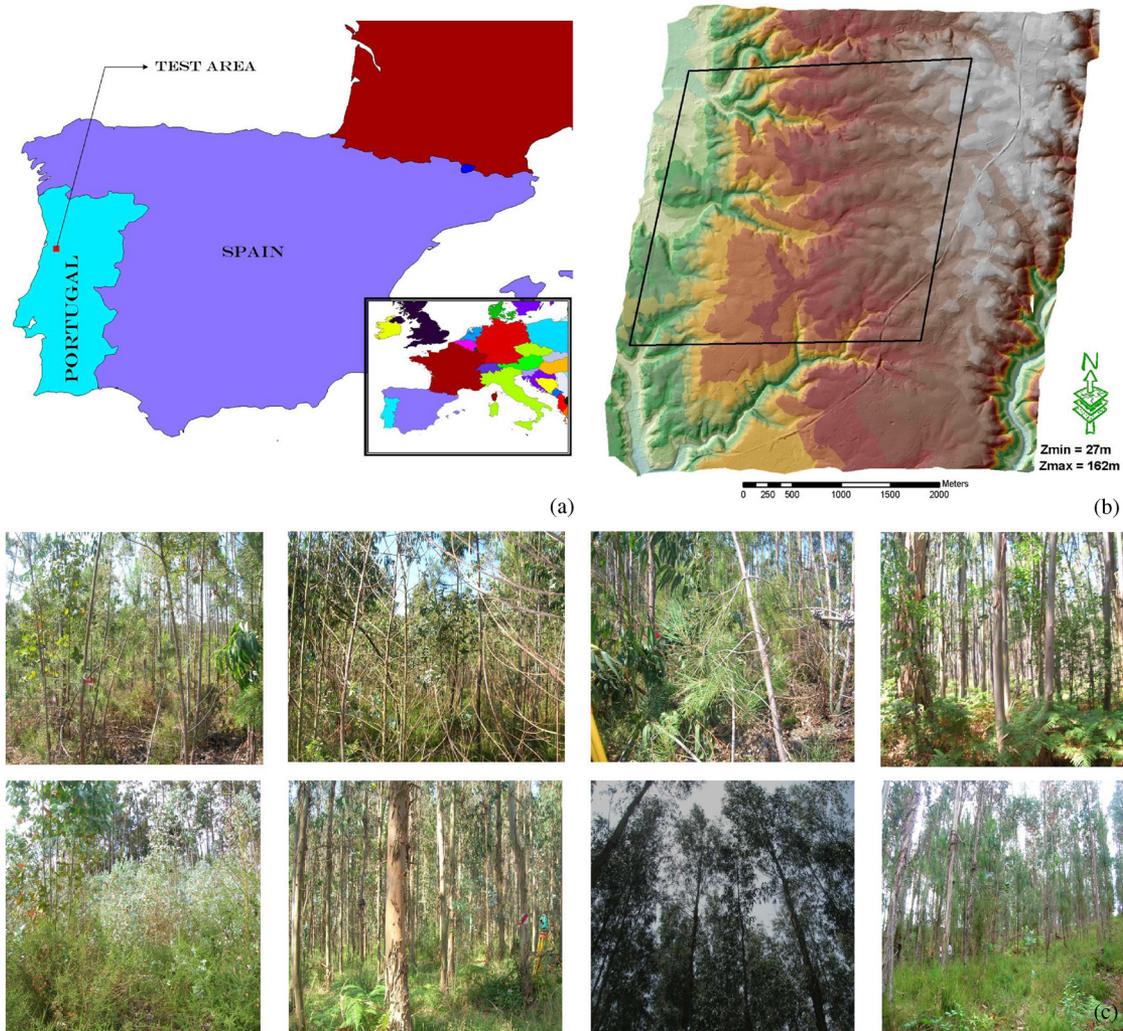


Fig. 1. (a) Localization of study area within Portugal and its delimitation. (b) DTM of the study area. (c) Examples of vegetation within study area.

140 software—the Riegl software, RiWorld for coordinate trans-  
 141 formation and RiAnalyze for processing of the FWALS data,  
 142 and the software bundle from TerraSolid. This comprises the  
 143 software modules TerraScan, TerraModeler, and TerraMatch.  
 144 TerraScan is the dedicated software solution for manipulation,  
 145 viewing, and classifying FWALS data. TerraModeler is a full-  
 146 featured terrain modeling application for creating digital terrain/  
 147 surface models (DTM/DSM). TerraMatch is used for automated  
 148 correction of ALS data ([http://www.igi-systems.com/products/](http://www.igi-systems.com/products/litemapper/components.htm)  
 149 [litemapper/components.htm](http://www.igi-systems.com/products/litemapper/components.htm), 19.01.2010).

150 Stereopairs of CIR images were acquired simultaneously  
 151 with the FWALS data. The parameters selected for FWALS  
 152 and image data acquisition are listed in Table II. The flight is  
 153 constituted by thirty plus two flight lines, the latter in a direction  
 154 perpendicular to the first and used for calibration purposes.

### 155 III. METHODS

#### 156 A. Procedures for Acquisition and Quality Assessment of 157 Reference Data

158 Reference data are needed to be verified, in terms of precision  
 159 and reliability, the DTM produced by means of the laser data  
 160 and a filtering algorithm (Section III-B-2). The strategy for

the reference data collection was not straightforward. In forest  
 161 areas, the collection of these data is time consuming, mainly in  
 162 plots with a high density of shrubs and trees. Furthermore, be-  
 163 cause the data were georeferenced, geodetic Global Navigation  
 164 Satellite Systems (GNSS) receivers had to be used. 165

The planning of the topographic survey was based on that of  
 166 the forest inventory, which started by selecting 43 plots, with  
 167 a radius of 11.28 m, within the study area [dots in Fig. 2(a)].  
 168 The DTM was represented by the coordinates of terrain points  
 169 located aside trees, which give also the locations of the trees,  
 170 and by the coordinates of prominent terrain points, like those on  
 171 breaklines [Fig. 2(b)]. The characterization of each plot in the  
 172 study area, in terms of mean slope (Section III-C), shrubbery  
 173 (height varying between 10 cm and 2 m), and high vegetation  
 174 (height > 2 m) is shown in Fig. 3. 175

In addition to the DTM quality assessment, it was also de-  
 176 cided to assess the quality of the delivered FWALS data. To this  
 177 end, it was measured points, in a grid format of approximately  
 178  $1 \times 1 \text{ m}^2$ , on bare surfaces [roads and fields; Fig. 4(a)–(c)]. 179

This information was collected by means of a topographic  
 180 survey using the radial method [16]. The coordinate system in  
 181 which the FWALS and image data were collected (WGS84-  
 182 UTM29) uses the reference ellipsoid of the World Geodetic  
 183

TABLE II  
FWALS DATA AND IMAGE ACQUISITION PARAMETERS

FWALS	Image
FWALS sensor: Riegl LMS-Q560	Camera sensor: Digicam H39 + 50 mm focal length
Wavelength: 1550 nm	Wavelengths in CIR mode (nm): B=500– 620; G=580– 800; R=800– 1000
Scan angle: 45°	Image repetition rate: 1.9 sec
Pulse rate: 150 kHz	
Effective Measurement rate: 75 kHz	
Beam divergence: 0.5 mrad	
Ground speed: 46.26 m/s	
Flying height above terrain: 600 m	Flying height above terrain: 600 m
Swath: 497 m	Overlap: 60%
Sidelap: 70%	Sidelap: 30%
Single run density of laser shots: 3.3 pts/m <sup>2</sup>	Nr. of pixels forward: 7 216
Expected final density: 9.9 pts/m <sup>2</sup>	Nr. of pixels sideward: 5 412
Distance between lines: 150 m	
Spot diameter: 30 cm	Ground sampling distance (GSD): 8.2 cm

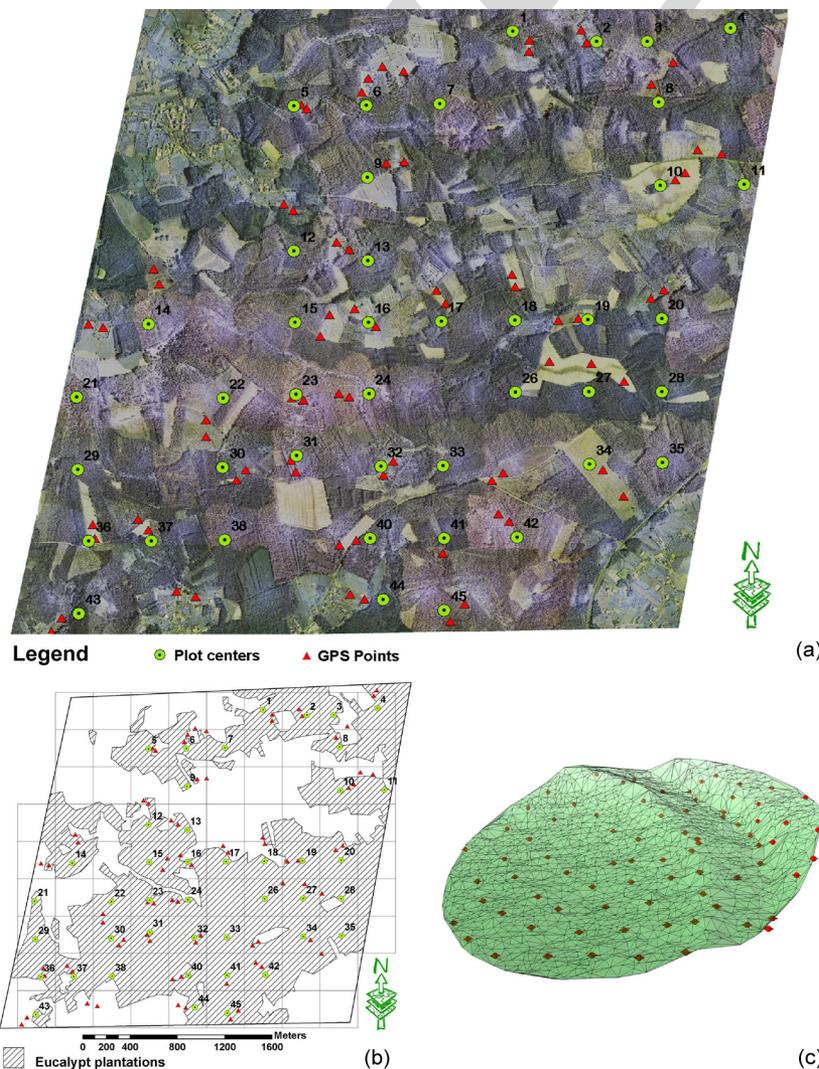


Fig. 2. Planning of the topographical survey: (a) and (b) location of the 43 forest plots; (c) DTM of plot no. 43 with breaklines, produced with the FWALS points classified as terrain. The small circles represent the points measured by topographic means.

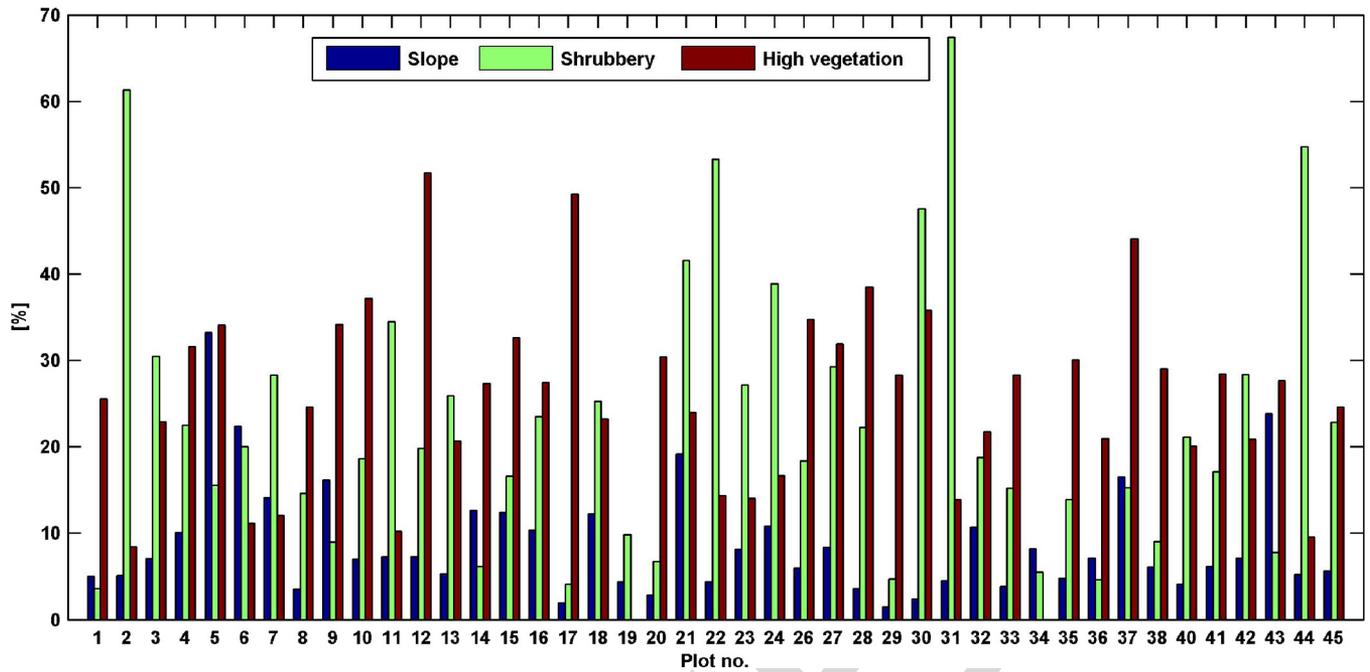


Fig. 3. Characterization of each plot in the study area.

