

Advances in horse morphometric measurements using LiDAR

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Abstract.

Zoometric measurements have a potential value in differentiating between individuals and within populations. The measurement of body size in horses and livestock plays a significant role in functional longevity, production, and reproductive performance and health. In this context, the measurements obtained without contact by detection systems and visualized by computers could represent a great advance over conventional measurements that are tedious, time consuming and stressful for the animals. This study presents a new approach to taking zoometric measurements of an animal's body based on digital three-dimensional modelling. The capture of the data series was carried out by a LiDAR sensor. The 16 laser beams of the sensor were able to fully scan a horse, performing a 3D reconstruction of the horse's side, through which body measurements were obtained. Five Pura Raza Española horses (PRE) (3 stallions and 2 mares) with ages ranging between 5 and 18 years old were scanned. The PRE is the most recognized native Spanish horse population for its census (national and international),

30 cultural and socioeconomic importance. For each horse, 17 zoometric measurements (linear
31 and angular) were taken both manually and using the LiDAR-based system to check the
32 usefulness of this non-invasive technology in obtaining quick livestock measurements while
33 causing minimal stress to the animals. Of the 17 zoometric measurements obtained manually
34 and with the sensor, 10 (58.82%) had a mean relative error that ranged between > 0 and <10 ; 5
35 (29.41%) had an error that ranged ≥ 10 and <20 ; and two (As and ACr) had an error ≥ 20
36 (11.76%). A total of 82.5% of the traits studied had an accuracy (v^2) lower than 5%. Therefore,
37 although this approach could still be improved, it verifies the viability of noncontact
38 measurements of large livestock.

39

40 **Keywords.** Zoometric measurements, digital three-dimensional modelling, LiDAR sensor,
41 Velodyne VLP-16

42 **1. Introduction**

43 The use of remote sensing in agriculture is not new; it dates back decades. However,
44 recent technological advances in sensors, instrumentation, and wireless networks have brought
45 innovative levels of monitoring into raising livestock (Guo et al., 2017, 2019; Le Cozler et al.,
46 2019). Sensors measuring animal welfare, health, growth rate, and physiological features,
47 including shape, size, and weight, either alone or in combinations, have been applied in animal
48 breeding. Kawasue et al. (2013) used three Kinect sensors to investigate the weight and size of
49 cattle, as well as their postures and shapes, with accuracies of up to 93% relative to manual
50 measurements. Other authors have also evaluated the body shape and temperature of black
51 cattle using a thermal camera and a Kinect sensor (Kawasue et al., 2017), the correlation
52 between manual measurements and stereo vision in horses (Pallottino et al., 2015), and the

53 combination of a binocular stereo vision system with the LabView development platform to
54 estimate pig body size and weight in indoor farm conditions (Shi et al., 2016).

55 The horse industry plays a part in national, state, and local economies in Europe, North
56 America, and Canada (Cross, 2019). The equine sector is diverse, involving agriculture,
57 business, recreation, racing, competitions, and the generation of specialized skills and general
58 employment across the board. Breeding, competition, and leisure activities involving horses are
59 important business concerns, and while horses are no longer used for primary transport, they
60 remain important assets. Although not currently popular in most countries, the consumption of
61 horse meat is increasing in several Western European countries due to its availability and
62 recognized nutritional value and mainly as an alternative to the traditional consumption of red
63 meat (Belaunzaran et al., 2015).

64 Equine production is one of the main agricultural activities in Spain, responsible for 0.5% of
65 the gross national product (Sánchez-Guerrero et al., 2018), with the Pura Raza Española horse
66 (PRE) being the most important breed in terms of historical census (328,706 horses in 65
67 countries) and economic impact on international trade (Sánchez-Guerrero et al., 2016). Typical
68 genetic selection programmes are based on functionality (dressage) and morphological
69 characteristics. Therefore, obtaining accurate zoometric measurements, characterized by high
70 repeatability and low variance among observers, can be considered ideal conditions in the
71 context of an animal breeding programme (Duensign et al., 2014). In the case of horses, the
72 close connection between biometric parameters, locomotion characteristics, and sports
73 performance is often the only tool for genetic improvement (Pallottino et al., 2015) and as an
74 indicator of functional longevity. In recent decades, indirect conformation assessment systems
75 based on linear morphological traits have been implemented for many equine breeds (Kuhnke
76 et al., 2019; Folla et al., 2019). This system has a series of advantages, as the scoring scale of
77 each region covers its biological range and a wide range of numerical classes can be used,
78 which allows subsequent statistical treatment of the data in a continuous scale. However, highly

79 trained qualifiers are required to apply the linear morphological system, and the relationship
80 between zoometric measurements and the linear scale used has not been established. The
81 main goal of the PRE breeding programme is to improve not only animal functionality but also
82 its conformation for sport performance (Sánchez-Guerrero et al., 2016). Therefore, in the PRE,
83 the importance of conformation has been shown in studies carried out to demonstrate its
84 relationship with dressage (Sánchez-Guerrero et al., 2017; Solé et al., 2013), functional
85 longevity (Solé et al., 2017), and genetic improvement of morphofunctional traits (Sánchez-
86 Guerrero et al., 2017).

