



Mapping the annual evolution of snow depth in a small catchment in the Pyrenees using the long-range terrestrial laser scanning

Journal:	<i>Journal of Maps</i>
Manuscript ID:	TJOM-2013-0041.R2
Manuscript Type:	Original Article
Date Submitted by the Author:	n/a
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Keywords:	snow depth distribution maps, terrestrial laser scanner, Pyrenees, application protocol

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Mapping the annual evolution of snow depth in a small catchment in the Pyrenees using the long-range terrestrial laser scanning

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Abstract: This paper presents the methodology used to develop snow depth distribution maps for a small catchment in the Central Spanish Pyrenees covering 55 ha in a 1:10000 scale. The maps were obtained using LiDAR (laser imaging detection and ranging) technology, using a long-range Terrestrial Laser Scanner (TLS) in six field surveys undertaken during the 2012 winter–spring period. This technique enabled the acquisition of information at a very high resolution concerning the spatial variability of snow cover, providing snow depth information for remote areas where data acquisition is complex and hazardous. We describe the methodological steps and the quality assessment applied in developing the maps. Comparison with manual measurements confirmed the reliability of the snow depth maps, including areas located at large distances from the scanner (800 m). This method provides a promising tool for future investigations of snow dynamics in mountainous environments.

Key words: snow depth distribution maps, terrestrial laser scanner, Pyrenees, application protocol

1. Introduction

In mountainous areas the snowpack has very high spatial variability as a consequence of vertical lapse rates in temperature (which affect the phases of precipitation), the distribution of precipitation amounts, terrain shadows caused by complex topography, ground irregularities, and snow redistribution by wind and avalanches (Deems et al., 2006; Grünewald et al., 2010; Liston and Sturm, 2002; López-Moreno et al., 2011). Accurate maps of snow duration and thickness provide valuable information for understanding many environmental processes that occur in mountainous areas, including the distribution and survival rates of plants (Keller et al., 2000; Körner, 1994), soil temperature and moisture (Groffman et al., 2001; Gross et al., 1990), identification of avalanche start and runout zones (Birkeland et al., 1995; Sovilla, 2004), erosion rates (Colbeck et al., 1979; Pomeroy and Gray, 1995), and runoff dynamics and snow cover depletion in mountainous catchments (Lehning et al., 2006; Pomeroy et al., 2004), among many other factors.

The importance of snow has focused scientific interest during the last two decades on the mapping of snow depth (SD) and the snow water equivalent (SWE) at various spatio-temporal scales (Bales and Harrington, 1995; Erickson et al., 2005). A widely used mapping procedure is focused on estimating SD and the SWE as a function of terrain characteristics. In this approach various topographical and geographical features, generally derived from a digital elevation model (DEM), are used as predictor variables on the basis that they are closely related to the accumulation, redistribution, and ablation of the snow cover (Bavera and De Michele, 2009; Erxleben et al., 2002; Jost et al., 2007; Molotch et al., 2005). When the relationships between snow accumulation and predictor variables have been established, the spatial distribution of SD can be determined for areas with well-known terrain characteristics. This technique is one of the so-called regression-based methods. It requires the acquisition of many SD or SWE measurements in complex and often dangerous terrain, and is subject to several uncertainties regarding the effects of grid and sample size (digital terrain model resolution and number of data points respectively) (López-Moreno et al., 2010).

As manual high resolution SD data acquisition over large areas is unfeasible, remote sensing techniques have been recently applied for this purpose. Examples include LiDAR (laser imaging

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3 detection and ranging) techniques, which commonly utilize airborne or terrestrial laser scanners
4 (respectively ALS and TLS). In recent years, these have provided accurate results in SD studies
5 compared with other techniques (Deems et al., 2006; Hopkinson et al., 2004; Jörg et al., 2006;
6 Prokop et al., 2008; Schaffhauser et al., 2008). For medium-sized and small areas the TLS
7 technique (Lundberg et al., 2008) represents a more flexible and inexpensive option as compared
8 to ALS, as TLSs are portable and snow can be surveyed without using an aircraft (note the costs of
9 hiring an aircraft). In addition, SD can be sampled at the most suitable spatial resolution for the
10 user's needs. Although the application of TLSs is not currently widespread in Alpine areas, an
11 increasing number of studies have highlighted its accuracy and usefulness for snow studies in
12 which high resolution data are required (Egli et al., 2012; Grünewald et al., 2010; Prokop, 2008).
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14 In this study a long range TLS was used to obtain high spatial resolution snow cover and depth
15 distribution maps in a small mountain catchment (55 ha) in the Central Spanish Pyrenees. We
16 describe the protocol followed for generating snow distribution maps during the snow accumulation
17 and melting periods in 2012. To assess the robustness of this protocol, the data obtained at various
18 distances from the TLS were compared with direct manual SD measurements.
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35 **2. Study Area**

36 The Izas experimental catchment has an area of approximately 55 ha and is located in the Central
37 Spanish Pyrenees (42°44' N, 0°25' W) between 2000 and 2300 m above sea level (asl). This area
38 is close to the main divide of the Pyrenees in the headwaters of the Gállego River, near the border
39 with France (Figure 1). The catchment is exposed to Atlantic climate conditions and, consequently,
40 the winters are relatively humid, and the snowpack covers most of the catchment from November to
41 the end of May (López-Moreno et al., 2010). The winter and spring of 2012 were particularly dry
42 and mild, and snowpack remained below the average throughout the winter and into the spring.
43 However, late snowfall events occurred by the end of April increased the amount of snow at the end
44 of the melting period. The vegetation cover is composed of high mountain pastures, although in
45 some areas the ground is bare of vegetation, with rocky outcrops occurring in the steeper areas
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3 (less than 15% of the study area). The mean slope of the catchment is 16° (López-Moreno et al.,
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6 2012).
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9 **3. Terrestrial Laser Scanner**

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11 LiDAR are those remote sensing technologies that emit laser light, and measures the distance
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13 between the target and the device based on the phase shifting of light, time of flight measurement
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15 or interferometry. LiDAR technology has undergone rapid development in recent years and both
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17 terrestrial and airborne LiDAR have been used in several types of geomorphological studies,
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19 including those investigating rock falls (Abellán et al., 2006, 2010), landslides (Bitelli et al., 2004;
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21 Jaboyedoff et al., 2012; Oppikofer et al., 2008), fluvial dynamics (Heritage and Hetherington, 2007)
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23 and soil degradation (Romanescu et al., 2011).
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25 The device used in the present study is a long-range TLS (RIEGL LPM-321), which uses time of
26
27 flight technology. The device measures the time between a light pulse emission and its detection.
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29 The resulting value is multiplied by the speed of light through air, and divided by two (the measured
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31 time is twice the time required for the light pulse to reach the target from the TLS; Figure 2). Each
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33 distance measurement is repeated for every single point acquired in the area of interest, obtaining a
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35 three-dimensional point cloud from real topography.
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37 The technical characteristics of the TLS used are: (a) light pulses of 905 nm wavelength (near-
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39 infrared), which is ideal for acquiring data from snow cover (Prokop, 2008); (b) the minimum angular
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41 step width of $0,018^\circ$; (c) the laser beam divergence of 0.046° ; and (d) the maximum work distance
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43 of 6000 m. The sampling speed depends on the operation mode, which in turn depends on the
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45 target distance, but ranges from 10 to 1000 Hz. This enables large areas to be surveyed in a
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47 reasonable time. This TLS has been shown to be an appropriate tool for deriving high resolution
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49 digital terrain models (DTM) applicable to cryosphere studies (Avian and Bauer, 2006; Egli et al.,
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51 2012; Grünwald et al., 2010; Schwalbe et al., 2008).
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53 Despite the number of recent studies in which TLSs have been used, the technique is still
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55 considered rather novel, with a limited number of applications to snow studies. When used for
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57 scanning at large distances, various error sources need to be taken into account and minimized
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3 (Reshetyuk, 2006). Thus, it is important to establish a protocol for obtaining data and generating
4 accurate DTMs that will facilitate development of accurate SD distribution maps. This paper
5 describes the operational procedure used for generating those maps for Izas experimental
6 catchment in Central Spanish Pyrenees.
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10 11 12 13 **4. Data collection**