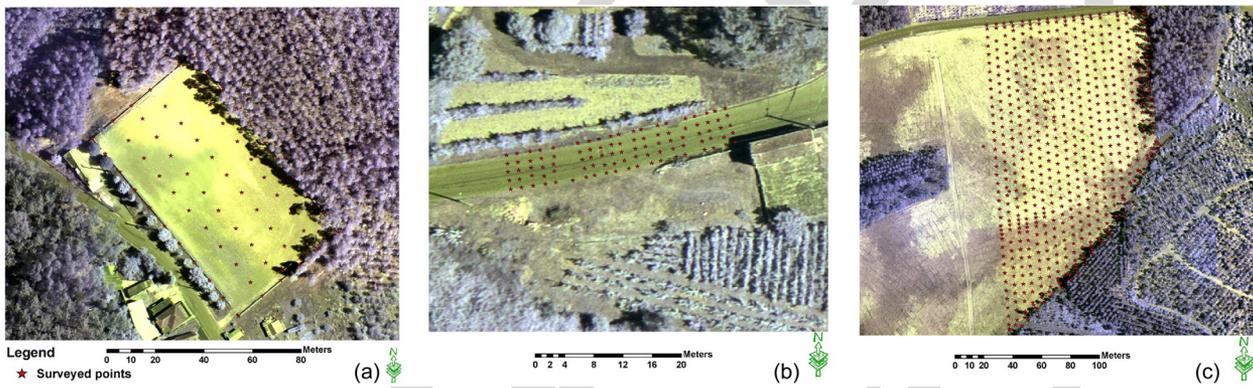


Fig. 4. Reference data for FWALS quality assessment. (a), (b), and (c) surveyed points on bare surfaces (paved road, football field and arable field, respectively).

184 System (WGS 84) and the cartographic projection Universal  
 185 Transverse Mercator (UTM) zone 29, for X and Y coordinates.  
 186 For the Z coordinate, the WGS84 ellipsoidal height was used.  
 187 These coordinates will be from now on referred to as absolute  
 188 coordinates X, Y, and h. Because this is not a local system,  
 189 the geographic information collected in the field had to be  
 190 converted to that system by using the GPS. To this end, it was  
 191 decided to attach to each plot two GPS-derived points, named  
 192 *GPS base* (triangles in Fig. 2(a)), whose coordinates were  
 193 measured with two GNSS receivers. They allow coordinating  
 194 the surveyed points directly in the referred coordinate system.  
 195 Two points are needed to orient the total station. These two  
 196 points are placed as close as possible to the plot and as much as  
 197 possible in an opened space (Fig. 5). This criterion turned out  
 198 to be difficult to fulfil in the study area.

199 The method used to measure the coordinates of the two GPS-  
 200 derived points was the relative positioning by using a fixed receiver  
 201 on a geodetic pillar with known coordinates on the WGS84-  
 202 UTM29. This method, in postprocessing, is the most precise  
 203 and may reach levels of precision in the order of the cm [17].

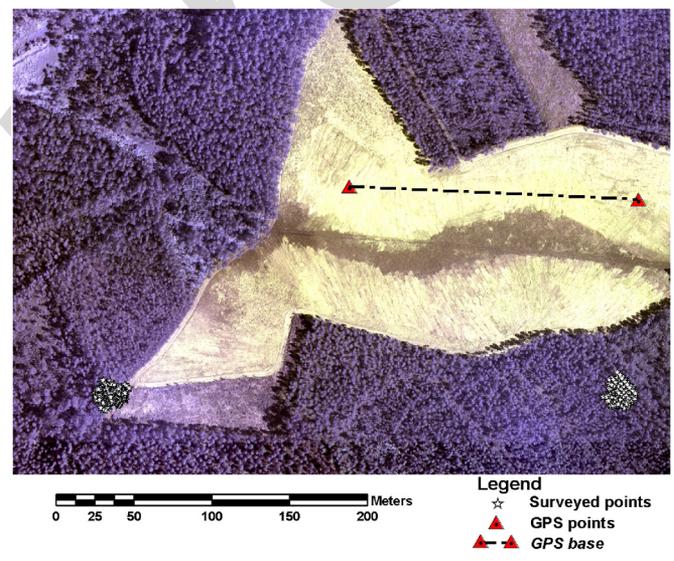


Fig. 5. Example of a *GPS base* used to georeference the measurements on two plots.

204 The GPS observations were collected according to a previous  
 205 planning dictated by the following specifications: minimum  
 206 time of observation of 60 min, depending on the point surround-  
 207 ings (normally it took 120 min), minimum number of seven  
 208 satellites, and value of the position dilution of precision less  
 209 than 3. In total, the coordinates of 82 points, i.e., 41 *GPS bases*  
 210 were measured. The number of *GPS bases* does not coincide  
 211 with that of plots because two *GPS bases* were assigned to two  
 212 plots each.

213 As some of the GPS-derived points of the *GPS bases* had  
 214 nearby some trees that might have degraded the GPS signal, we  
 215 developed a strategy to control the quality of their coordinates.  
 216 This strategy is based on the distances and height differences  
 217 between the two points of the *GPS bases* computed in two  
 218 coordinate systems. One is a local system, and the other the  
 219 absolute system (*WGS84 – UTM29*). Those variables are  
 220 compared to each other. It is assumed that the variables obtained  
 221 by topographic methods, using a total station, are error free. It  
 222 was decided that if the differences in the distance values and in  
 223 the height differences were bigger than 6 cm, the coordinates  
 224 of the GPS-derived points had to be remeasured by stationing  
 225 once again the GPS on the points. The mean error and RMSE of  
 226 the differences in distance are 1.7 cm and 2.5 cm, respectively,  
 227 while the mean error and RMSE of the differences in height  
 228 differences have the values of 1.9 mm and 2.6 cm, respectively.

## 229 B. FWALS Data Processing

230 1) *Full-Waveform Data Preparation*: The full-waveform  
 231 laser data were purchased already processed by the company  
 232 that acquired the data. The processing was done with the  
 233 RiAnalyze software from Riegl. The basic processing is based  
 234 on a Gauss pulse estimation technique. This aims to combine  
 235 the execution time of a simple center of gravity approach  
 236 ([18]) with the accuracy of Gaussian pulse fitting technique and  
 237 provides multiple returns ([19]). A maximum of five returns  
 238 were obtained for all the FWALS echoes.

239 The strips of FWALS data were also adjusted and geo-  
 240 referenced by the referred company using the TerraMatch  
 241 and the RiWorld softwares and were stored in the form of  
 242 LAS blocks. TerraMatch compares overlapping laser strips  
 243 with each other and corrects their orientation to obtain  
 244 the best fit and improved accuracy of the laser data. The  
 245 adjustment is based on measured differences between the  
 246 xyz shape or the intensity of laser points from differ-  
 247 ent strips (<http://www.terrasolid.fi/system/files/TerraMatch.pdf>,  
 248 23.12.2010). With the TerraMatch software, the misalignment  
 249 between the navigation and the laser systems is estimated with  
 250 calibration strips. The misalignment is then corrected when  
 251 georeferencing the cloud of points with the RiWorld software.  
 252 The TerraMatch software is once again used to adjust the height  
 253 values between points in overlapping strips by smoothing GPS  
 254 fluctuations. Finally, the data are verified for systematic errors  
 255 by comparing the h values of ground control points with those  
 256 of laser points.

257 2) *Point Cloud Filtering*: The point cloud obtained by pro-  
 258 cessing the full-waveform data contains both laser returns from  
 259 objects located on the ground (buildings, trees, understory

vegetation) as the ground itself. Moreover, in vegetated areas  
 the laser pulse can penetrate the canopy, travelling between the  
 branches and leaves and be reflected by the understory (e.g.,  
 subcanopy trees and shrubs) and by the ground.

Since we are interested in the reconstruction of the terrain  
 surface, it is necessary to remove the nonground returns from  
 the point cloud. Removing nonground measurements from the  
 point cloud is here referred to as filtering. In order to find the  
 most probable ground returns from the initial point cloud, the  
 filtering process uses the last return of each pulse.

Points in a point cloud are classified as either ground or non-  
 ground returns using a progressive triangular irregular network  
 (TIN) densification algorithm implemented in the TerraScan  
 software v4.006 ([www.terrasolid.fi/en](http://www.terrasolid.fi/en), 11.05.2010) and based  
 on the algorithm published in [20], [21]. This algorithm begins  
 by constructing a sparse TIN with the lowest point inside a  
 square neighborhood of a certain size. The TIN is densified by  
 adding points of the set of unclassified points that satisfy certain  
 criteria. The densifying of the TIN stops when no more unclas-  
 sified points satisfy the criteria or if a certain ground point den-  
 sity is achieved. An additional threshold may be used to control  
 the maximum terrain slope allowed at each triangle of the TIN  
 surface. In this paper, the parameters used for filtering the laser  
 point cloud were tuned empirically and happen to coincide with  
 the default values ([22]) with exception of the neighborhood  
 size parameter that has the value of 10 m (the default is 60 m  
 and therefore too big when compared to each plot size).

## C. DTM Accuracy Assessment

In addition to the accuracy assessment of the DTM produced  
 by the filtering process of the laser data, the contribution of  
 the quality of the FWALS data alone into that assessment was  
 verified. To this end, three DTM were produced, by using topo-  
 graphic means, on flat bare surfaces (a paved road, a football  
 field, and an arable field; see Section III-A). The accuracy  
 assessment of these DTM as well as the DTM produced after  
 the filtering process of the laser data relates to the estimation  
 of the mean and RMSE of the differences (dh) between the  
 h values of FWALS terrain points and those of the reference  
 points (Section III-A) at the same planimetric locations. These  
 FWALS terrain points are interpolated from a TIN computed  
 with the filtered FWALS terrain points (Section III-B-2). The  
 TIN was created from the filtered FWALS data instead of from  
 the reference data because the density of FWALS terrain points  
 is higher than that of reference points (3.8 versus 0.2 pts/m<sup>2</sup>).  
 This means that many points on the terrain presented in the  
 FWALS filtered data are not present in the reference data.  
 Therefore, it is more appropriate to interpolate the laser data  
 to the planimetric positions of the reference data. The TIN  
 format is also chosen, instead of the grid format, to handle the  
 breaklines. Furthermore, the interpolation methods associated to  
 this format are computationally simpler when compared, for  
 example, with the kriging method, at expenses of being more  
 sensitive to measurements errors.

The deviation of the distribution of the residuals dh from the  
 normal distribution implies the use of other accuracy estimators  
 than the mean and RMSE ([23], [24]). Therefore, that deviation

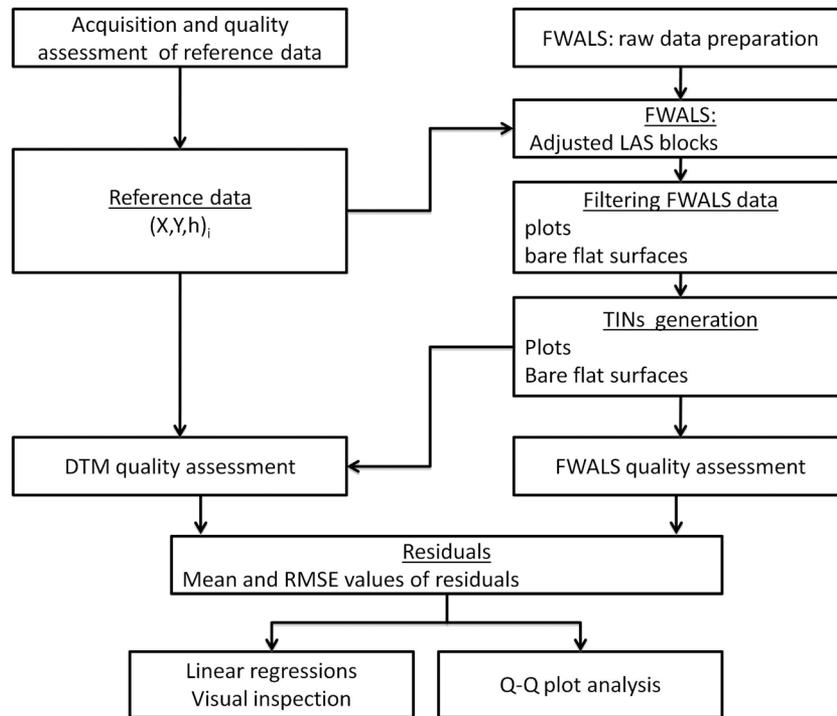


Fig. 6. Concise presentation of the developed methodology.

316 has to be evaluated. This is done using two different methods. 317 One method superimposes on the histogram of the residuals a 318 curve for a normal distribution obtained by ordinary estimation 319 of mean error and variance. If the Gaussian curve does not fit 320 the data very well, it means that the residuals are not originating 321 from a normal distribution, e.g., because the distribution is 322 not symmetric around its mean or because the distribution is 323 more peaked around its mean than the normal distribution while 324 having heavier tails [24]. These effects are measured by the 325 skewness and kurtosis of the distribution. The analysis of the 326 so-called quantile-quantile (Q-Q) plot is another method for 327 verifying a deviation from the normal distribution. The quan- 328 tiles of the empirical distribution function are plotted against 329 the theoretical quantiles of the normal distribution. If the distri- 330 bution is normal, the Q-Q plot should yield a straight line. As 331 a robust and distribution free approach handling outliers and 332 non-normal distribution, [24] suggest as accuracy measures the 333 median  $m_{dh}$  of residuals, the normalized median absolute de- 334 viation ( $NMAD = 1.4826 \cdot median_j(|dh_j - m_{dh}|)$ ), and the 335 68.3% and 95% quantiles of the absolute value of residuals.