87 In the last decade, several computerized techniques have been proposed for the collection
88 of biometric data through images (Wu et al., 2004; Viazzi et al., 2014). Novel three-dimensional
89 systems can solve the problems posed by conventional two-dimensional vision systems,
90 including stereo photogrammetric techniques. However, these photogrammetric systems are
91 difficult to implement.

92 In recent years, new techniques that provide point clouds in a fast, non-invasive and
93 inexpensive way, such as the Kinect v2® (Microsoft Corp., Redmond, Washington) or Xtion Pro
94 (USUSTeK COMPUTER Inc., Taipei, Taiwan), as well as infrared (IR) light (Salau et al., 2017;
95 Kawasue et al., 2017; Viazzi et al., 2014; Mortensen et al., 2016; Guo et al., 2017; Pezzuolo et
96 al., 2018; Song et al., 2018; Guo et al., 2019), are being used to develop new method to
97 automate the collection of morphological information from livestock. In fact, Pezzuolo et al.
98 (2019) performed a metrological analysis of the Structure from Motion (SfM) approach, low-cost
99 LiDAR scanning, and the Microsoft Kinect v1 depth camera for three-dimensional measurement
100 of an animal's body, with specific reference to pigs. The results obtained demonstrated the high
101 potential of the 3D Kinect and a higher root mean square (RMS) value of LiDAR with respect to
102 Kinect and SfM, probably due to the collection approach based on individual profiles instead of
103 surfaces.

104 However, despite progress in the three-dimensional reconstruction of the bodies of animals,
 105 no robust descriptor for the automatic estimation of conformation has currently been found in
 106 equine species. Therefore, the main objective of this study is to explore a 3D reconstruction
 107 approach for body measurements in horses using point cloud analysis. Our specific objectives
 108 were (1) to develop a portable scanning system for domesticated animals; (2) to obtain valuable
 109 zootechnical information through 3D point cloud analysis (that would avoiding wasting time and
 110 money); and (3) to compare the accuracy between the digital and conventional manual
 111 measurements on five PRE horses.

112

113 **2. Materials and methods**

114 ***2.1. Animals and morphological variables under study***

115 Five PRE horses, belonging to the breeding programme of the “Yeguada Cartujana del
 116 Hierro del Bocado” (Jerez de la Frontera, Cádiz, Spain), situated at latitude 36°41’ north and
 117 longitude 06°09’ west, were analysed (Table 1).

118 **Table 1.** Ages and coat colours of the five Pura Raza Española horses used in this study.

	Stallion 1	Stallion 2	Stallion 3	Mare 1	Mare 2
Age	5 years old	6 years old	8 years old	12 years old	18 years old
Coat Colour	Grey	Bay	Grey	Grey	Grey

119

120 The animals had been used for breeding and dressage exhibitions. To reduce possible
 121 variations, the zoometric measurements taken by the manual method were always carried out
 122 by the same veterinarian. For each horse, 17 zoometric measurements (15 linear and 2
 123 angular) relating to sport performance (Sánchez-Guerrero et al., 2016, 2017) were collected
 124 (Fig. 1).