14 In this section the methodological steps applied for collecting data are fully explained. First, a short
15 introduction of the experimental setup is presented, followed by the scanning procedure applied for
16 each survey.
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20 21 **Experimental setup design**

22 The first step, which precedes data collection with the TLS, is to consider several factors for
23 choosing the optimal position (or positions) to place the TLS (scan station). LiDAR is a line-of-sight
24 technique (i.e., any object placed between the TLS and the study area will prevent the acquisition of
25 information behind it) where the cost of multiple scan stations should be weighed against the benefit
26 (i.e., occlusion reduction). As the Izas Experimental Catchment presents a relatively smooth
27 topography with a single scan station 62% of the area is covered (scan station 1 in Figure 3), but
28 the south zone of the study area is hidden from this point of view. Consequently, a second scan
29 station, placed by around 450 m away from scan station 1, was established in the central area of
30 the catchment (see scan station 2 in Figure 3). This enabled the reduction of topographic shadows
31 in the final points cloud as is shown in Figure 4 and therefore 86% of the study area was scan
32 covered in final maps. Furthermore, carrying the TLS equipment and accessories (e.g., tripod,
33 batteries, laptop and cables), with a weight of around 45 kg, in alpine terrain with snow presence,
34 must be considered. Thus, experimental area and scan stations should be chosen properly to
35 accommodate logistical considerations.
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50 Indirect registration, also called target-based registration, was used to merge the
51 information from the different scan stations, and for transforming the local coordinates of the
52 obtained point clouds into a global coordinate system. It enabled the comparison of scans made on
53 different survey days. Indirect registration uses fixed reference points (targets) that are located in
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3 the study area. Those targets must be visible to all scan stations because they are used for
4 transforming coordinates, as described in the next section. Reflective targets of known shape and
5 dimensions are placed at those reference points. Although a minimum of three targets should be
6 used, we installed 12 within the experimental basin, each comprising a cylindrical reflector 0.25 m
7 high \times 0.25 m diameter mounted on a pole 2 m tall (Figure 3). The distance of the targets to the two
8 scan stations ranged from 110 to 730 m. This number of targets increased the robustness of the
9 experimental setup and prevented any target loss. Using standard topographic methods we
10 obtained accurate global coordinates for the targets using a differential GPS with post-processing
11 and a total station. Two GPS base-station receivers with a long measuring time (up to 2 h) should
12 be established in the study area (note that all targets should be visible from at least one of them to
13 undertake afterwards an optical survey), and their positions corrected using data from a permanent
14 GPS station. When both bases were correctly established and marked, a total station was
15 positioned at the base where all targets were visible, and orientated towards the second base. The
16 global coordinates of the targets were acquired using this total station, undertaking an optical
17 survey. In our study the global coordinates were acquired in the UTM 30 coordinate system in the
18 ETRS89 datum. The final precision for the global target coordinate was 0.05 m by planimetry and
19 0.1m by altimetry.

Scanning procedure

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21 In order to reduce one of the most important error sources related with the equipment stability,
22 some considerations about assembling and positioning the TLS must be taken into account. The
23 TLS is mounted using a tribrach (for leveling the device) on a topographic tripod. The equipment
24 must be properly assembled and stabilized to avoid any disturbance (e.g. vibration) to the scanner
25 position, which can introduce considerable errors in measurements made over large distances; for
26 example, a deviation of 8' produces a deviation of 2.32 m at a distance of 1000 m. This issue is
27 particularly relevant in snow-covered areas, where the weight of the TLS device and the vibration
28 produced during the scanning process can cause movement if the tripod is placed on snow or
29 frozen surfaces. Hence, it is strongly recommended to: (1) mount the tripod on bare rock surfaces;

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3 (2) compact the snow around the tripod; or (3) build stable platforms (e.g. concrete surfaces) where
4 possible.
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7 The last stage before acquiring the point cloud of the study area is the registration of the scan
8 station. This step includes the atmospheric correction and the scan of the targets. Light pulses
9 propagate through the air, with the speed of propagation depending on the air refraction index,
10 which is dependent on atmospheric pressure and air temperature and humidity. For this reason an
11 atmospheric correction is needed. When the equipment is assembled, these three atmospheric
12 variables are measured using a digital thermohygrometer and a barometer; these measurements
13 are used by the software that controls the TLS (Ripprofile®) to correct the light speed propagation
14 value (Ullrich, 2005). Using this correction, light speed in the air can be inferred and the distance to
15 each point accurately obtained by the TLS. After atmospheric correction, the 12 targets of the study
16 area are scanned to provide the reference point coordinates that will be used for the indirect
17 registration. In this study, atmospheric correction and target coordinates acquisition were repeated
18 at 1 or 1.5 hour intervals. The purpose of this, is to detect instability problems, and atmospheric or
19 snow surface changes. This time interval is consistent with the recommendation of repeating both
20 operations at intervals of less than two hours (Prokop, 2008).
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34 Once all the above mentioned steps have been completed, the scan of the study area can be
35 started. First, the area of interest is selected and the TLS is aimed at the two opposite corners of
36 the area to be scanned. In this way a rectangular area is chosen, in which a virtual spatial matrix
37 having the desired resolution is generated. Several factors should be considered when selecting the
38 appropriate spatial resolution. The TLS works on polar coordinates, and automatically transforms
39 the acquired point coordinates to the Cartesian coordinate system. It is important to differentiate
40 between angular and linear resolution. While angular resolution remains constant, linear resolution
41 will change depending on distance to the TLS. Thus, the greatest and least distances in the
42 selected area should be measured (using the TLS), and averaged to estimate the middle distance.
43 Following this, the angular resolution is calculated for the desired linear resolution, and this angular
44 resolution is used for selecting the angular step width for the scan. As the working frequency of the
45 TLS depends on the operational mode, which is dependent on distance, the acquisition speed
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3 depends on the average distance between the TLS and the selected area. As the scanning duration
4 is limited, a compromise between resolution, operational mode and the area to be scanned is
5 needed.
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9 In this study, the average resolution in point cloud acquisition (spacing between points) was 0.35 m,
10 with resolutions between 0.1 m (areas nearest to the TLS) and 0.6 m (areas furthest from the TLS).
11 To achieve those values, the angular scanning resolution has been selected depending on the
12 range between the scanned surface and the TLS, as it is presented in Table 1. In TLS literature the
13 standardized unit for measuring point cloud density corresponds to points per square meter
14 (points/m²). For a normal incidence angle to the scanned surface (0° between laser beam direction
15 and the surface normal vector, see upper panel of Figure 2) the number of points/m² is maximum.
16 When the incidence angle increases the number of points/m² is reduced, as shown in Table 1. This
17 study presents a final cell size value of 1m, and therefore the number of points/m² can be directly
18 related to Table 1 values. From this table, it is clear that only for the highest incidence angles at
19 large distances (75° at 1000 m), the points cloud density is inadequate (lower than 1 point/m²),
20 which occurs in a negligible area of the catchment. To summarize, the average number of points
21 scanned in each survey was around 4 million. Moreover, in order to scan the entire study area,
22 respecting the time interval of 1.5 hours in target acquisition and atmospheric correction, between
23 six and eight scans (from both scan stations) were undertaken.
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38 It is important to consider the beam width, because the laser beam divergence (0.8 mrad = 0.046°)
39 leads to an elliptical footprint for what is considered to be a single point (Figure 2). This means that
40 each point is the average distance to several points of real topography with the calculation
41 depending on the operational mode of the TLS. As the object distance increases so does the
42 footprint, but distance is not the only variable affecting footprint area. When the incident angle
43 between the laser beam direction and the normal vector of the scanned surface increases, the
44 footprint area also increases (upper panel in Figure 2). For instance, for a distance of 200 m and an
45 incident angle of 0°, the footprint area is 0.02 m², increasing to 0.11 m² with an incident angle of
46 80°. For these angles, these areas increase to 0.5 m² and 2.9 m² for a distance of 1000 m, and 8
47 m² and 46 m² for a distance of 4000 m respectively (Prokop, 2009). The relationships between
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3 incident angle and terrain reflectivity affects pulse intensity reflection (Wagner et al., 2004), so
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5 varying behavior of the signal return occurs depending on terrain properties. Another consideration
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7 in beam divergence occurs when the maximum resolution (0.018°, the minimum angular step width)
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9 is selected, as adjacent footprints may overlap. However, it has been established that the optimal
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11 overlap of footprints is 85% of laser beam divergence (Lichti and Jamtsho, 2006) and the TLS used
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13 in this study satisfies this overlapping optimum. Nevertheless, recent studies have suggested that
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15 footprint increases at high incidence angles has a negligible effect on the quality of final point cloud
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17 (Schürch et al., 2011). Therefore it is an effect that must be considered but it is not a limiting factor
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19 of the data quality.