336 It was also verified if the quality of the DTM has a relation 337 with the vertical structure of the vegetation and with the ter- 338 rain slope. Therefore, a linear regression analysis between the 339 quality of the DTM produced by filtering the laser data and 340 the vertical structure of the vegetation was computed. Linear 341 regression was carried out between the values estimated for the 342 mean and the RMSE of residuals per plot and, respectively, the 343 percentages of the number of points classified by the filtering 344 process as terrain, as shrubbery with height varying from 10 cm 345 until 2 m and as higher vegetation cover.

346 The linear regression analysis between the quality of the 347 referred DTM and the mean slope of the terrain within plots 348 was also performed. Linear regression was carried out between

the values estimated for the RMSE of residuals per plot and 349 the mean slope within each plot. The mean slope was estimated 350 by fitting a plane to the reference terrain points. The plane is 351 in fact the geometric shape that represents best the geometry 352 of all the terrain surfaces within the plots. A relation between 353 the RMSE values and the spatial distribution of the plots as 354 well as the spatial point density was also studied by means of 355 visual inspection. The spatial point density was computed for 356  $1 \times 1 \text{ m}^2$  cell using only the last returns [25]. For those plots 357 on the same flight line, the associated RMSE values were also 358 inspected for a relation. 359

The developed methodology is summarized in the form of a 360 flow diagram in Fig. 6. 361

#### IV. RESULTS

362 The quality of the DTM obtained with the laser data was 363 computed for each plot individually, by using the strategy 364 described above. The estimated values for the mean, standard 365 deviation, and RMSE are plotted in Fig. 7. As it may be seen in 366 that figure, the values of the mean of residuals vary, in absolute 367 value, between 5 and 28 cm, while the RMSE varies between 4 368 and 29 cm. When considering all the plots together, i.e., 3174 369 points, the value of the mean of residuals is 8 cm, while the 370 RMSE is 15 cm. These are very good values considering the 371 cover characteristic within the plots. Furthermore, there has 372 been no manual editing, i.e., manual removal of gross errors. 373

The bias of 8 cm shows that the FWALS filtered points are 374 above the terrain surface. This may result from the vegetation 375 cover within the plots. Nonetheless, this value is quite low 376 considering the characteristics of the plots (Fig. 3), which 377 proves that the strategy for laser data acquisition and processing 378 is effective. 379

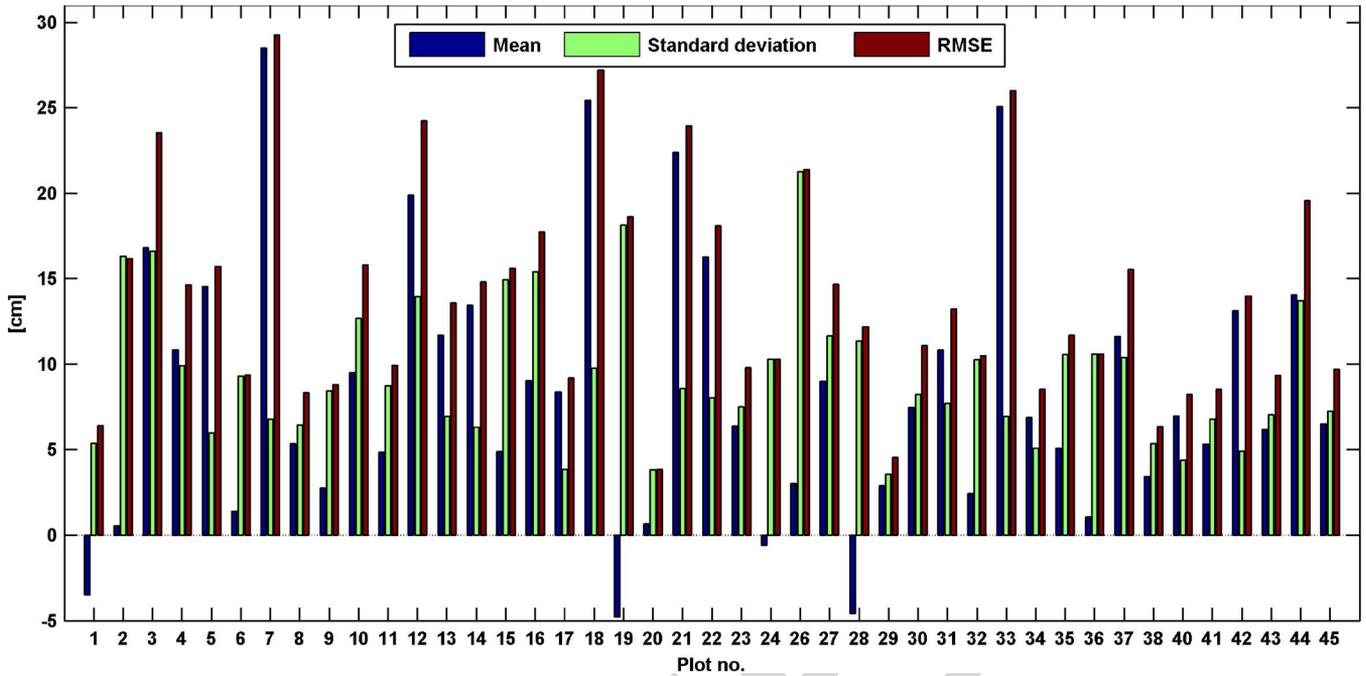


Fig. 7. Values of the mean, standard deviation, and RMSE of residuals per plot. The plots 19 and 34 are the only ones in open areas. They have young eucalypt plantations with mean heights smaller than 30 and 20 cm, respectively. Although situated in an open area, plot 19 has a bigger RMSE because of the terrain relief (it contains a breakline).

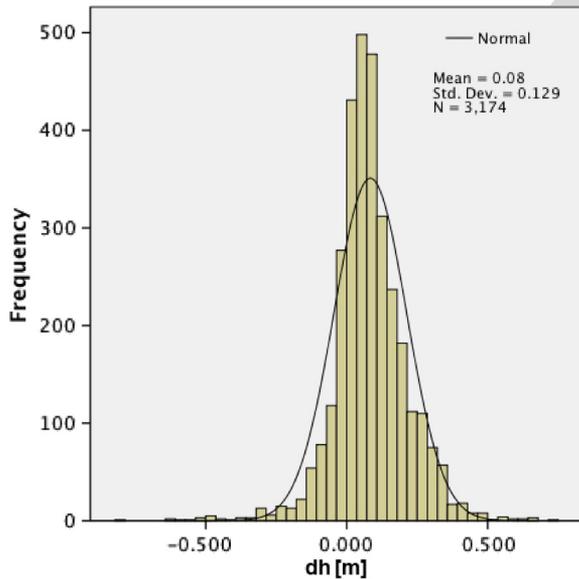


Fig. 8. Histogram of the residuals. Superimposed on the histogram are the expected counts from a normal distribution with mean and standard deviation estimated from the data.

380 The gross errors present in the results (min = -79 cm;  
 381 max = 72 cm) deviate from the distribution of the residuals  
 382 from a normal distribution. Fig. 8 shows the histogram of  
 383 the residuals on which is superimposed a curve for a normal  
 384 distribution obtained by ordinary estimation of mean error and  
 385 variance. As mentioned in Section III-C, the deviation from nor-  
 386 mality is measured by the skewness and kurtosis of the distribu-  
 387 tion which, in this case, have the values of  $-0.177$  (std. error =  
 388  $0.043$ ) and of  $3.979$  (std. error =  $0.087$ ), respectively.  
 389 The analysis of the so-called Q-Q plot provides another way  
 390 of verifying a deviation from the normal distribution. Fig. 9

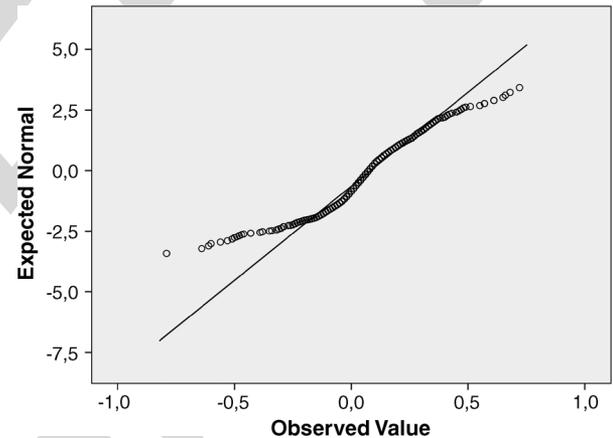


Fig. 9. Quantile-quantile plot of the distribution of residuals computed to assess the DTM accuracy.

TABLE III  
 ROBUST ACCURACY ESTIMATORS

Accuracy estimator	Value (cm)
$m_{dh}$	7
NMAD	10
68.3% Quantile	13
95% Quantile	32
RRMSE	12

shows the Q-Q plot for the distribution of the residuals. A  
 391 deviation from a straight line may be seen at both extremes,  
 392 which indicates that the distribution of the residuals is not  
 393 normal. The obtained robust accuracy measures (Section III-C)  
 394 are tabulated in Table III. Although the distribution of the  
 395 residuals deviate from the normal distribution, the differences  
 396

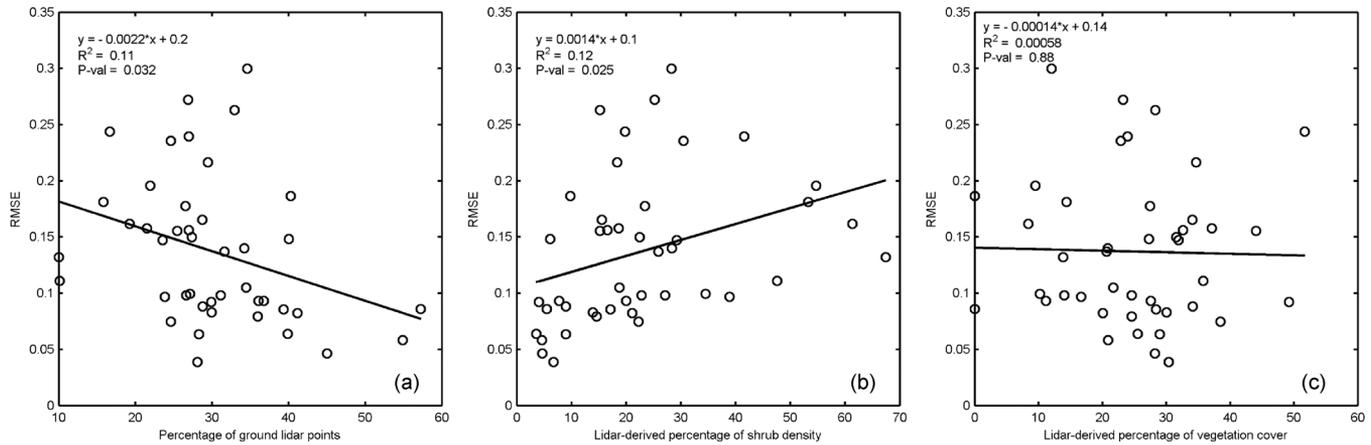


Fig. 10. Linear regression between the value of RMSE and the process-derived percentage of. (a) Terrain points. (b) Shrubberty. (c) Higher ( $\geq 2$  m) vegetation cover per plot.

397 between the robust accuracy measures and those estimated  
 398 by assuming a normal distribution of the residuals are not  
 399 significant. The difference between the values of the standard  
 400 deviation of residuals and of the  $NMAD$  is only about 3 cm.  
 401 If one takes the robust measures  $m_{dh}$  and  $NMAD$  as values  
 402 representative for the mean and standard deviation of residuals,  
 403 a robust estimator for the RMSE value (RRMSE), estimated as  
 404  $\sqrt{m_{dh} + NMAD^2}$ , is 12 cm (Table III).

405 The existence of a linear relation between the quality of the  
 406 DTM and the vertical structure of the vegetation and the mean  
 407 slope of the terrain in each plot was studied by means of linear  
 408 regression. It is expected that this relation exists. Fig. 10(a)–(c)  
 409 shows the linear regression between the values estimated for  
 410 the RMSE per plot and, respectively, the percentages of the  
 411 number of points classified by the filtering process as terrain,  
 412 as shrubberty with height varying from 10 cm until 2 m and as  
 413 high vegetation cover (i.e., with height  $> 2$  m).

414 Fig. 11 shows linear regression between the values estimated  
 415 for the RMSE per plot and the mean terrain slope within each  
 416 plot.

417 Contrary to what was expected, the low values obtained  
 418 for the coefficient of determination ( $R^2_{(a)} = 0.08$ ;  $R^2_{(b)} = 0.13$ ;  
 419  $R^2_{(c)} = 0.10$ ) indicate that there is no linear relation between  
 420 the quality of the DTM and the vertical structure of vegetation.  
 421 In fact, there is no relation judging by the point dispersion  
 422 characteristics shown in Fig. 10. The same tests and results  
 423 were obtained when using the mean of the residuals per plot  
 424 instead of the RMSE. In relation to the characteristics of plots  
 425 in terms of vegetation cover, this means that the algorithm  
 426 published in [21], together with the high density of the laser  
 427 points per  $m^2$  (in mean 10), is quite robust and insensitive to  
 428 the obstacles on the terrain. In relation to the terrain slope, the  
 429 results show that there is no relation between the slope and  
 430 the RMSE of residuals, which contradict those found in other  
 431 studies [12] but also support results from others [5].