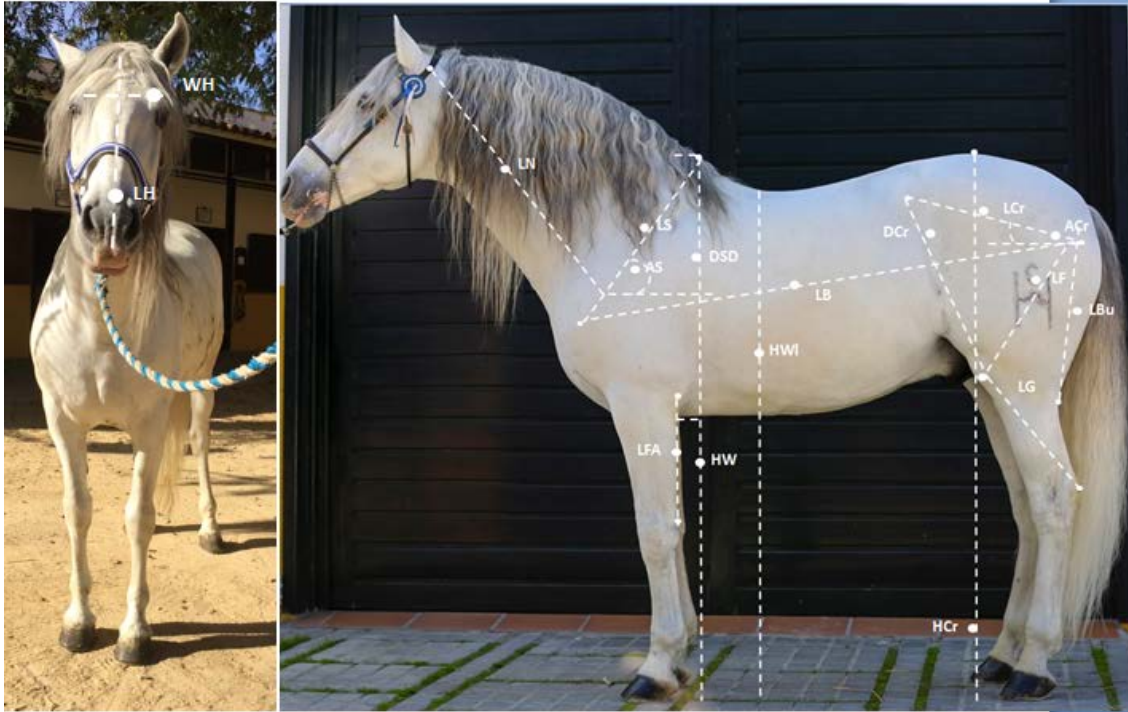


Figure. 1. Graphical representation of the morphological measurements taken in the Pura Raza Espanola stallion 1 of the *Yeguada Cartujana del Hierro del Bocado*. Height at the withers (HW); Height to the lowest point at the withers (HWI); Height at the croup (HCr); Length of the body (LB); Length of the head (LH); Width of the head (WH); Length of the neck (LN); Length of the shoulder (LS); Length of the forearm (LFA); Dorso-sternum diameter (DSD); Length of the croup (LCr); Length of the femur (LF); Depth of the croup (DCr); Length of the gaskin (LG); Length of the buttock (LBu); Angle of the shoulder (AS); Angle of the croup (ACr).

125

126 The morphological measurements taken were as follows:

127 1. Height at the withers (HW): measured from the ground to the highest point of the
 128 withers.

129 2. Height to the lowest point at the withers (HWI): measured from the ground to the lowest
 130 point of the withers.

131 3. Height at the croup (HCr): measured from the ground to the highest point of the tuber
 132 coxae.

133 4. Length of the body (LB): measured between the tubercle of the humerus and the ischial
 134 tuberosity.

- 135 5. Length of the head (LH): distance from the nape to the cranial border of the snout.
- 136 6. Width of the head (WH): distance between the most protruding edge of the zygomatic
- 137 arches.
- 138 7. Length of the neck (LN): distance between the base of the ear and the middle point of
- 139 the spine of the scapula.
- 140 8. Length of the scapula (LS): distance between the wither and shoulder.
- 141 9. Length of the forearm (LFA): distance between the parallel straight lines drawn down
- 142 from the elbow and carpal joint midpoint.
- 143 10. Dorso-sternum diameter (DSD): distance measured from the lowest point in the wither
- 144 decline to the sternal area.
- 145 11. Length of the croup (LCr): distance between the coxal tuberosity at its midpoint and the
- 146 ischial tuberosity.
- 147 12. Length of the femur (LF): distance between the buttock and stifle
- 148 13. Depth of the croup (DCr): distance between the hip and stifle.
- 149 14. Length of the gaskin (LG): distance between the stifle and the hock
- 150 15. Length of the buttock (LBU): distance between the coxal tuberosity of the ilium and the
- 151 ischial tuberosity.
- 152 16. Angle of the shoulder (AS): angle formed by the line from the withers to the shoulder
- 153 with the horizontal.
- 154 17. Angle of the croup (ACr): angle formed by the line from the ischial tuberosity to the
- 155 tuber coxae with the horizontal.

156

157 **2.2. Data acquisition and data processing**

158 *2.2.1 Conventional measurement system*

159 For each horse, zoometric measurements were systematically collected using standard
160 measuring sticks, non-elastic measuring tape and zoometric compasses (Sánchez-Guerrero et
161 al., 2016). All measurements were taken from the left side of the horse while it was standing on
162 a hard surface and flat ground, assuming a natural position. The horses were positioned for
163 measurement with the front legs and hind feet parallel and as near to perpendicular as possible;
164 the toes were in line. No sedatives were used.