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21 The emission spectrum of the sun includes near infrared wavelengths. Thus, if there is sunlight
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23 reaching the scanned area, or sunlight impacts directly on the TLS receptor, the signal–noise ratio
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25 could affect signal detection (Wehr and Lohr, 1999). If this effect is combined with high incident
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27 angles on wet snow surfaces (angles > 30°), no signal is detected by the TLS detector (Prokop,
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29 2008). Furthermore, with scanning at large distances (> 1000 m), some acquisition points are lost
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31 because of the combination of sunlight and large incident angles. It has been observed that this
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33 problem is largely avoided when no sunlight affects the scan, so for areas with signal detection
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35 problems (long range and large incident angles) the scan must be performed in the absence of
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37 sunlight (at sunset or during the night). At this point, it is interesting to note that no effect on signal
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39 detection has been observed with the different snow albedos given during the snow season.

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41 Finally, for an optimal results the TLS should not be operated during unstable weather conditions
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43 (e.g. cloudy, rainy or windy days), because of erroneous points being acquired during precipitation,
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45 destabilization of the system by wind, and the absorption of near infrared wavelengths by clouds.

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47 When a scan has been completed, pictures are acquired using a digital camera coupled to the TLS.
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49 Images are useful for obtaining RGB (red/green/blue) information for each point, which can be used
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51 for further analysis and verification. For example, this information is useful to account SD errors in
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53 the limits of snow patches (not presented in this paper).

5. Data processing

Several studies have highlighted the importance of choosing the appropriate geo-referencing procedure (Williams, 2012), and the need to establish a rigid geodetic network for reducing misalignments and systematic errors (Rub, 2007). For geo-referencing data we used an indirect procedure, named indirect registration, that uses reference point coordinates (Reshetyuk, 2006) where targets are located. The point clouds obtained are recorded in the relative coordinate system of the TLS (SOCS). Thus, targets enable the transformation of coordinates for each point cloud into a project coordinate system (PRCS) common for all scan stations of a single survey. Subsequently, transformation of the point cloud into a global coordinate system (GLCS) is needed to generate a geo-referenced DTM.

Based on the coordinates of the targets, a transformation matrix (translation and rotation matrix) is generated (using Riprofile® software), which may merge point clouds from the different scan stations and the SOCS coordinate transformation to PRCS, and finally to GLCS. Each matrix calculation is associated with a standard deviation calculation that provides information about the error in target coordinate acquisition for all registration processes during the survey project. This error increases with distance. In this study we found that the standard deviation was approximately 0.008 m, which increased to 0.03 m when the global coordinates of the target were included (note that errors in global coordinate acquisition increase the standard deviation). This standard deviation is adequate for the study purpose, because with this value, errors originated in point cloud registration are below the SD differences, even at large distances (for example, at 730m, the furthest target from scan stations, an average altitudinal error on its location of 0.023m was obtained).

From the obtained point clouds for every survey day, a DTM is derived. Given the high density of points in our study area and the relatively smooth topography of the catchment, point data were directly rasterized at a grid size resolution of 1 m². In this manner, for every single survey, the elevations of all points within each raster cell were averaged. The cells with no point information were discarded and assigned as "no data". Raster masks were created to avoid erroneous calculations for areas in shadow.

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3 Finally, each SD map was obtained by subtracting the snow free catchment DTM (scanned in July)
4 from each DTM created for the six survey days from mid-February to the end May. This is the last
5 product obtained out of the presented methodology.
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10 11 **6. Data Analysis**