432 The spatial distribution of the RMSE values per plot, the  
 433 spatial point density and the flight lines, within the study area  
 434 is shown in Fig. 12. As in previous studies [5], there seems  
 435 to be no relation between the RMSE values and the spatial  
 436 distribution of the plots. There appears to be also no relation  
 437 between the RMSE values and the spatial point density. This is

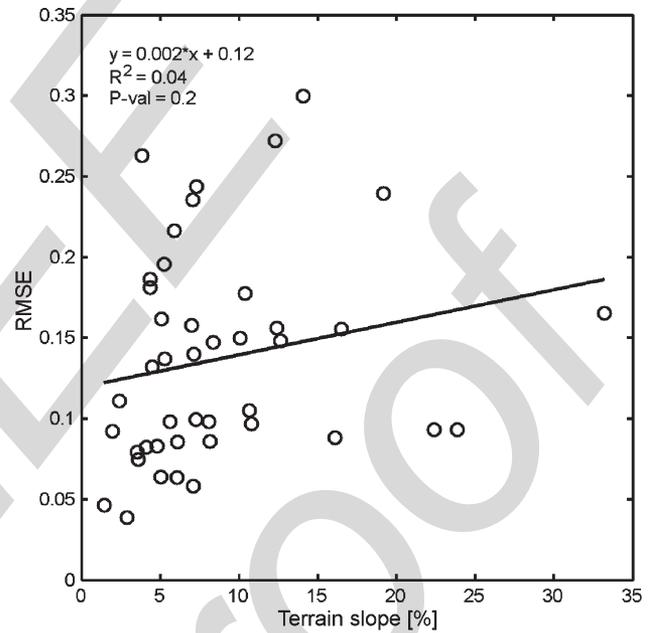


Fig. 11. Linear regression between the values of RMSE and the mean terrain slope per plot.

not surprising once the mean point density is high. In Fig. 12, it  
 438 may be also seen that the RMSE values for plots situated along  
 439 the same strip have no relation among them. Because the strips  
 440 have been adjusted, and thus systematic errors removed, it may  
 441 be stated that there is as well no relation between the RMSE  
 442 values and bias strip differences.

443  
 444 In order to verify the contribution of the quality of the  
 445 FWALS data alone into the assessment of the quality of the  
 446 DTM, the quality of the laser data was verified, as described  
 447 in Section III-C. The accuracy measures, i.e., the mean and  
 448 the RMSE of residuals are used instead of the robust esti-  
 449 mators aforementioned because the distribution of residuals is  
 450 considered normal (skewness and kurtosis equal to 0.067 and  
 451 0.268, respectively). See Fig. 13 representing the Q-Q plot). The  
 452 computed mean and RMSE are listed in Table IV. These values  
 453 are quite good and similar to those obtained in previous studies  
 454 ([26]). The worst is obtained for the arable land and results from  
 455 the fact that the pole used for measuring sunk into the fresh

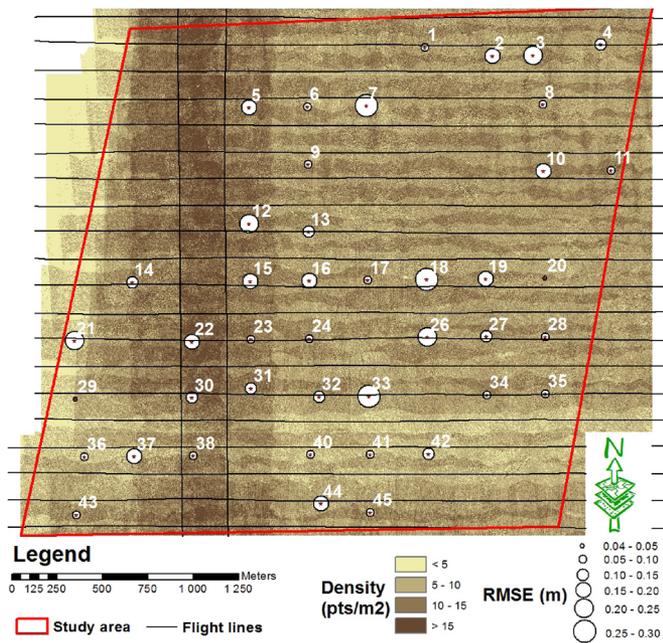


Fig. 12. Spatial distribution of the RMSE value per plot.

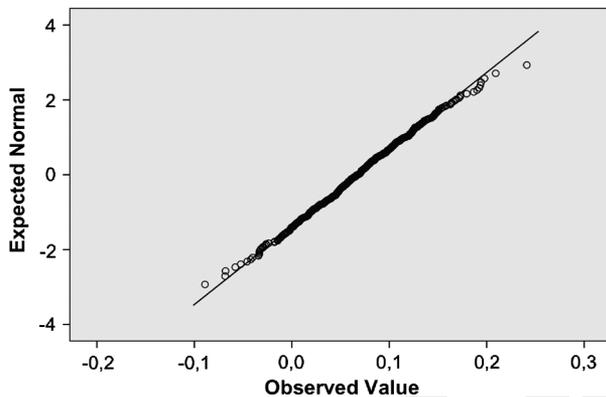


Fig. 13. Quantile-quantile plot of the distribution of residuals computed to assess the accuracy of the delivered laser data.

TABLE IV  
MEAN AND RMSE VALUES OF RESIDUALS, PER SURFACE

	Mean (cm)	RMSE (cm)
Paved road	2.8	3.2
Football field	1.8	3.4
Arable field	6.8	8.4

456 soil for a few centimeters. This pole is used to transport the  
457 prism needed to measure the coordinates of the DTM points by  
458 topographic means. There is a bias that shows that the FWALS  
459 filtered points are above the terrain surface. Nonetheless, this  
460 bias value is insignificant for the majority of applications.

461 In addition to this evaluation of the accuracy of the FWALS  
462 data, the mean and standard deviation of the residuals between  
463 the  $h$  values of the *GPS bases* and those derived from the laser  
464 data at the same location were computed. A mean of 0.3 cm and  
465 a RMSE of 6.8 cm were obtained. This mean value of residuals

is quite different from those computed in flat surfaces because  
466 while the *GPS bases* are spread all over the study area, and thus  
467 involving all the 30 laser strips, the other use only seven laser  
468 strips, five of which are common. 469

## V. DISCUSSION AND CONCLUSION 470

A study as the one here presented, carried out in an area of  
471 900 ha with unmanaged eucalypt forest with great amount of  
472 high shrubbery, chaotic tree distribution, and high tree density,  
473 is quite demanding. It requires an exhaustive and well-planned  
474 collection of reliable control data to properly assess the quality  
475 of FWALS data and of a derived DTM. The georeferencing  
476 of these data makes use of GNSS techniques, which are very  
477 time consuming in environments where a high density of high  
478 vegetation predominates. This is because under high vegetation,  
479 the GNSS receiver has to be stationed on the GPS-derived  
480 points for long periods of time (the georeferencing of the  
481 GPS-derived points took approximately 170 h). Furthermore,  
482 topographic measurements in plots covered with high shrubs  
483 are also difficult to execute. 484

The accuracy of the produced DTM is very good and suitable  
485 for a great number of applications. The value of 12 cm in RMSE  
486 is smaller than that published in several studies carried out in  
487 less chaotic forests (Table I; [1]). It also compares to the RMSE  
488 value claimed 12 years ago in [2], but for flat surfaces, and  
489 still referred in several studies. For similar surfaces, a value  
490 of 3 cm in RMSE was instead obtained. Although such a high  
491 DTM accuracy is not needed per se for this type of applications,  
492 such as forest inventory, acquiring very dense point clouds is  
493 recommended to describe the crown characteristics accurately,  
494 including total tree height ([1]). 495

Considering that the outgoing pulse width is 4 ns, the  
496 expected echo separation is 60 cm. Nonetheless, the full-  
497 waveform data processing algorithm detected 93 pairs of con-  
498 secutive echoes with a separation smaller than 60 cm, among  
499 the 40 523 recorded pairs of consecutive echoes. It may be said  
500 that there is a slight improvement in echo separation, due to the  
501 postprocessing wave analysis, but it remains to be verified if all  
502 of the 93 pairs of echoes really relate to reflections. 503

The filtering algorithm based on the filter of [21] also proved  
504 to be very robust to terrain with the aforesaid characteristics.  
505 The low values obtained for the coefficients of determination  
506  $R^2$  between the RMSE values per plot and the percentages  
507 of the number of points classified by the filtering process as  
508 terrain, as shrubbery with height varying from 10 cm until 2  
509 m and as higher vegetation cover, indicate that there is not  
510 a relation between the quality of the DTM and the vertical  
511 structure of vegetation. This means that the Axelsson algorithm,  
512 together with a high density of the laser points per  $m^2$  (in  
513 mean 10), is quite robust and insensitive to the obstacles on  
514 the terrain. The spatial distribution of the RMSE values per plot  
515 shows also no relation between the RMSE values and the spatial  
516 distribution of the plots as well as the spatial point density.  
517 The same conclusion was attained for the relation between the  
518 RMSE values and the mean terrain slope. Because this is not  
519 what was expected, further studies are needed. 520

521 To conclude, the FWALS system proved to be adequate, in  
522 terms of data accuracy, for the production of DTM in areas of  
523 unmanaged high-density eucalypt forest. The accuracy figure  
524 of 12 cm for the DTM produced in such areas, using the data  
525 acquisition parameters listed in Table II and the Axelsson fil-  
526 tering algorithm [21], is also acceptable to derive forest canopy  
527 structure metrics.

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# A Thorough Accuracy Estimation of DTM Produced From Airborne Full-Waveform Laser Scanning Data of Unmanaged Eucalypt Plantations

Gil Gonçalves and Luísa Gomes Pereira

**Abstract**—A digital terrain model (DTM) is needed to extract forest variables. Its accuracy affects that of these variables, and thus it has to be known. Airborne laser scanning (ALS) is being increasingly used to produce DTM. The potential of full-waveform ALS (FWALS), however, for DTM production in forest areas has not been thoroughly evaluated, particularly under adverse conditions of heterogeneous forest stands, such as the typically extensively managed eucalypt plantations of north-central Portugal studied here. In the study area of 900 ha, an exhaustive assessment of FWALS-derived DTM accuracy was carried out based on the field survey of 43 plots. In each of these circular plots of roughly 400 m<sup>2</sup>, the coordinates were measured for all trees and all prominent terrain points using Global Navigation Satellite Systems receivers and total station. In addition, the quality of the FWALS data was assessed by surveying 1 m × 1 m grids of points on three bare areas. The accuracy assessment measures are computed by assuming normal and nonnormal distributions of the differences between the h values of FWALS terrain points and those of control points at the same planimetric locations. While the accuracy of the FWALS data is as expected, around 3 cm, that of the DTM of the forested area exceeds expectations. In fact, the DTM accuracy of 12 cm is higher than that obtained in prior ALS studies carried out in less adverse conditions than those studied here. Thus, the current FWALS-based methodology to derive DTM appears highly suitable to extracting forest variables.

**Index Terms**—Author, please supply index terms/keywords for your paper. To download the IEEE Taxonomy go to [http://www.ieee.org/documents/2009Taxonomy\\_v101.pdf](http://www.ieee.org/documents/2009Taxonomy_v101.pdf).

## I. INTRODUCTION

ONE OF the indispensable conditions for a sustainable development is a detailed and up-to-date knowledge of the available natural resources. In the case of Portugal, an important natural resource is that of its forests. Crucial information for foresters and land-use planners concerns, in particular, tree species, tree position, number of trees, mean height, volume, and biomass per hectare. In relation to forest fires—which have attained dramatic proportions in Portugal in the last few years—the need for additional information on vegetation variables like shrub density and height, as well as on terrain topography must

be stressed. Such information is fundamental for determining the risk of ignition and for predicting fire dynamics.

In Portugal, the technological delay has hampered the development of a suitable methodology for a fast and reliable survey of forestry data, adapted to new technologies and techniques. The majority of the forest inventories are done in situ by specialized teams. This makes them expensive, and their quality is greatly correlated with the experience of the teams. The relief is modeled by the so-called digital terrain models (DTM) produced by photogrammetric means. In forested areas, photogrammetry is rarely adequate for DTM production because the trees hide the terrain underneath. Airborne laser scanning (ALS), better known as light detection and range, has shown itself to be an alternative/complementary technique to photogrammetry for the production of high-resolution DTM, particularly in forested areas. Furthermore, it also allows the estimation of shrub density and height as well as tree variables such as height, localization, and mean tree crown and volume [1]. Hence, the errors in the DTM will propagate to the normalized digital surface model and, therefore, to the derivation of the forest inventory and fuel variables, like tree and shrub heights.

The vertical accuracy of a DTM produced with ALS data is influenced by the errors/inaccuracies originated from the [1]–[4]:

- ALS system (global positioning system (GPS), inertial navigation system, and laser scanner);
- strategy for acquisition of ALS data (point density, first/last pulse, several pulses, flight height, scan angle);
- conversion of ALS data (full-waveform modeling, filtering and interpolation methods);
- characteristics of the target surface, that is, topography and type(s), density, and spatial distribution of vegetation.