165

166 *2.2.2. Measurements based on the point cloud system*

167 The zoometric measurements were acquired through an optical remote sensing technique
168 using a LiDAR sensor. The sensor uses laser light to obtain a dense sampling of the target,
169 producing accurate measurements in three dimensions. The reflection of laser light off the target
170 is detected and analysed by the receivers in the LiDAR sensor. These receivers, consisting of a
171 receiver telescope, filters, and an optical detector, record the precise time between when the
172 laser pulse left the system and when it returned to calculate the limit distance between the
173 sensor and the target. These spatially organized post-processed LiDAR data are known as point
174 cloud data, which is a large data set consisting of 3D point data. Each LiDAR point can have an
175 assigned classification defining the type of object that reflected the laser pulse.

176 The LiDAR used was a Velodyne VLP-16 (Velodyne Lidar, Inc., California, EEUU), which is
177 a high-precision 3D laser sensor LiDAR with a rotating head containing a certain number of
178 semiconductor lasers or laser diodes. Each of the lasers has its own detector (Fig. 2).

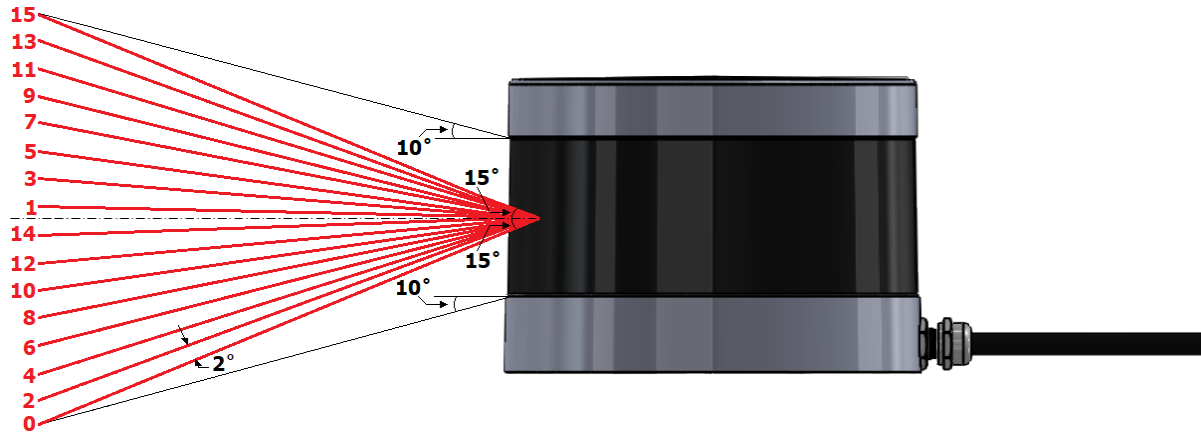


Figure 2. Representation of the lasers of the Velodyne VLP-16

179

180 The VLP-16 sensor measures the reflectivity of an object with 256-bit resolution
 181 independent of laser power and distance within an interval of 1-100 m. Commercially available
 182 reflectivity standards were used for absolute reflectivity calibration, which are stored in a
 183 calibration table within the field-programmable gate array (FPGA) of the VLP-16.

184 The VLP-16 scanner has 16 individual lasers/detectors arranged in a 30° FOV, which yields
 185 a vertical resolution of 2.0°. This sensor has an FOV symmetrical with the respect to the
 186 horizontal plane, and points can be obtained up to 100 m away at a rate of approximately
 187 300,000 per second in single return and 600,000 for dual return. The horizontal FOV is 360°,
 188 with an adjustable rotation frequency between 5 and 20 Hz.

189 Thanks to the divergence of the laser beam, a single shot can hit multiple objects, and
 190 different returns will occur. The VLP-16 has the ability to analyze the multiple returns and report
 191 the strongest return, the last return, or both (dual return or dual mode). Multiple returns occur
 192 when a laser pulse strikes the horse in a location that does not completely block the path of the
 193 pulse, allowing the remaining portion of the pulse to continue to the next seen object.

194 Communication between the sensor and the computer performing the data analysis was
 195 performed was carried out through an interface box with an Ethernet cables and user-assigned
 196 IP addresses.