12 The SD maps obtained for each date are presented in the feature map. The distribution of the
13 snowpack reveals the high variability of snow, even over very short distances. For example, in
14 periods of very little snow accumulation (as occurred in winter–spring 2011–2012) there are
15 concavities in the terrain that accumulate more than 5 m of snow depth and, at short distances, the
16 SD could be reduced to less than 1 m. In addition, it is clearly observed how in those concavities
17 the snow lasts until the end of the melting period. Qualitatively, this evolution can be seen in the
18 snow distribution maps for 2 May 2012 (obtained from measurements made immediately following
19 an accumulation event) and the two subsequent surveys (14 and 24 May 2012), made when
20 meteorological conditions led to rapid snow melt. Thus, if the snowpack evolution is monitored
21 during the snow season, it is clear that SD maps represent a powerful data base for studying snow
22 cover variability and also for studying snow accumulation and fusion patterns, relating the obtained
23 maps with topographic variables, and meteorological data, if these are available.
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36 Many authors have analyzed the uncertainty in data derived from TLS in relation to (1) errors in
37 point cloud acquisition (Abellán et al., 2009; Hodge et al., 2009) and (2) in DTM generation (Lane et
38 al., 2003; Wheaton et al., 2010). More recently, Schürch et al. (2011) concluded that the elevation
39 ambiguity in points within the same raster cell obtained from different scan stations, is principally
40 originated in errors of point cloud registration and surface roughness. Despite the differences in
41 scale with other studies and the particularities of Izas Experimental Catchment, those works are
42 highlighting the importance of establishing an appropriate experimental setup and following a well-
43 established procedure for minimizing error sources as it is presented here. Considering Schürch et
44 al. (2011), elevation ambiguities could be related with errors from point cloud registration and
45 surface roughness. In this study, instead of considering this numerical analysis of errors in the DTM
46 generation; the altitudinal differences in raster cells for the base DTM (i.e., snow free) and the
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3 obtained DTMs snow-covered, are compared to direct measurements of snow depth, which
4 represents a good option to consider errors of final SD maps.
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7 In order to evaluate the quality and the accuracy of the obtained snow maps from TLS, manual SD
8 measurements were made during each survey. However, it was not possible to directly relate the
9 locations of manual measurements (determined using a hand-held GPS (Garmin GPSmap 60CSx),
10 which had an estimated error of ± 5 m) to those of scanned measurements, where the acceptable
11 error should not exceed a few tens of centimeters. Moreover, site-specific SD acquisition should
12 take into account that high snow variability occurs over short distances and is affected by
13 topographic irregularities at very small scales (decimeters; e.g. the presence of rocks, bushes and
14 other terrain features).
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17 For these reasons, three "validation" plots were defined in the catchment, having average distances
18 from the TLS of 200 m, 400 m and 800 m. The plots were 20×20 m and included flat, concave and
19 convex areas. The SD in these plots was measured using a 5m long anodized aluminum snow
20 probe in a grid of 2×2 m. Using this approach it was possible to assess whether information
21 obtained using the TLS accurately represented the mean and the variability of the snowpack depth
22 at the plot scale, and also enabled assessment of whether distance to the scan position affected the
23 accuracy of the acquired data. López-Moreno et al. (2011) used an extensive dataset of manually
24 measured snow variability to demonstrate that using only eight SD measurements in a plot of 10x10
25 meters was enough to drastically reduce the error when estimating the measured average ground
26 truth and the variance at the plot scale (below 5% in 75 % of analyzed cases). Similar conclusions
27 can be derived from the study of Fassnacht and Deems (2006) that compared different resolutions
28 and resampling methods to quantify local snow variability and scale breaks. These results suggest
29 that the different resolutions of manual (2 meters) and TLS snow measurements (1 meter) should
30 not affect seriously the comparison of average snow estimation and its variability in the validation
31 plots.
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34 The non-parametric Whitney-Wilcoxon statistical test (Siegel and Castelan, 1988) was used to
35 determine the possible statistically significant differences between SD measured manually and from
36 TLS at a significance level of 0.05. Differences between SD measured manually and from TLS were
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3 not statistically different in more than 80% of the cases, with only three exceptions: for medium
4 range the 02/04/2012 and the 17/04/2012; and for long range the 14/05/2012 (marked with * in
5 Figure 5), so it is reasonable to compare TLS and manual measurements, as these do not show
6 differences (at a significance threshold of 95%). Figure 5 shows box plots representing the mean
7 and variability of the snowpack measured using the manual snow probe, and the values of the TLS-
8 based gridded maps measured on each surveyed day. Accurate snowpack quantification was
9 achieved at the three considered scanning distances, with a mean absolute error of 0.07 m. Error
10 does not seem to be directly dependent on scanning distance as they are 0.06 m, 0.09 m, 0.07 m
11 for 200m, 400m and 800m respectively.
12

13 Some differences can be found in the variance of the snowpack measured with TLS and those
14 obtained with manual measurements (Figure 5). In general, higher variances are observed in the
15 plots 400 m and 800 m distant from the TLS in both measurements, which corresponded to areas
16 having more complex topography relative to the flat conditions that dominated plots at 200 m
17 distance.
18

19 Prokop (2008) found a SD accuracy within 0.1 m for distances < 500 m. The accuracy increased for
20 shorter distances, with a maximum SD deviation of ± 0.08 m and a mean deviation of 0.045 m at a
21 distance of 300 m. This explains why TLS is considered the best technology for studying snow
22 distribution at medium distances (up to 500 meters) (i.e. Lundberg et al., 2008). Our study has
23 revealed that mean errors lower than 0.1 m can be also obtained at longer distances, as it has been
24 demonstrated for 800 meters. These results indicate that the protocol presented in this study
25 minimizes error sources, and suggests the applicability of the technique for large distances (up to
26 800-1000m).
27

28 **7. Conclusions**

29 The SD maps generated in the study enable improved monitoring and understanding of snow
30 dynamics in the study area. For this area of 55 ha they provide the highest resolution SD
31 distributions maps (spatial resolution 1 m) available for the Pyrenees. In addition, the maps are not
32 subject to interpolation uncertainties because cells with no point information were discarded.
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3 The results of the manual and TLS measurements of SD show good performance, with only minor
4 differences in the average values between the two measurement techniques, even for long-range
5 TLS measurements. The increase in SD variability for medium (400 m) and large (800 m) distances
6 are explained by differences in concavities and convexities among the selected plots and is
7 observed in both TLS and manual measurements. The results reinforce the applicability of this
8 technique for studying SD distribution and its relation with topographic variables. For future studies
9 it is recommended that SDs be measured using snow poles fixed in the terrain, which should
10 improve the match between manual SD measurements and the SD derived from the TLS. In this
11 way precise positioning would be achieved and manual GPS measurement uncertainties would be
12 avoided.

13 Despite meteorological and stability limitations, the TLS technology was shown to generate
14 accurate high resolution snow distribution maps in alpine terrain, obtained by subtraction of DTMs.
15 Long-range data acquisition from spatially distributed locations in the experimental area is a major
16 advantage of this technology, because it enables data acquisition in remote areas that are
17 dangerous or impossible to reach.

18 The application of the TLS technique in mountainous environments represents a powerful tool for
19 understanding snow dynamics. The main objective in generating SD maps is to represent the
20 spatial variability of snow and snow permanence over time at specific locations in the catchment.

21 **8. Software**

22 Data acquisition using the TLS was achieved using Riprofile® 1.6.2 software (Riegl, Austria). This
23 enabled point cloud overlapping from different scan stations, and the coordinate transformations
24 from SOCS to PRCS and to GLCS. The final point cloud coordinates were exported as ASCII files.
25 Miramon v6.4 software was used for final rasterization of point clouds data, because it is able to
26 manage large databases. Miramon v6.4 was also used for generating the SD maps, and ArcGis 9.3
27 was used to generate the final cartographies.

9. Map Design

The presented maps have been designed for showing snow depth in the Izas Experimental Catchment. In order to improve the understanding of the maps, the information is provided as follows: (i) in the upper part of the figure the title of the map, authors and affiliations are shown; (ii) in the center of the figure the six snow depth maps with the corresponding date of each survey campaign are shown; and (iii) finally in the bottom part of the figure some thematic information is provided: i.e., location of the study area, scales for snow depths (in cold colors) with the appropriate intervals for all obtained maps, positioning accuracy and the estimated mean error in SD estimation, graphical scale, and global coordinate system.