While a general understanding of the accuracy of the ALS systems has been achieved, the accuracy of the derived DTM from ALS data in a forest environment has not been thoroughly evaluated [3], [5], mainly in unmanaged eucalypt forests. As far as we know, there are only few studies concerning eucalypt forest, like those of [6], [7], the later using even regular eucalypt plantations, in which the methods for quality assessment are not fully described. Furthermore, while in the recommendations of the article of [1], it is said that the extraction of DTM for forest areas is well established. This conclusion is based on a list of works using other forests than eucalypt forests. Moreover, by digitizing and recording the complete echo waveform, the full-waveform ALS (FWALS) systems have, in comparison

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TABLE I  
PUBLISHED ACCURACY VALUES OF DTM PRODUCED WITH ALS DATA (n.r = not reported)

Authors	Area extent (ha)	ALS density (pts/m <sup>2</sup> )	No. of check pts.	RMSE ; bias (cm)	Land cover
Kraus and Pfeifer (1998), [9]	9 100	0.1	466	57; 20	Beech.
Hyypya et al. (2000), [10]	1.4	08-10	740	22 (Std)	Boreal forest. Norway spruce and Scots pine.
Reutebuch et al. (2003), [5]	500	4	347	32; 22	Coniferous forest (Douglas-fir in western Washington region).
Wack et al. (2003) [7]	0.48	5	602	19 (Std); 14	Eucalypt forest (Portugal)
Hodgson and Bresnahan (2004), [3]	200 000	0.25	654	21; n.r	Pavement, low grass, high grass, brush/low trees, evergreen and deciduous forests.
Takahashi et al. (2005), [11]	2 areas: 250 000 and 62 500	11.2	283	39; 14	Sugi plantations ( <i>Cryptomeria japonica</i> D. Don)
Yu et al. (2005) [13]	8 areas of 100	10	1 474	9 (Std) 30 (Std)	Boreal forest (Kalkkinen) Norway spruce and Scots pine.
Gonçalves-Seco et al. (2006) [6]	281	7.8	2207	0.19; 0.16	Eucalypt forest (Spain)
Hollaus et al. (2006) [14]	12 800	1.8	2 200	10 <sup>1</sup> ; n.r 50 <sup>2</sup> ; n.r	ALS data along road sides.
Su and Bork (2006) [4]	2 700	0.75	256	59; 2	Riparian Meadows, Upland Grasslands, Shrublands and Aspen Forest.
Evans and Hudak (2007) [15]	2 areas of 85 000	0.3	165 39	70; -70 62; 60	Coniferous Forest

<sup>1</sup>For flat areas

<sup>2</sup>For sloped areas (>60°)

89 to the conventional real-time processing multi-echo systems,  
90 the advantage of permitting echo detection in postprocessing.  
91 This makes the ranging determination process more robust.  
92 While this is expected to lead to higher accuracy of the derived  
93 distances and thus to a more accurate DTM [8], it remains to be  
94 proved in areas with the characteristics of the one here studied.  
95 Table I lists some published accuracy values of DTM cre-  
96 ated by using ALS data of several forests/plantations ([3]–[7],  
97 [9]–[15]). A limitation of the studies is that they do not allow  
98 one to predict the DTM accuracy because they do not address  
99 all the four accuracy-influencing factors above listed.

100 In the context of a research project for estimating forest  
101 inventory parameters and fuel variables under eucalypt stands  
102 in Mediterranean climates, the vertical precision of the DTM  
103 obtained by automatic filtering of FWALS data had to be  
104 evaluated. In this paper, it is thus assessed the vertical accuracy  
105 of a DTM obtained by automatic filtering of FWALS data of  
106 a eucalypt forest. In addition to the peculiar characteristics of  
107 the study area, which are often avoided in related studies, the  
108 accuracy estimation is carried out in a novel way. By novel way,  
109 it meant an exhaustive, well-planned collection of reliable con-  
110 trol data under a eucalypt forest comprising regular as well as  
111 irregular spacing plantations. The understory is heterogeneous  
112 and is mainly composed of whin, fern, heath, and gorse. Prior  
113 to the vertical accuracy assessment of the DTM, the quality of  
114 the delivered laser data was also evaluated. To this end, mea-  
115 surements on horizontal and inclined bare surfaces were used.

## II. STUDY AREA AND DATA ACQUISITION

116

### A. Study Area

117

The study area was selected nearby the city of Águeda, in the  
118 district of Aveiro, situated in the Northern part of Portugal. The  
119 selected area measures 900 ha [Fig. 1(a)]. Its topography varies  
120 from gentle (2.5%) to steep slopes (34.2%), with altitudes  
121 varying from 27 to 162 m [Fig. 1(b)]. The area is dominated by  
122 eucalypt plantations, but it also includes some pine stands and  
123 few built-up areas. The mean tree density is around 1600 trees  
124 per hectare. The forest stands in the area have regular as well as  
125 irregular spaced plantations, both even and uneven-aged stands,  
126 and stands with and without extensive undergrowth [Fig. 1(c)].  
127

### B. Acquisition of the FWALS and Image Data

128

The FWALS data were acquired on the 14th of July 2008. The  
129 laser system utilized was the Litmapper 5600. This system has,  
130 as main hardware components the high-resolution laser scanner  
131 from RIEGL, the LMS-Q560 with full-waveform processing, the  
132 AEROcontrol GPS/IMU system and the DigiCAM, a medium-  
133 format Airborne Digital Camera System operated simultane-  
134 ously with the FWALS system. The AEROcontrol system is  
135 utilized for the precise determination of position and attitude  
136 with an inertial measurement unit (IMU) of 256 Hz raw data rate  
137 and a differential GPS. The main system software relates to  
138 the AEROoffice—the GPS and the IMU data postprocessing  
139

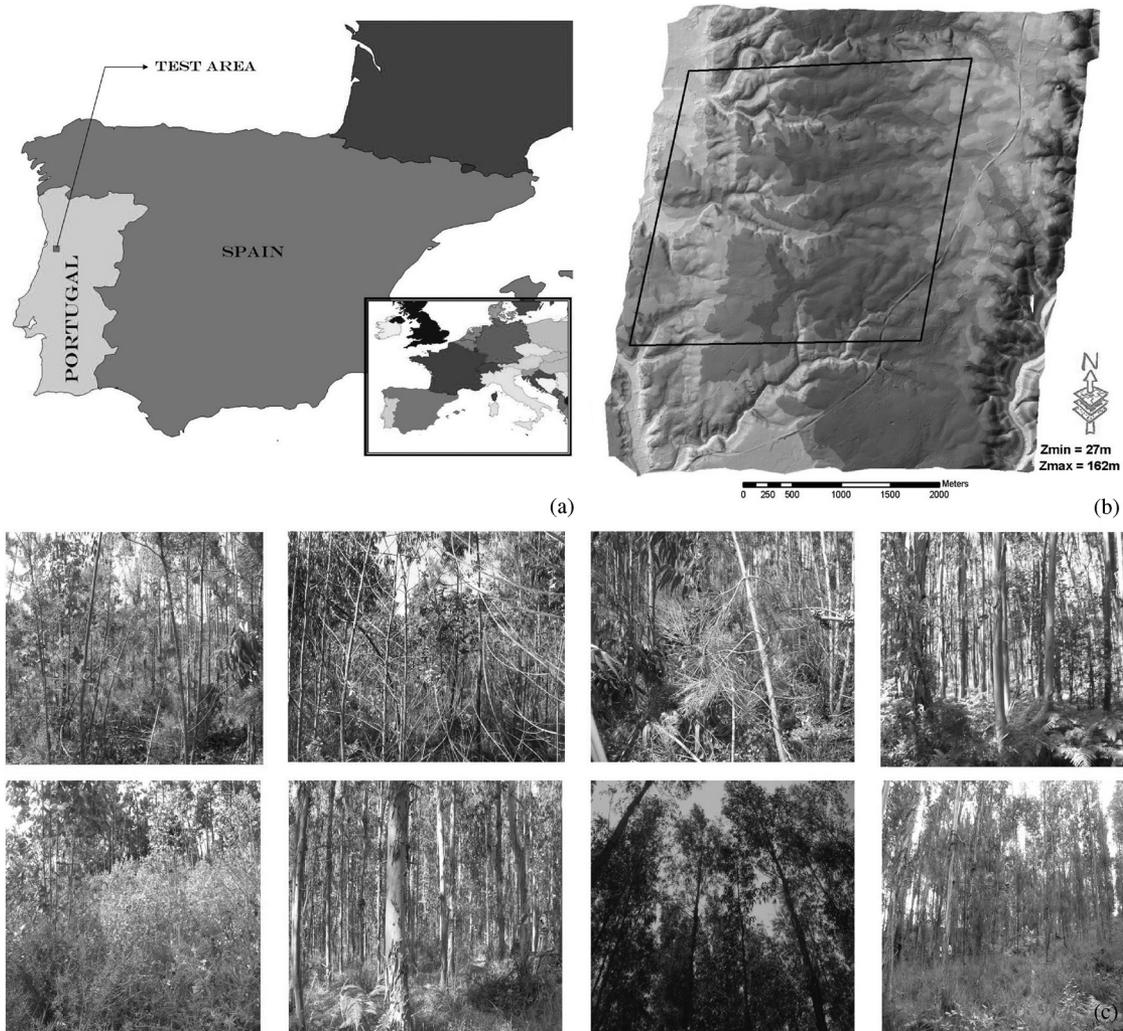


Fig. 1. (a) Localization of study area within Portugal and its delimitation. (b) DTM of the study area. (c) Examples of vegetation within study area.

140 software—the Riegl software, RiWorld for coordinate trans-  
 141 formation and RiAnalyze for processing of the FWALS data,  
 142 and the software bundle from TerraSolid. This comprises the  
 143 software modules TerraScan, TerraModeler, and TerraMatch.  
 144 TerraScan is the dedicated software solution for manipulation,  
 145 viewing, and classifying FWALS data. TerraModeler is a full-  
 146 featured terrain modeling application for creating digital terrain/  
 147 surface models (DTM/DSM). TerraMatch is used for automated  
 148 correction of ALS data ([http://www.igi-systems.com/products/  
 149 litemapper/components.htm](http://www.igi-systems.com/products/litemapper/components.htm), 19.01.2010).

150 Stereopairs of CIR images were acquired simultaneously  
 151 with the FWALS data. The parameters selected for FWALS  
 152 and image data acquisition are listed in Table II. The flight is  
 153 constituted by thirty plus two flight lines, the latter in a direction  
 154 perpendicular to the first and used for calibration purposes.

### 155 III. METHODS

#### 156 A. Procedures for Acquisition and Quality Assessment of 157 Reference Data

158 Reference data are needed to be verified, in terms of precision  
 159 and reliability, the DTM produced by means of the laser data  
 160 and a filtering algorithm (Section III-B-2). The strategy for

the reference data collection was not straightforward. In forest  
 161 areas, the collection of these data is time consuming, mainly in  
 162 plots with a high density of shrubs and trees. Furthermore, be-  
 163 cause the data were georeferenced, geodetic Global Navigation  
 164 Satellite Systems (GNSS) receivers had to be used. 165

The planning of the topographic survey was based on that of  
 166 the forest inventory, which started by selecting 43 plots, with  
 167 a radius of 11.28 m, within the study area [dots in Fig. 2(a)].  
 168 The DTM was represented by the coordinates of terrain points  
 169 located aside trees, which give also the locations of the trees,  
 170 and by the coordinates of prominent terrain points, like those on  
 171 breaklines [Fig. 2(b)]. The characterization of each plot in the  
 172 study area, in terms of mean slope (Section III-C), shrubbery  
 173 (height varying between 10 cm and 2 m), and high vegetation  
 174 (height > 2 m) is shown in Fig. 3. 175

In addition to the DTM quality assessment, it was also de-  
 176 cided to assess the quality of the delivered FWALS data. To this  
 177 end, it was measured points, in a grid format of approximately  
 178  $1 \times 1 \text{ m}^2$ , on bare surfaces [roads and fields; Fig. 4(a)–(c)]. 179

This information was collected by means of a topographic  
 180 survey using the radial method [16]. The coordinate system in  
 181 which the FWALS and image data were collected (WGS84-  
 182 UTM29) uses the reference ellipsoid of the World Geodetic  
 183

TABLE II  
FWALS DATA AND IMAGE ACQUISITION PARAMETERS

FWALS	Image
FWALS sensor: Riegl LMS-Q560	Camera sensor: Digicam H39 + 50 mm focal length
Wavelength: 1550 nm	Wavelengths in CIR mode (nm): B=500– 620; G=580– 800; R=800– 1000
Scan angle: 45°	
Pulse rate: 150 kHz	Image repetition rate: 1.9 sec
Effective Measurement rate: 75 kHz	
Beam divergence: 0.5 mrad	
Ground speed: 46.26 m/s	
Flying height above terrain: 600 m	Flying height above terrain: 600 m
Swath: 497 m	Overlap: 60%
Sidelap: 70%	Sidelap: 30%
Single run density of laser shots: 3.3 pts/m <sup>2</sup>	Nr. of pixels forward: 7 216
Expected final density: 9.9 pts/m <sup>2</sup>	Nr. of pixels sideward: 5 412
Distance between lines: 150 m	
Spot diameter: 30 cm	Ground sampling distance (GSD): 8.2 cm

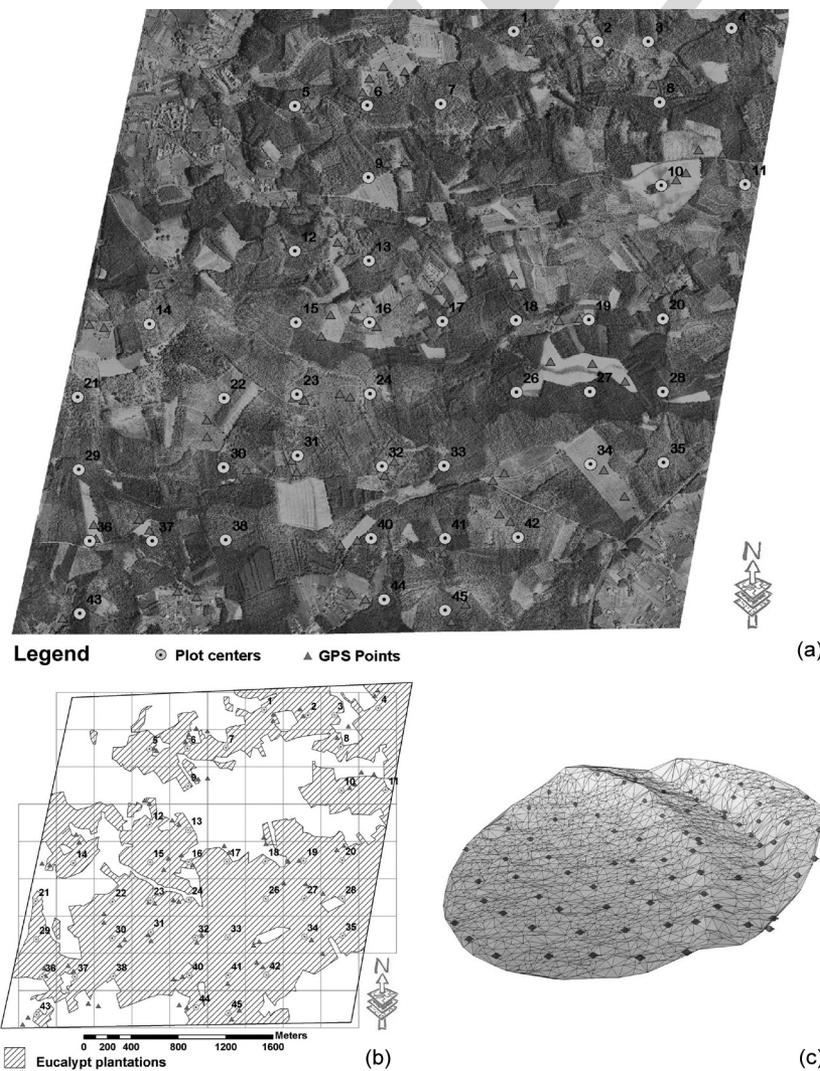


Fig. 2. Planning of the topographical survey: (a) and (b) location of the 43 forest plots; (c) DTM of plot no. 43 with breaklines, produced with the FWALS points classified as terrain. The small circles represent the points measured by topographic means.

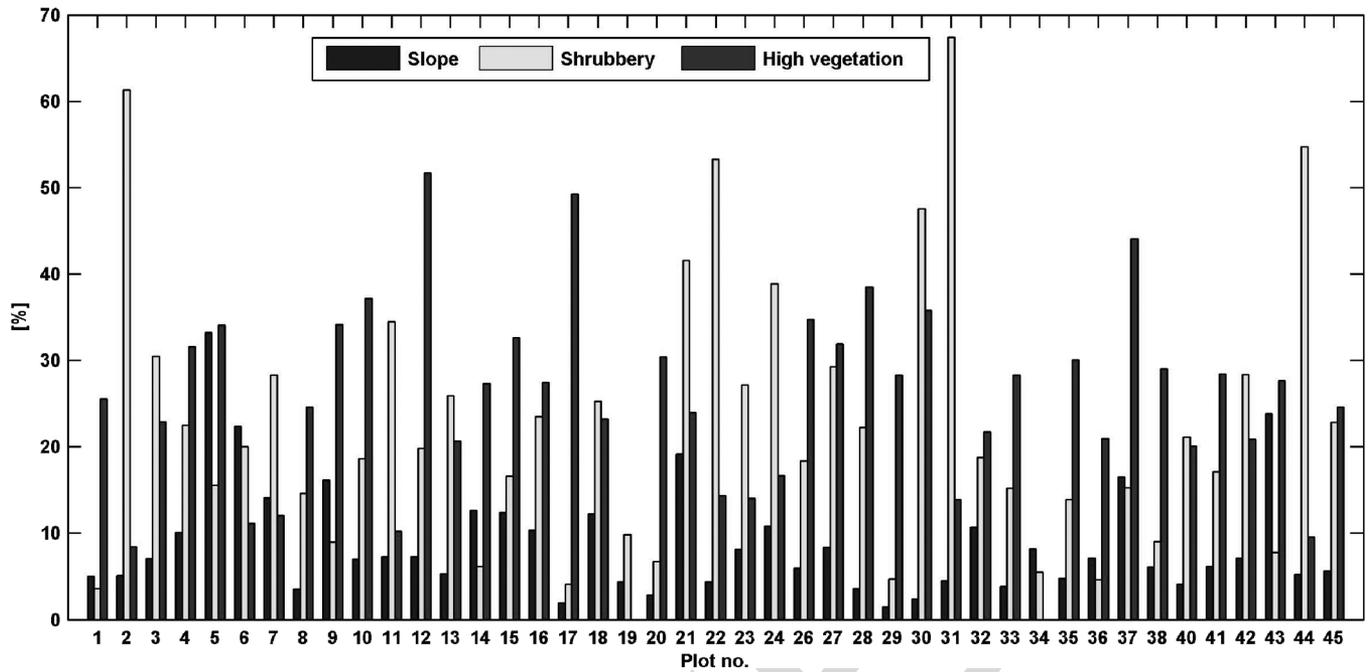


Fig. 3. Characterization of each plot in the study area.

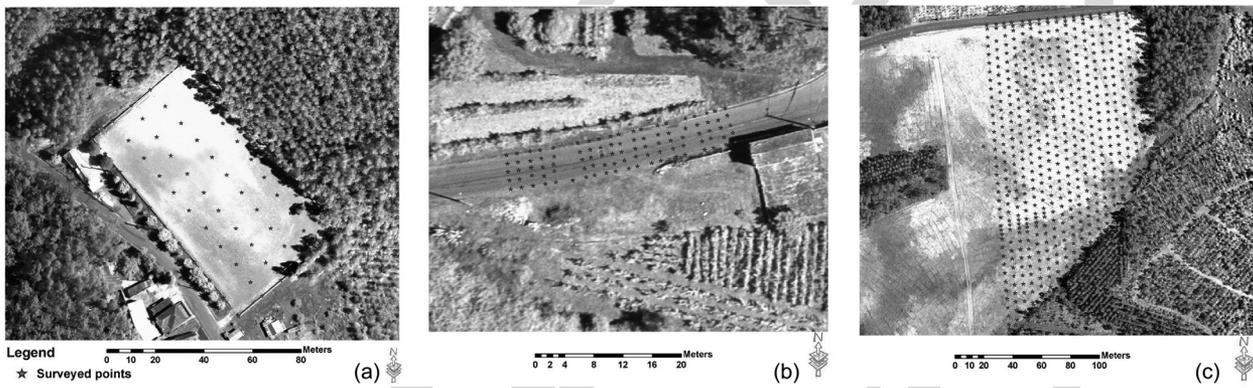


Fig. 4. Reference data for FWALS quality assessment. (a), (b), and (c) surveyed points on bare surfaces (paved road, football field and arable field, respectively).

184 System (WGS 84) and the cartographic projection Universal  
 185 Transverse Mercator (UTM) zone 29, for X and Y coordinates.  
 186 For the Z coordinate, the WGS84 ellipsoidal height was used.  
 187 These coordinates will be from now on referred to as absolute  
 188 coordinates X, Y, and h. Because this is not a local system,  
 189 the geographic information collected in the field had to be  
 190 converted to that system by using the GPS. To this end, it was  
 191 decided to attach to each plot two GPS-derived points, named  
 192 *GPS base* (triangles in Fig. 2(a)), whose coordinates were  
 193 measured with two GNSS receivers. They allow coordinating  
 194 the surveyed points directly in the referred coordinate system.  
 195 Two points are needed to orient the total station. These two  
 196 points are placed as close as possible to the plot and as much as  
 197 possible in an opened space (Fig. 5). This criterion turned out  
 198 to be difficult to fulfil in the study area.  
 199 The method used to measure the coordinates of the two GPS-  
 200 derived points was the relative positioning by using a fixed receiver  
 201 on a geodetic pillar with known coordinates on the WGS84-  
 202 UTM29. This method, in postprocessing, is the most precise  
 203 and may reach levels of precision in the order of the cm [17].

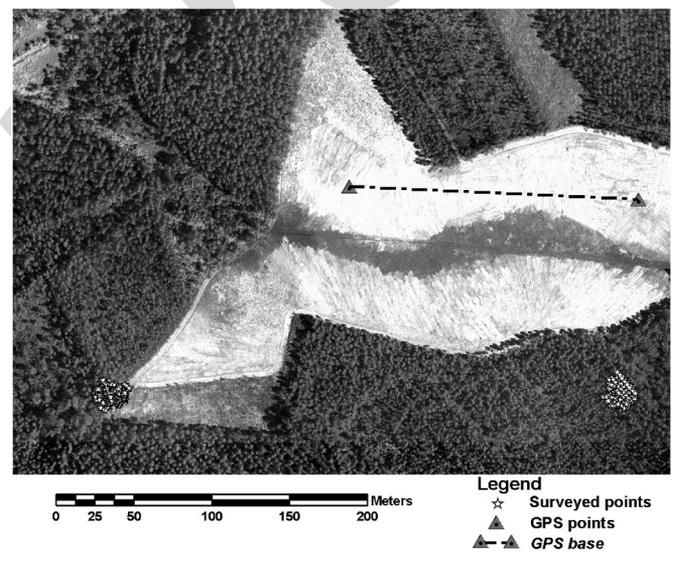


Fig. 5. Example of a *GPS base* used to georeference the measurements on two plots.

204 The GPS observations were collected according to a previous  
 205 planning dictated by the following specifications: minimum  
 206 time of observation of 60 min, depending on the point surround-  
 207 ings (normally it took 120 min), minimum number of seven  
 208 satellites, and value of the position dilution of precision less  
 209 than 3. In total, the coordinates of 82 points, i.e., 41 *GPS bases*  
 210 were measured. The number of *GPS bases* does not coincide  
 211 with that of plots because two *GPS bases* were assigned to two  
 212 plots each.

213 As some of the GPS-derived points of the *GPS bases* had  
 214 nearby some trees that might have degraded the GPS signal, we  
 215 developed a strategy to control the quality of their coordinates.  
 216 This strategy is based on the distances and height differences  
 217 between the two points of the *GPS bases* computed in two  
 218 coordinate systems. One is a local system, and the other the  
 219 absolute system (*WGS84 – UTM29*). Those variables are  
 220 compared to each other. It is assumed that the variables obtained  
 221 by topographic methods, using a total station, are error free. It  
 222 was decided that if the differences in the distance values and in  
 223 the height differences were bigger than 6 cm, the coordinates  
 224 of the GPS-derived points had to be remeasured by stationing  
 225 once again the GPS on the points. The mean error and RMSE of  
 226 the differences in distance are 1.7 cm and 2.5 cm, respectively,  
 227 while the mean error and RMSE of the differences in height  
 228 differences have the values of 1.9 mm and 2.6 cm, respectively.

## 229 B. FWALS Data Processing

230 1) *Full-Waveform Data Preparation*: The full-waveform  
 231 laser data were purchased already processed by the company  
 232 that acquired the data. The processing was done with the  
 233 RiAnalyze software from Riegl. The basic processing is based  
 234 on a Gauss pulse estimation technique. This aims to combine  
 235 the execution time of a simple center of gravity approach  
 236 ([18]) with the accuracy of Gaussian pulse fitting technique and  
 237 provides multiple returns ([19]). A maximum of five returns  
 238 were obtained for all the FWALS echoes.

239 The strips of FWALS data were also adjusted and geo-  
 240 referenced by the referred company using the TerraMatch  
 241 and the RiWorld softwares and were stored in the form of  
 242 LAS blocks. TerraMatch compares overlapping laser strips  
 243 with each other and corrects their orientation to obtain  
 244 the best fit and improved accuracy of the laser data. The  
 245 adjustment is based on measured differences between the  
 246 xyz shape or the intensity of laser points from differ-  
 247 ent strips (<http://www.terrasolid.fi/system/files/TerraMatch.pdf>,  
 248 23.12.2010). With the TerraMatch software, the misalignment  
 249 between the navigation and the laser systems is estimated with  
 250 calibration strips. The misalignment is then corrected when  
 251 georeferencing the cloud of points with the RiWorld software.  
 252 The TerraMatch software is once again used to adjust the height  
 253 values between points in overlapping strips by smoothing GPS  
 254 fluctuations. Finally, the data are verified for systematic errors  
 255 by comparing the h values of ground control points with those  
 256 of laser points.

257 2) *Point Cloud Filtering*: The point cloud obtained by pro-  
 258 cessing the full-waveform data contains both laser returns from  
 259 objects located on the ground (buildings, trees, understory

vegetation) as the ground itself. Moreover, in vegetated areas  
 the laser pulse can penetrate the canopy, travelling between the  
 branches and leaves and be reflected by the understory (e.g.,  
 subcanopy trees and shrubs) and by the ground.

Since we are interested in the reconstruction of the terrain  
 surface, it is necessary to remove the nonground returns from  
 the point cloud. Removing nonground measurements from the  
 point cloud is here referred to as filtering. In order to find the  
 most probable ground returns from the initial point cloud, the  
 filtering process uses the last return of each pulse.

Points in a point cloud are classified as either ground or non-  
 ground returns using a progressive triangular irregular network  
 (TIN) densification algorithm implemented in the TerraScan  
 software v4.006 ([www.terrasolid.fi/en](http://www.terrasolid.fi/en), 11.05.2010) and based  
 on the algorithm published in [20], [21]. This algorithm begins  
 by constructing a sparse TIN with the lowest point inside a  
 square neighborhood of a certain size. The TIN is densified by  
 adding points of the set of unclassified points that satisfy certain  
 criteria. The densifying of the TIN stops when no more unclas-  
 sified points satisfy the criteria or if a certain ground point den-  
 sity is achieved. An additional threshold may be used to control  
 the maximum terrain slope allowed at each triangle of the TIN  
 surface. In this paper, the parameters used for filtering the laser  
 point cloud were tuned empirically and happen to coincide with  
 the default values ([22]) with exception of the neighborhood  
 size parameter that has the value of 10 m (the default is 60 m  
 and therefore too big when compared to each plot size).