10. Acknowledgements

This work was supported by the research projects CGL2011-27536 and CGL2011-27574-CO2-02, financed by the Spanish Commission of Science and Technology and FEDER; “Efecto de los escenarios de cambio climático sobre la hidrología superficial y la gestión de embalses del Pirineo Aragonés”, financed by “Obra Social La Caixa”; and CTP1/12 “Creación de un modelo de alta resolución espacial para cuantificar la esquiabilidad y la afluencia turística en el Pirineo bajo distintos escenarios de cambio climático”, financed by the Comunidad de Trabajo de los Pirineos. First author is a recipient under the predoctoral grant FPU program 2010 (Spanish Ministry of Education Culture and Sports). The third author was the recipient of the postdoctoral grants JAE-DOC043 (CSIC) and JCI-2011-10263 (Spanish Ministry of Science and Innovation). We offer thanks to www.oshitoaudiovisual.com for the vectorization of the map. The authors wish to acknowledge the editor and the three reviewers for their detailed and helpful comments to the original manuscript.

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12. Figures and tables

Figure 1: The study area showing the location of the Izas Experimental Catchment. The left panel represents the geographical location in the Iberian Peninsula, and the right panel depicts a zoom in of Central Spanish Pyrenees for a better location of the Izas Experimental Catchment.

Figure 2: Time of flight technology and footprint characteristics of the LPM-321 TLS. This is a simplified figure with footprint perspective for different distances and incident angles, where d_i are distances [m], t is time of flight [s], c is light speed propagation [m/s], δ is laser beam divergence [°], α is incident angle between surface normal and laser beam direction [°], and F_i are footprint sizes [m²]. In the upper panel, two footprints for the same distance are represented with two different incident angles, showing the increase of footprint size when incident angle increases. In the main panel, two footprints with normal incident angle between surface and laser beam direction for two distances, show the increase of footprint size with distance.

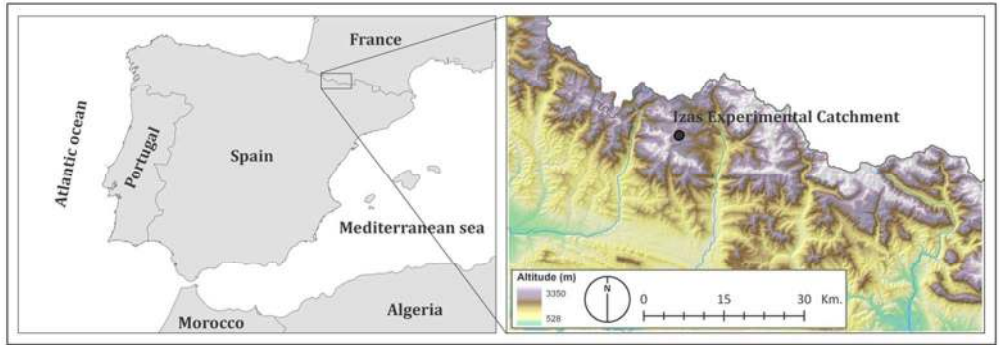
Figure 3: Digital terrain model of the Izas experimental basin showing the locations of the two scan stations (1 and 2) and the 12 targets used in this study. The locations of plots where manual SD measurements were taken are shown.

Figure 4: Point cloud with topographic shadows from Riprofile © software. This is a perspective view from a three-dimensional view. The magenta color shows areas visible from scan station 1, and cyan shows areas visible from the scan station 2. Areas with points of both colors (magenta and cyan) are visible from both scan stations. The black shading shows areas in which no measurements were possible and the white lines depict the visible edges (from the image point of view) of the bounding box that enclose the point cloud.