## C. DTM Accuracy Assessment

In addition to the accuracy assessment of the DTM produced  
 by the filtering process of the laser data, the contribution of  
 the quality of the FWALS data alone into that assessment was  
 verified. To this end, three DTM were produced, by using topo-  
 graphic means, on flat bare surfaces (a paved road, a football  
 field, and an arable field; see Section III-A). The accuracy  
 assessment of these DTM as well as the DTM produced after  
 the filtering process of the laser data relates to the estimation  
 of the mean and RMSE of the differences (dh) between the  
 h values of FWALS terrain points and those of the reference  
 points (Section III-A) at the same planimetric locations. These  
 FWALS terrain points are interpolated from a TIN computed  
 with the filtered FWALS terrain points (Section III-B-2). The  
 TIN was created from the filtered FWALS data instead of from  
 the reference data because the density of FWALS terrain points  
 is higher than that of reference points (3.8 versus 0.2 pts/m<sup>2</sup>).  
 This means that many points on the terrain presented in the  
 FWALS filtered data are not present in the reference data.  
 Therefore, it is more appropriate to interpolate the laser data  
 to the planimetric positions of the reference data. The TIN  
 format is also chosen, instead of the grid format, to handle the  
 breaklines. Furthermore, the interpolation methods associated to  
 this format are computationally simpler when compared, for  
 example, with the kriging method, at expenses of being more  
 sensitive to measurements errors.

The deviation of the distribution of the residuals dh from the  
 normal distribution implies the use of other accuracy estimators  
 than the mean and RMSE ([23], [24]). Therefore, that deviation

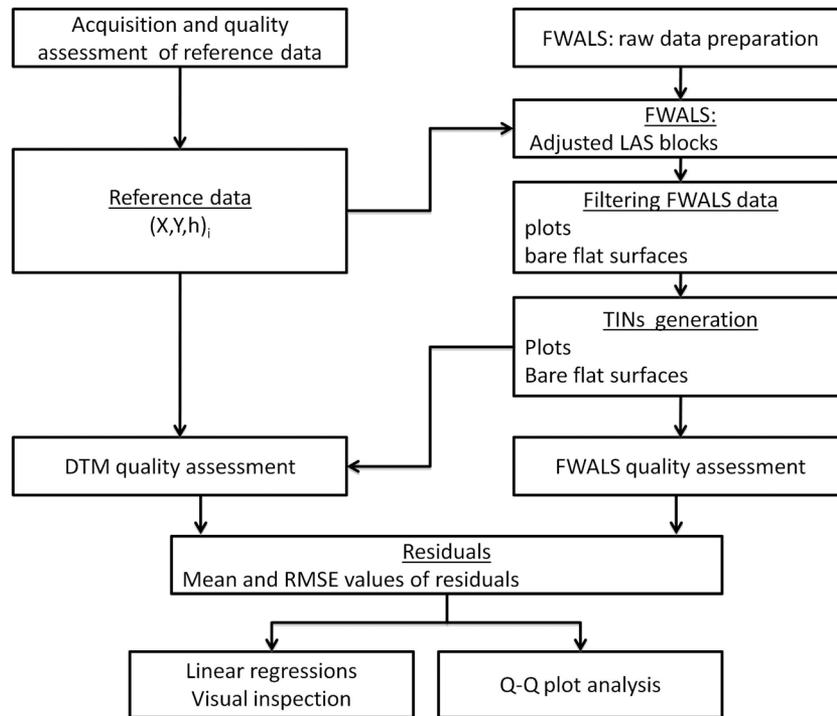


Fig. 6. Concise presentation of the developed methodology.

316 has to be evaluated. This is done using two different methods. 317 One method superimposes on the histogram of the residuals a 318 curve for a normal distribution obtained by ordinary estimation 319 of mean error and variance. If the Gaussian curve does not fit 320 the data very well, it means that the residuals are not originating 321 from a normal distribution, e.g., because the distribution is 322 not symmetric around its mean or because the distribution is 323 more peaked around its mean than the normal distribution while 324 having heavier tails [24]. These effects are measured by the 325 skewness and kurtosis of the distribution. The analysis of the 326 so-called quantile-quantile (Q-Q) plot is another method for 327 verifying a deviation from the normal distribution. The quan- 328 tiles of the empirical distribution function are plotted against 329 the theoretical quantiles of the normal distribution. If the distri- 330 bution is normal, the Q-Q plot should yield a straight line. As 331 a robust and distribution free approach handling outliers and 332 non-normal distribution, [24] suggest as accuracy measures the 333 median  $m_{dh}$  of residuals, the normalized median absolute de- 334 viation ( $NMAD = 1.4826 \cdot median_j(|dh_j - m_{dh}|)$ ), and the 335 68.3% and 95% quantiles of the absolute value of residuals.

336 It was also verified if the quality of the DTM has a relation 337 with the vertical structure of the vegetation and with the ter- 338 rain slope. Therefore, a linear regression analysis between the 339 quality of the DTM produced by filtering the laser data and 340 the vertical structure of the vegetation was computed. Linear 341 regression was carried out between the values estimated for the 342 mean and the RMSE of residuals per plot and, respectively, the 343 percentages of the number of points classified by the filtering 344 process as terrain, as shrubbery with height varying from 10 cm 345 until 2 m and as higher vegetation cover.

346 The linear regression analysis between the quality of the 347 referred DTM and the mean slope of the terrain within plots 348 was also performed. Linear regression was carried out between

the values estimated for the RMSE of residuals per plot and 349 the mean slope within each plot. The mean slope was estimated 350 by fitting a plane to the reference terrain points. The plane is 351 in fact the geometric shape that represents best the geometry 352 of all the terrain surfaces within the plots. A relation between 353 the RMSE values and the spatial distribution of the plots as 354 well as the spatial point density was also studied by means of 355 visual inspection. The spatial point density was computed for 356  $1 \times 1 \text{ m}^2$  cell using only the last returns [25]. For those plots 357 on the same flight line, the associated RMSE values were also 358 inspected for a relation. 359

The developed methodology is summarized in the form of a 360 flow diagram in Fig. 6. 361

#### IV. RESULTS

362

The quality of the DTM obtained with the laser data was 363 computed for each plot individually, by using the strategy 364 described above. The estimated values for the mean, standard 365 deviation, and RMSE are plotted in Fig. 7. As it may be seen in 366 that figure, the values of the mean of residuals vary, in absolute 367 value, between 5 and 28 cm, while the RMSE varies between 4 368 and 29 cm. When considering all the plots together, i.e., 3174 369 points, the value of the mean of residuals is 8 cm, while the 370 RMSE is 15 cm. These are very good values considering the 371 cover characteristic within the plots. Furthermore, there has 372 been no manual editing, i.e., manual removal of gross errors. 373

The bias of 8 cm shows that the FWALS filtered points are 374 above the terrain surface. This may result from the vegetation 375 cover within the plots. Nonetheless, this value is quite low 376 considering the characteristics of the plots (Fig. 3), which 377 proves that the strategy for laser data acquisition and processing 378 is effective. 379

379

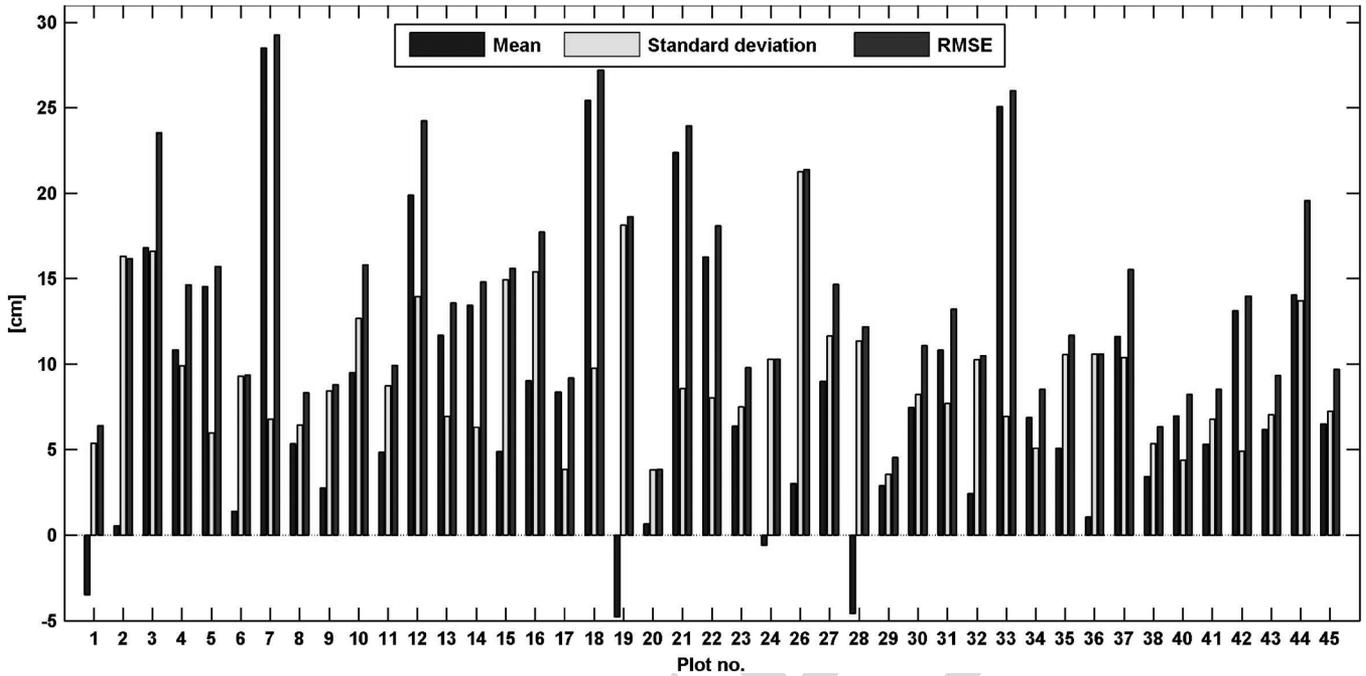


Fig. 7. Values of the mean, standard deviation, and RMSE of residuals per plot. The plots 19 and 34 are the only ones in open areas. They have young eucalypt plantations with mean heights smaller than 30 and 20 cm, respectively. Although situated in an open area, plot 19 has a bigger RMSE because of the terrain relief (it contains a breakline).

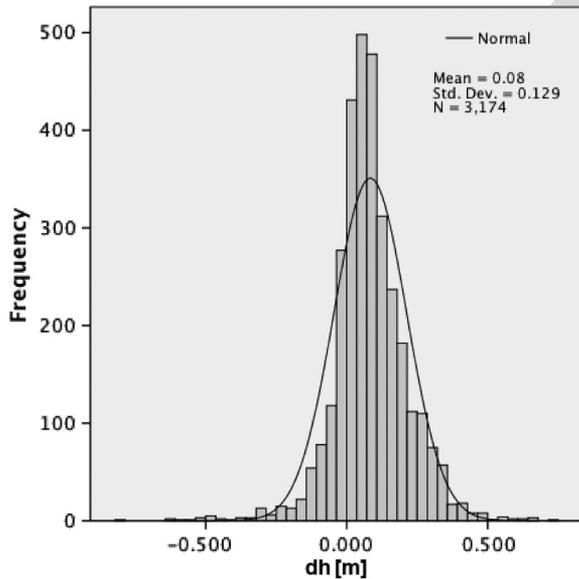


Fig. 8. Histogram of the residuals. Superimposed on the histogram are the expected counts from a normal distribution with mean and standard deviation estimated from the data.

380 The gross errors present in the results (min = -79 cm;  
 381 max = 72 cm) deviate from the distribution of the residuals  
 382 from a normal distribution. Fig. 8 shows the histogram of  
 383 the residuals on which is superimposed a curve for a normal  
 384 distribution obtained by ordinary estimation of mean error and  
 385 variance. As mentioned in Section III-C, the deviation from nor-  
 386 mality is measured by the skewness and kurtosis of the distribu-  
 387 tion which, in this case, have the values of  $-0.177$  (std. error =  
 388  $0.043$ ) and of  $3.979$  (std. error =  $0.087$ ), respectively.  
 389 The analysis of the so-called Q-Q plot provides another way  
 390 of verifying a deviation from the normal distribution. Fig. 9

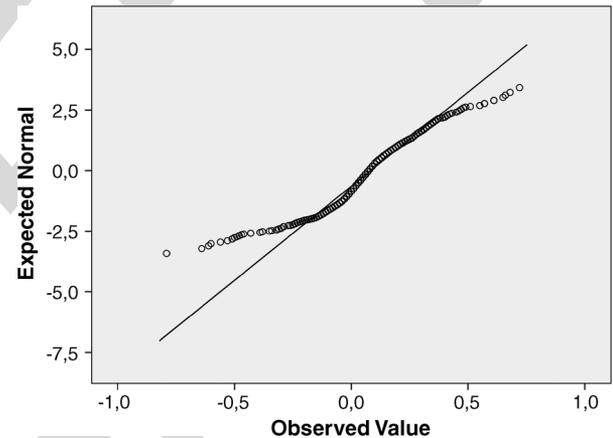


Fig. 9. Quantile-quantile plot of the distribution of residuals computed to assess the DTM accuracy.

TABLE III  
 ROBUST ACCURACY ESTIMATORS

Accuracy estimator	Value (cm)
$m_{dh}$	7
NMAD	10
68.3% Quantile	13
95% Quantile	32
RRMSE	12

shows the Q-Q plot for the distribution of the residuals. A  
 391 deviation from a straight line may be seen at both extremes,  
 392 which indicates that the distribution of the residuals is not  
 393 normal. The obtained robust accuracy measures (Section III-C)  
 394 are tabulated in Table III. Although the distribution of the  
 395 residuals deviate from the normal distribution, the differences  
 396

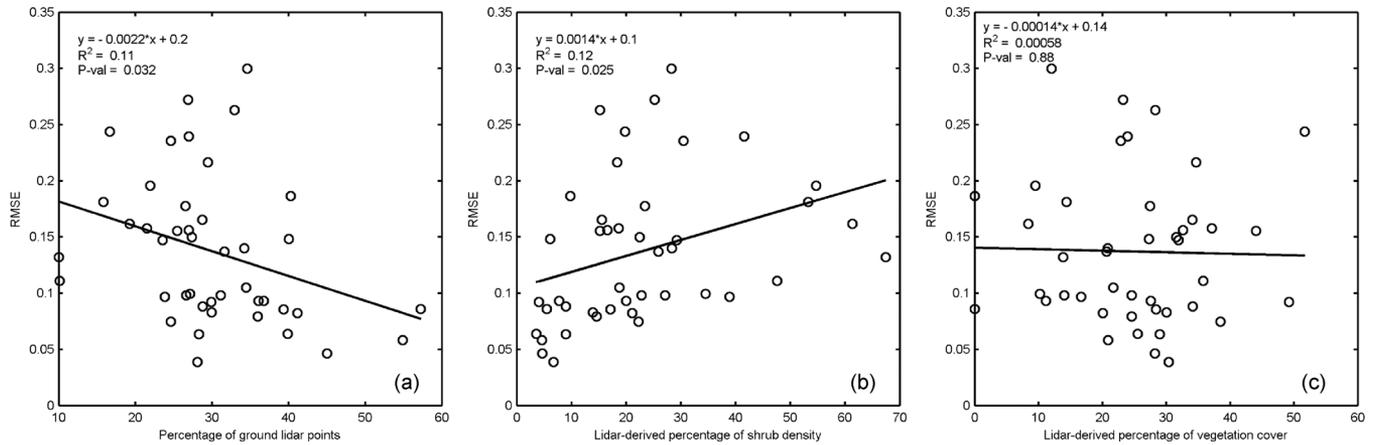


Fig. 10. Linear regression between the value of RMSE and the process-derived percentage of. (a) Terrain points. (b) Shrubberty. (c) Higher ( $\geq 2$  m) vegetation cover per plot.

397 between the robust accuracy measures and those estimated  
 398 by assuming a normal distribution of the residuals are not  
 399 significant. The difference between the values of the standard  
 400 deviation of residuals and of the  $NMAD$  is only about 3 cm.  
 401 If one takes the robust measures  $m_{dh}$  and  $NMAD$  as values  
 402 representative for the mean and standard deviation of residuals,  
 403 a robust estimator for the RMSE value (RRMSE), estimated as  
 404  $\sqrt{m_{dh} + NMAD^2}$ , is 12 cm (Table III).

405 The existence of a linear relation between the quality of the  
 406 DTM and the vertical structure of the vegetation and the mean  
 407 slope of the terrain in each plot was studied by means of linear  
 408 regression. It is expected that this relation exists. Fig. 10(a)–(c)  
 409 shows the linear regression between the values estimated for  
 410 the RMSE per plot and, respectively, the percentages of the  
 411 number of points classified by the filtering process as terrain,  
 412 as shrubberty with height varying from 10 cm until 2 m and as  
 413 high vegetation cover (i.e., with height  $> 2$  m).

414 Fig. 11 shows linear regression between the values estimated  
 415 for the RMSE per plot and the mean terrain slope within each  
 416 plot.

417 Contrary to what was expected, the low values obtained  
 418 for the coefficient of determination ( $R^2_{(a)} = 0.08$ ;  $R^2_{(b)} = 0.13$ ;  
 419  $R^2_{(c)} = 0.10$ ) indicate that there is no linear relation between  
 420 the quality of the DTM and the vertical structure of vegetation.  
 421 In fact, there is no relation judging by the point dispersion  
 422 characteristics shown in Fig. 10. The same tests and results  
 423 were obtained when using the mean of the residuals per plot  
 424 instead of the RMSE. In relation to the characteristics of plots  
 425 in terms of vegetation cover, this means that the algorithm  
 426 published in [21], together with the high density of the laser  
 427 points per  $m^2$  (in mean 10), is quite robust and insensitive to  
 428 the obstacles on the terrain. In relation to the terrain slope, the  
 429 results show that there is no relation between the slope and  
 430 the RMSE of residuals, which contradict those found in other  
 431 studies [12] but also support results from others [5].

432 The spatial distribution of the RMSE values per plot, the  
 433 spatial point density and the flight lines, within the study area  
 434 is shown in Fig. 12. As in previous studies [5], there seems  
 435 to be no relation between the RMSE values and the spatial  
 436 distribution of the plots. There appears to be also no relation  
 437 between the RMSE values and the spatial point density. This is

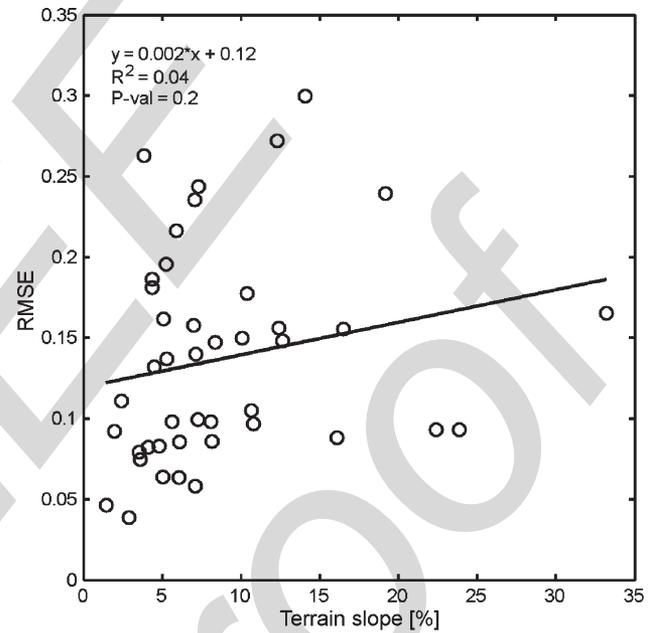


Fig. 11. Linear regression between the values of RMSE and the mean terrain slope per plot.

not surprising once the mean point density is high. In Fig. 12, it  
 438 may be also seen that the RMSE values for plots situated along  
 439 the same strip have no relation among them. Because the strips  
 440 have been adjusted, and thus systematic errors removed, it may  
 441 be stated that there is as well no relation between the RMSE  
 442 values and bias strip differences.

443  
 444 In order to verify the contribution of the quality of the  
 445 FWALS data alone into the assessment of the quality of the  
 446 DTM, the quality of the laser data was verified, as described  
 447 in Section III-C. The accuracy measures, i.e., the mean and  
 448 the RMSE of residuals are used instead of the robust esti-  
 449 mators aforementioned because the distribution of residuals is  
 450 considered normal (skewness and kurtosis equal to 0.067 and  
 451 0.268, respectively). See Fig. 13 representing the Q-Q plot). The  
 452 computed mean and RMSE are listed in Table IV. These values  
 453 are quite good and similar to those obtained in previous studies  
 454 ([26]). The worst is obtained for the arable land and results from  
 455 the fact that the pole used for measuring sunk into the fresh

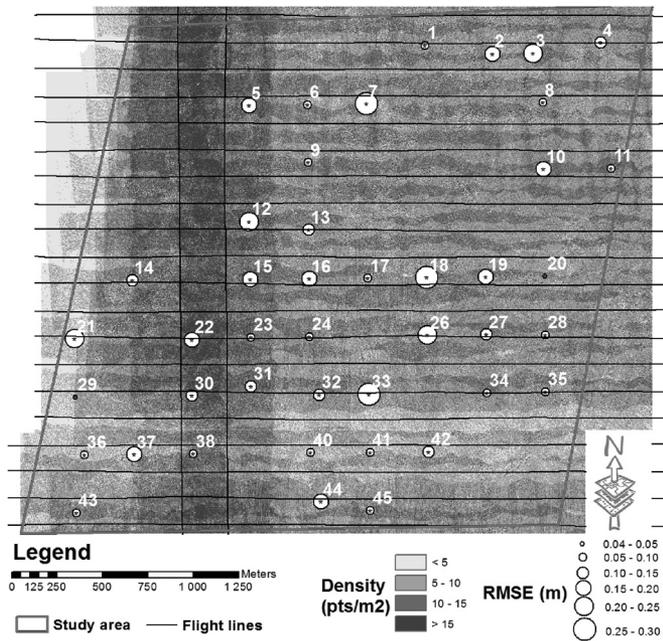


Fig. 12. Spatial distribution of the RMSE value per plot.

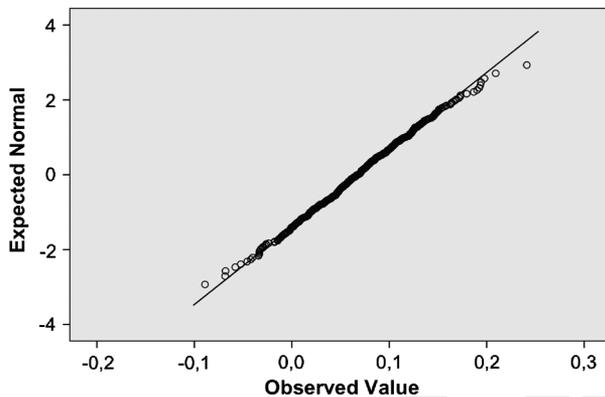


Fig. 13. Quantile-quantile plot of the distribution of residuals computed to assess the accuracy of the delivered laser data.

TABLE IV  
MEAN AND RMSE VALUES OF RESIDUALS, PER SURFACE

	Mean (cm)	RMSE (cm)
Paved road	2.8	3.2
Football field	1.8	3.4
Arable field	6.8	8.4

456 soil for a few centimeters. This pole is used to transport the  
457 prism needed to measure the coordinates of the DTM points by  
458 topographic means. There is a bias that shows that the FWALS  
459 filtered points are above the terrain surface. Nonetheless, this  
460 bias value is insignificant for the majority of applications.

461 In addition to this evaluation of the accuracy of the FWALS  
462 data, the mean and standard deviation of the residuals between  
463 the  $h$  values of the *GPS bases* and those derived from the laser  
464 data at the same location were computed. A mean of 0.3 cm and  
465 a RMSE of 6.8 cm were obtained. This mean value of residuals

is quite different from those computed in flat surfaces because  
466 while the *GPS bases* are spread all over the study area, and thus  
467 involving all the 30 laser strips, the other use only seven laser  
468 strips, five of which are common. 469

## V. DISCUSSION AND CONCLUSION

470

A study as the one here presented, carried out in an area of  
471 900 ha with unmanaged eucalypt forest with great amount of  
472 high shrubbery, chaotic tree distribution, and high tree density,  
473 is quite demanding. It requires an exhaustive and well-planned  
474 collection of reliable control data to properly assess the quality  
475 of FWALS data and of a derived DTM. The georeferencing  
476 of these data makes use of GNSS techniques, which are very  
477 time consuming in environments where a high density of high  
478 vegetation predominates. This is because under high vegetation,  
479 the GNSS receiver has to be stationed on the GPS-derived  
480 points for long periods of time (the georeferencing of the  
481 GPS-derived points took approximately 170 h). Furthermore,  
482 topographic measurements in plots covered with high shrubs  
483 are also difficult to execute. 484

The accuracy of the produced DTM is very good and suitable  
485 for a great number of applications. The value of 12 cm in RMSE  
486 is smaller than that published in several studies carried out in  
487 less chaotic forests (Table I; [1]). It also compares to the RMSE  
488 value claimed 12 years ago in [2], but for flat surfaces, and  
489 still referred in several studies. For similar surfaces, a value  
490 of 3 cm in RMSE was instead obtained. Although such a high  
491 DTM accuracy is not needed per se for this type of applications,  
492 such as forest inventory, acquiring very dense point clouds is  
493 recommended to describe the crown characteristics accurately,  
494 including total tree height ([1]). 495

Considering that the outgoing pulse width is 4 ns, the  
496 expected echo separation is 60 cm. Nonetheless, the full-  
497 waveform data processing algorithm detected 93 pairs of con-  
498 secutive echoes with a separation smaller than 60 cm, among  
499 the 40 523 recorded pairs of consecutive echoes. It may be said  
500 that there is a slight improvement in echo separation, due to the  
501 postprocessing wave analysis, but it remains to be verified if all  
502 of the 93 pairs of echoes really relate to reflections. 503

The filtering algorithm based on the filter of [21] also proved  
504 to be very robust to terrain with the aforesaid characteristics.  
505 The low values obtained for the coefficients of determination  
506  $R^2$  between the RMSE values per plot and the percentages  
507 of the number of points classified by the filtering process as  
508 terrain, as shrubbery with height varying from 10 cm until 2  
509 m and as higher vegetation cover, indicate that there is not  
510 a relation between the quality of the DTM and the vertical  
511 structure of vegetation. This means that the Axelsson algorithm,  
512 together with a high density of the laser points per  $m^2$  (in  
513 mean 10), is quite robust and insensitive to the obstacles on  
514 the terrain. The spatial distribution of the RMSE values per plot  
515 shows also no relation between the RMSE values and the spatial  
516 distribution of the plots as well as the spatial point density.  
517 The same conclusion was attained for the relation between the  
518 RMSE values and the mean terrain slope. Because this is not  
519 what was expected, further studies are needed. 520

521 To conclude, the FWALS system proved to be adequate, in  
522 terms of data accuracy, for the production of DTM in areas of  
523 unmanaged high-density eucalypt forest. The accuracy figure  
524 of 12 cm for the DTM produced in such areas, using the data  
525 acquisition parameters listed in Table II and the Axelsson fil-  
526 tering algorithm [21], is also acceptable to derive forest canopy  
527 structure metrics.

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