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3 **Figure 5:** Comparison of snow depths determined by the TLS and manual methods. The red line
4 shows the mean value, the black continuous line shows the statistical mode, the boxes show the
5 25th and 75th percentile range, the black dots show the 5th and 95th percentiles, and the whiskers
6 show the 10th and 90th percentiles. * indicate those cases in which statistical significant differences
7 exist between TLS and manual measurements.
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15 **Table 1:** Angular resolutions, linear resolutions, and the density of the obtained point cloud at
16 different distances from the TLS. *Note that values for 200m and 600m ranges are duplicated,
17 because 200 m range is normally the limit for using near or medium range, and 600 m range is the
18 limit for using medium or long range.
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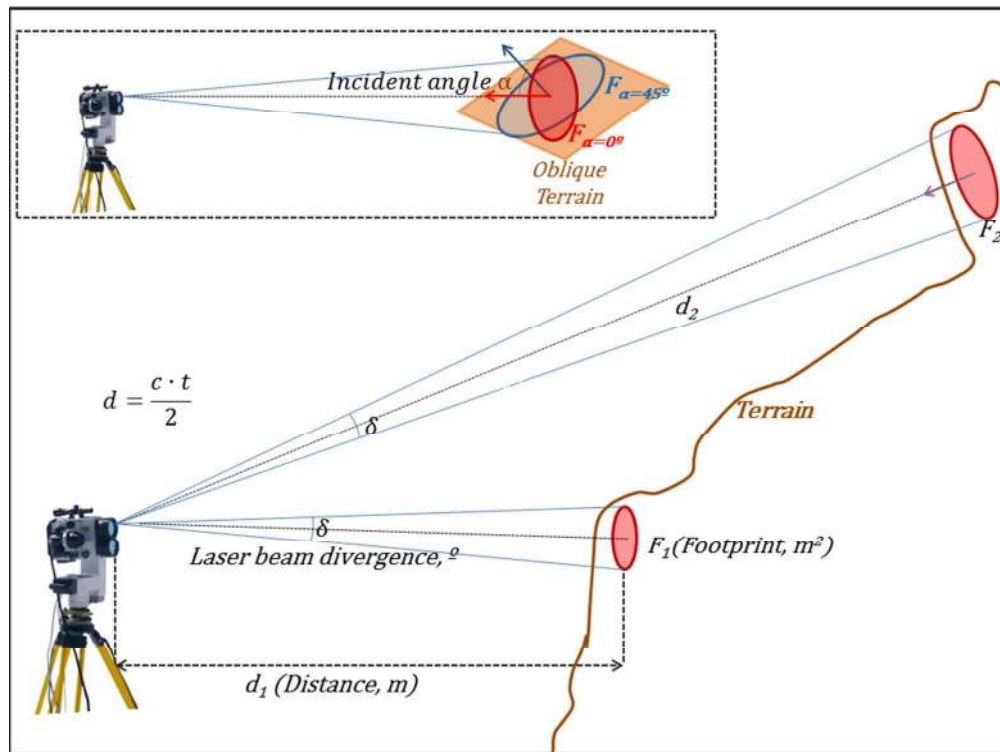
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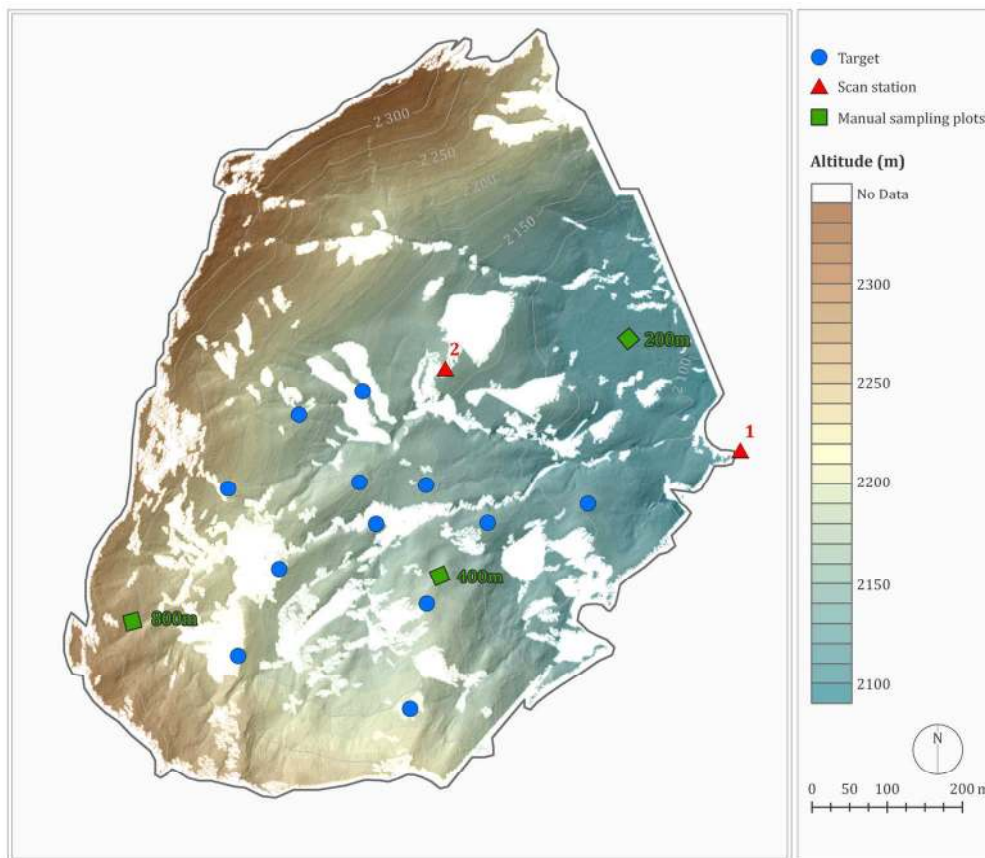
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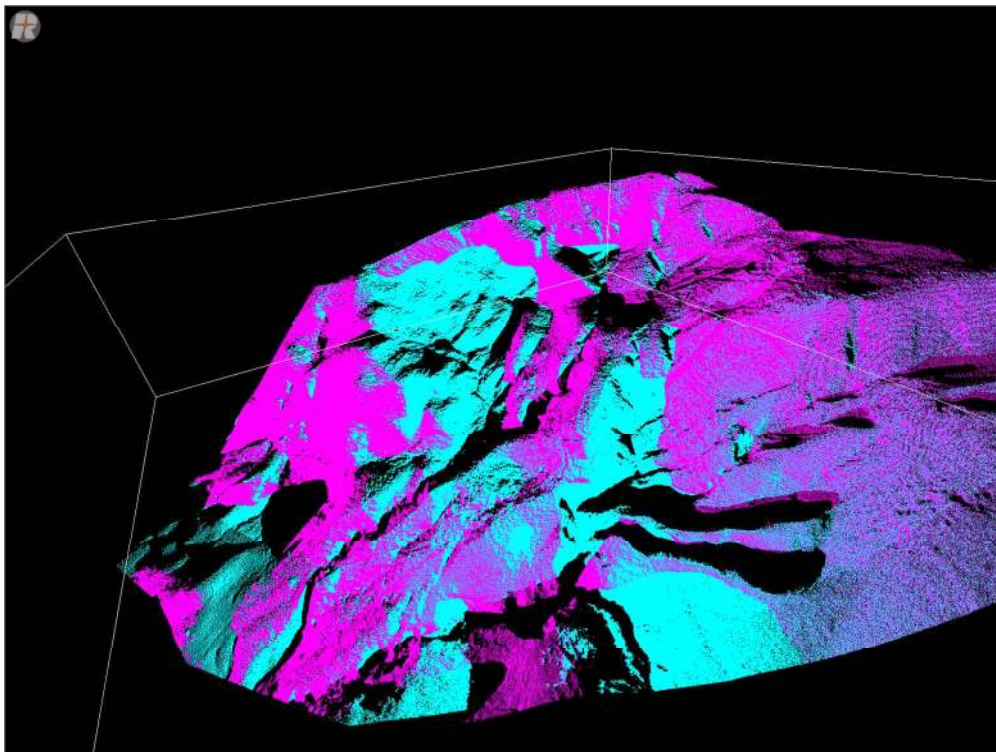
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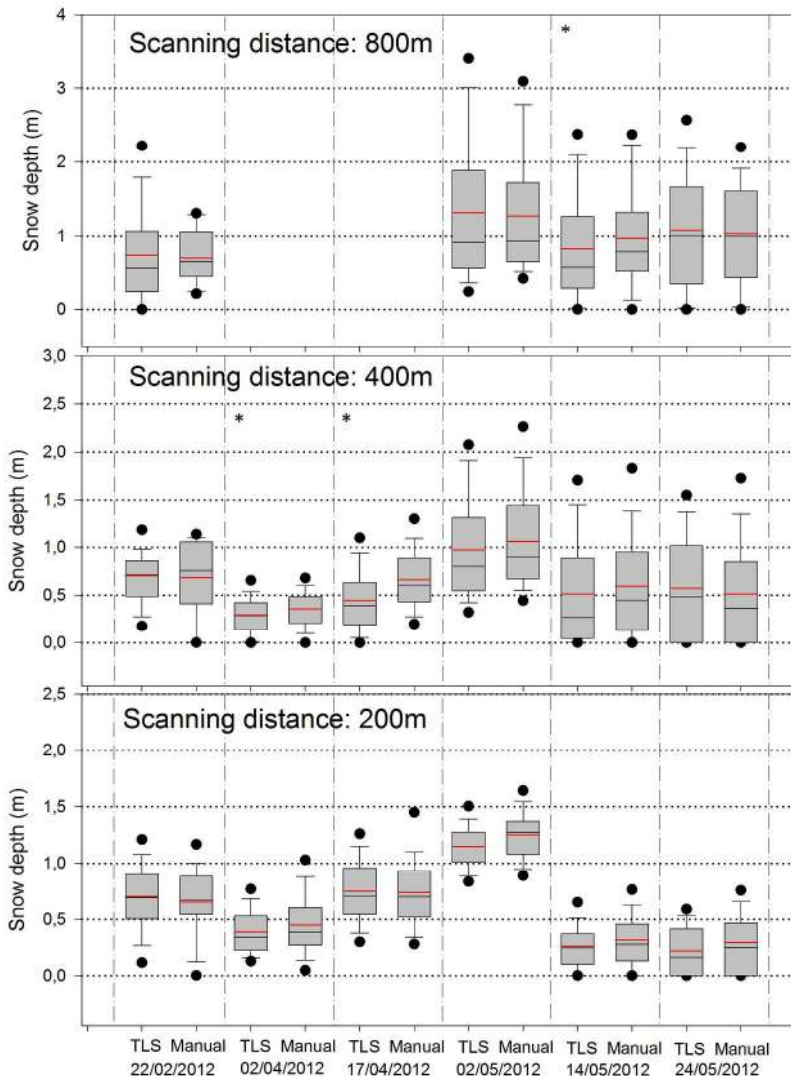


135x101mm (300 x 300 DPI)

View Only

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Comparison between TLS / Manual depth acquisition



176x250mm (300 x 300 DPI)

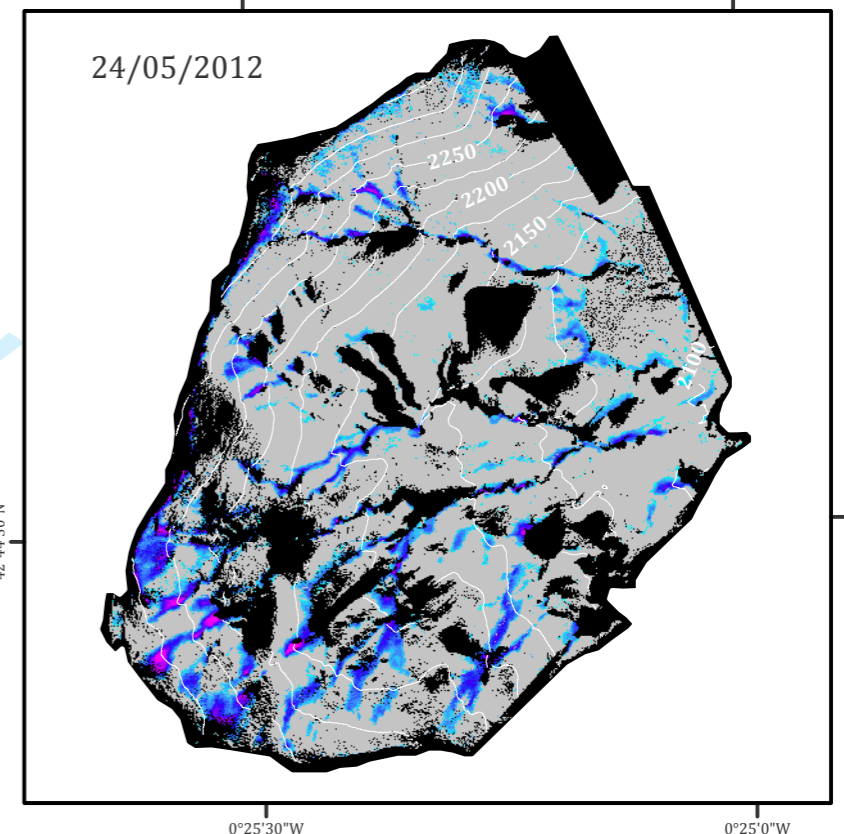
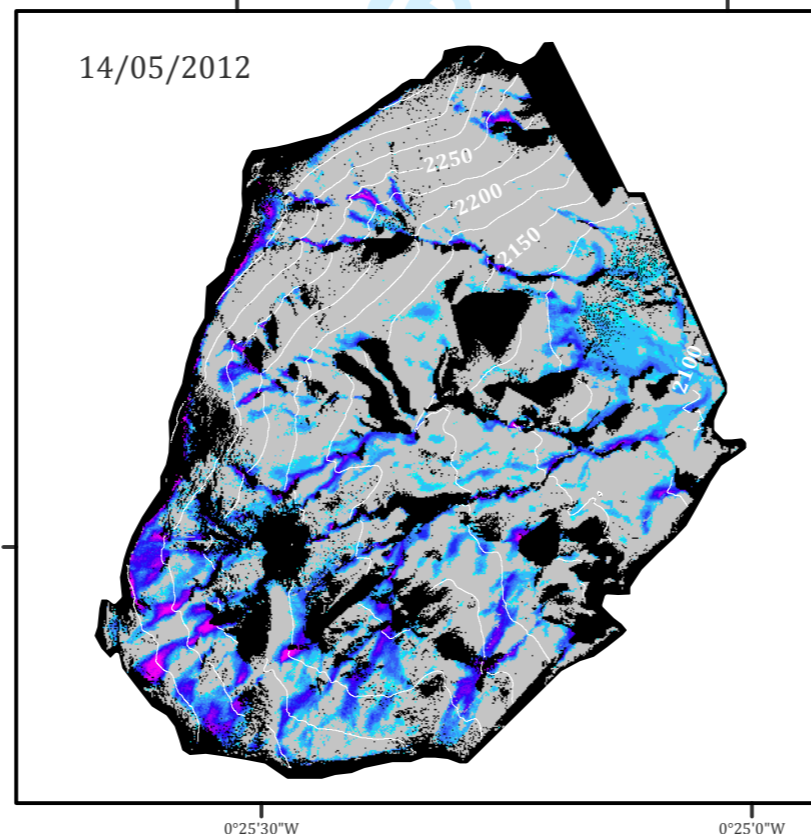
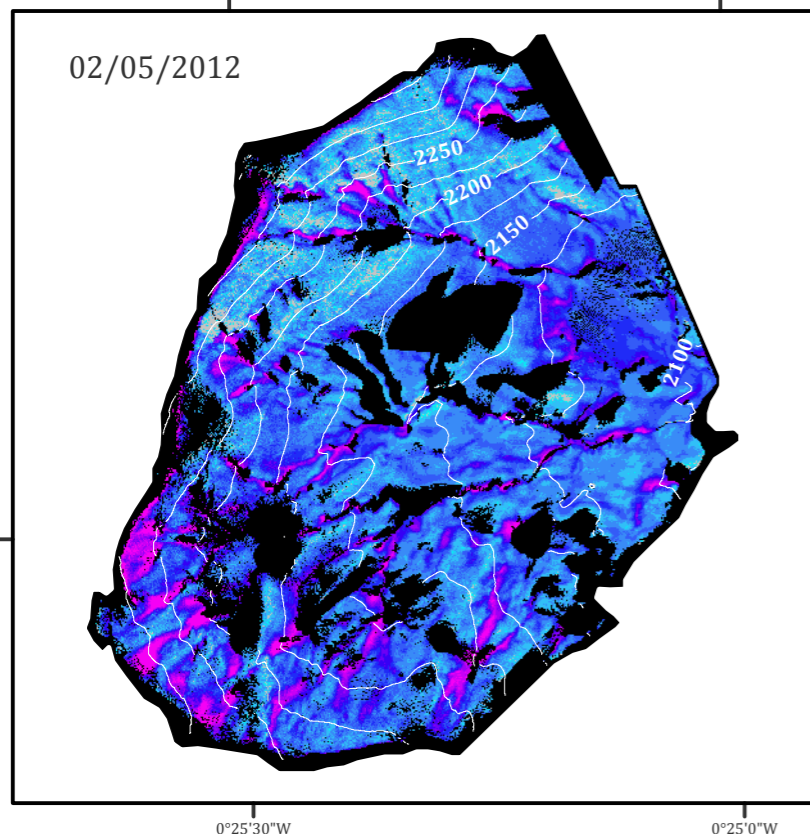
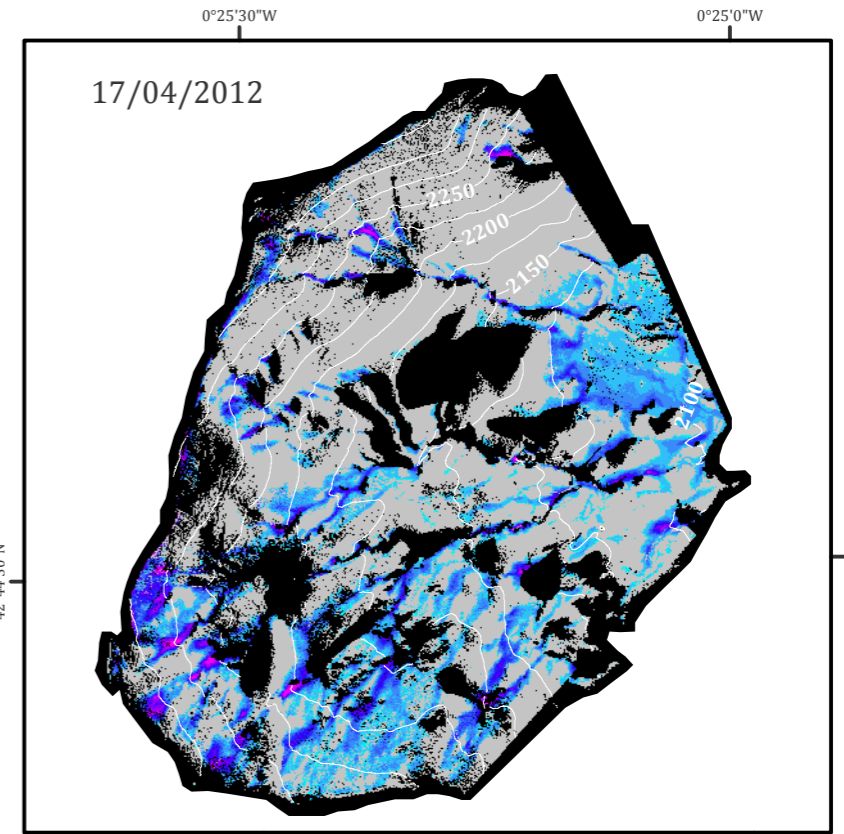
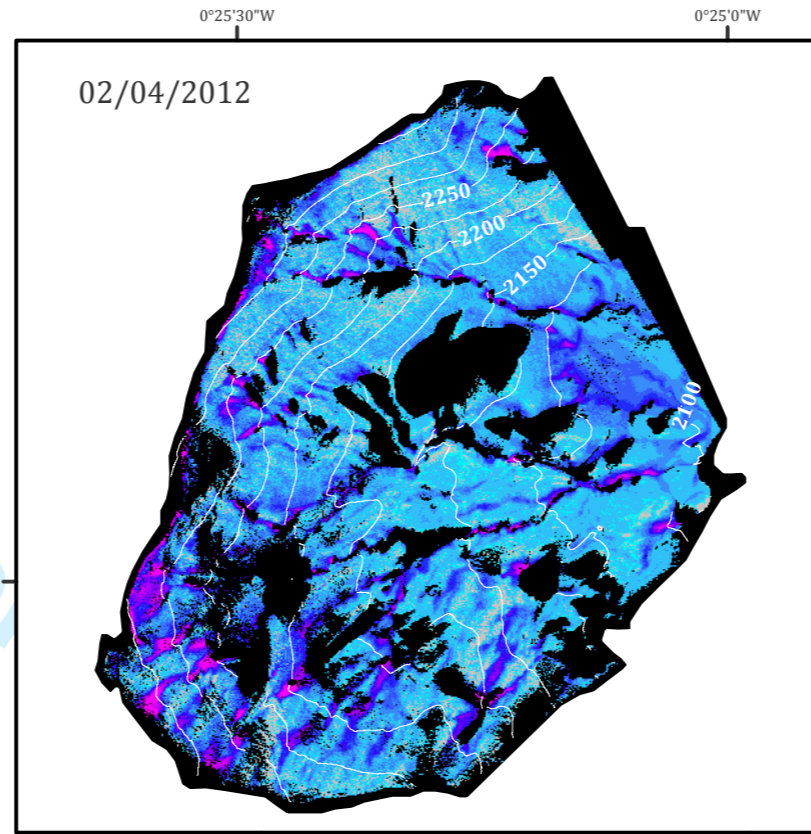
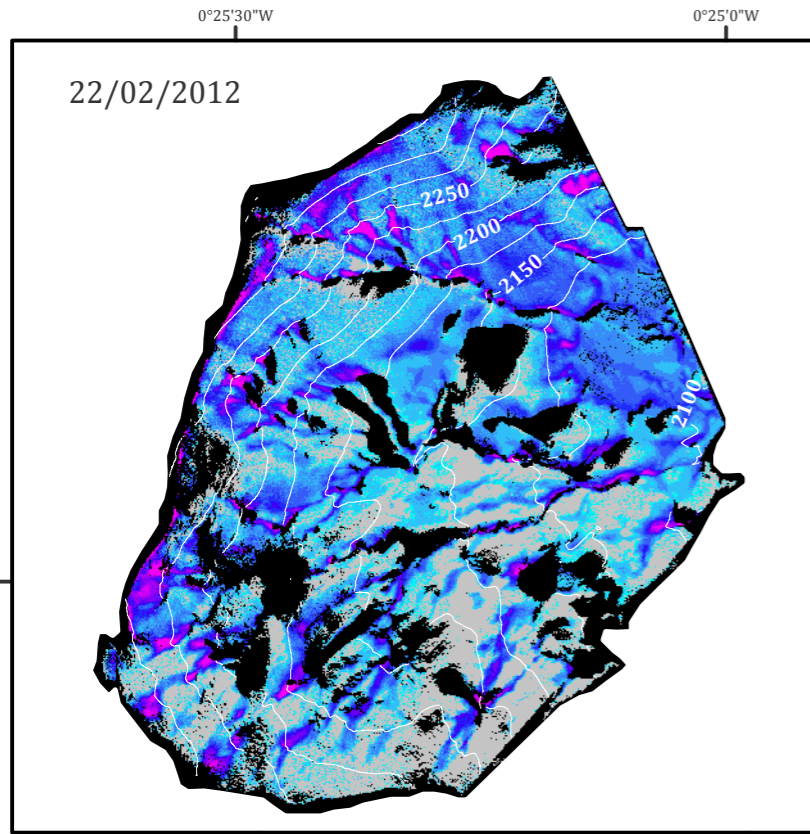
Range (m)	Angular Resolution (°)	Linear Resolution (m)	Points/m ² for different incidence angles (α)				
			0°	15°	30°	60°	75°
1000	0,036	0,63	2,5	2,4	2,2	1,3	0,9
900	0,036	0,57	3,1	3,0	2,7	1,6	1,1
800	0,036	0,50	4,0	3,8	3,4	2,0	1,4
600	0,036	0,38	7,0	6,8	6,1	3,5	2,4
*600	0,064	0,67	2,2	2,2	1,9	1,1	0,8
400	0,064	0,45	5,0	4,8	4,3	2,5	1,7
200	0,064	0,22	20,0	19,4	17,4	10,0	6,9
*200	0,11	0,38	6,8	6,6	5,9	3,4	2,3
100	0,11	0,19	27,1	26,2	23,5	13,6	9,3
50	0,11	0,10	108,5	104,8	94,0	54,3	37,1

Snow Depth in Izas Experimental Catchment

J. Revuelto¹, J.I. López-Moreno¹, C. Azorín-Molina¹, J. Zabalza¹, G. Arguedas² and S.M. Vicente-Serrano¹ (2013)

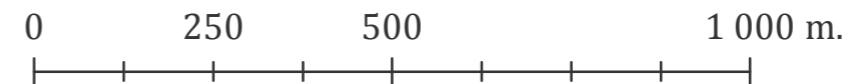
Instituto Pirenaico de Ecología - CSIC : 1
Escuela Politécnica Superior de Edificación de Barcelona (UPC) : 2

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Snow Depth (m)

Positioning accuracy: 10 cm
Estimated mean absolute error: 0.07m



Cell size for gridded data: 1 m
Contour interval (m): 25
Universal Transverse Mercator
UTM Zone 30N