197 As a robust and easily portable solution, the mini PC used was a Raspberry Pi, a small,
198 low-cost, single-board computer. This system (Fig. 3) is very versatile and allows the installation
199 of several operating systems, whether official or otherwise, as well as different content
200 managers depending on the use for which it is intended.
201

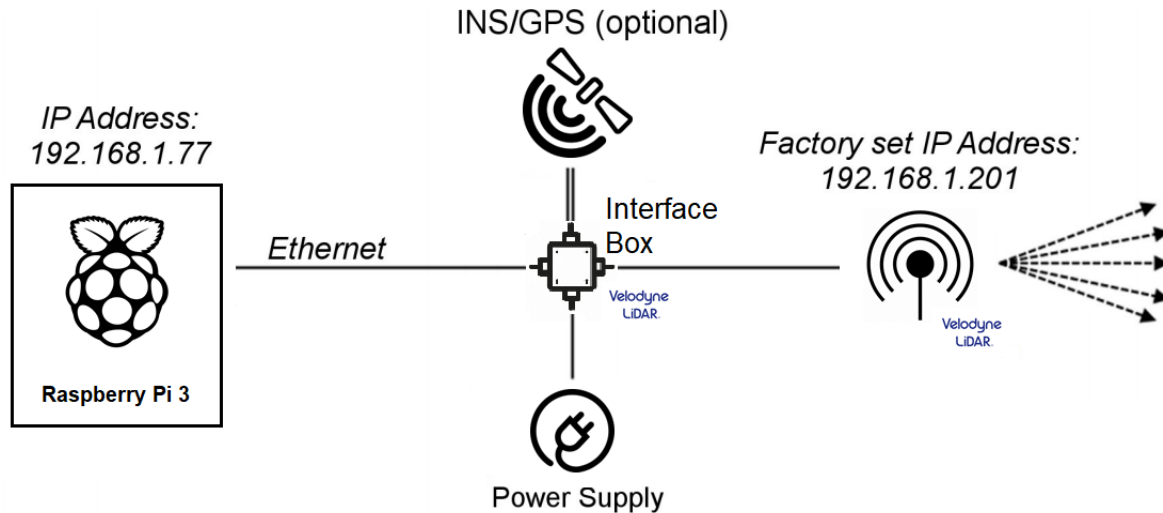


Figure 3. Hardware connections.

202
203 The operating system used was UBUNTU 18.04.2 (Bionic). Ubuntu was chosen because it
204 is an open operating system and the preferred software for development under Linux, especially
205 in terms of artificial intelligence solutions, automatic learning, and deep learning, because it
206 offers a wide breadth of libraries, tutorials, and examples; furthermore, no other operating
207 system offers the same level of support for the latest versions of software and free open source
208 platforms. Additionally, UBUNTU is one of the best optimal platforms for robotics operating
209 systems (ROS) to function correctly and simply.

210 An ROS is an operating meta-system, as it includes an operating system that is installed
211 over another one. This flexible framework has a wide variety of tools, libraries, and packages
212 that considerably simplify the creation of applications for robots.

213 Working with the Ubuntu operating system along with the ROS allows us to obtain
214 information through a command terminal, configure the sensor, the interface and execution
215 time, and work with a simple graphical interface (such as ROS visualization (RViz)), among
216 other functions.

217 RViz is a 3D visualization tool that can represent robots, point clouds, sensor data, and so
218 on. RViz simply reads and interprets the different data contained in the ROS messages.
219 Therefore, it is necessary to have an external generator for these messages, such as a robot, or
220 in this case, a LiDAR sensor.

221 Once RViz was configured, the horse could be scanned. A tripod was used to stabilize the
222 sensor and regulate its height. Once screwed to the tripod, the sensor must be placed far
223 enough away from the horse to capture the animal completely and through as many acquisition
224 channels as possible. The LiDAR is placed on the tripod at a height of 90 cm for the ground,
225 enough height to obtain a complete scan of the horse. This height favors that the sensors rays
226 strike as as little as possible on the ground and on the top of the horse.

227 As the vertical field of view of the sensor covers an angle from -15° below to 15° above the
228 horizontal, the laser was installed in the horizontal orientation, perpendicular to the ground,
229 ensuring a complete sweep of the horse (Fig. 4). The horse was only restrained by a halter and
230 a rope, and the handler attempted to keep it as stationary as possible.



Figure 4. Measuring a mare with the LiDAR system at the *Yeguada Cartujana Hierro del Hierro del Bocado*.

231
232 From the LiDAR sensor configuration screen, it is possible to reduce the vertical field of
233 view, preventing the sensor from scanning the background, focusing only on the surface of the
234 horse. The point cloud was generated by the RViz visualization tool. Through this point cloud,
235 we could find the surface and reflectance of the materials scanned (Fig. 5).

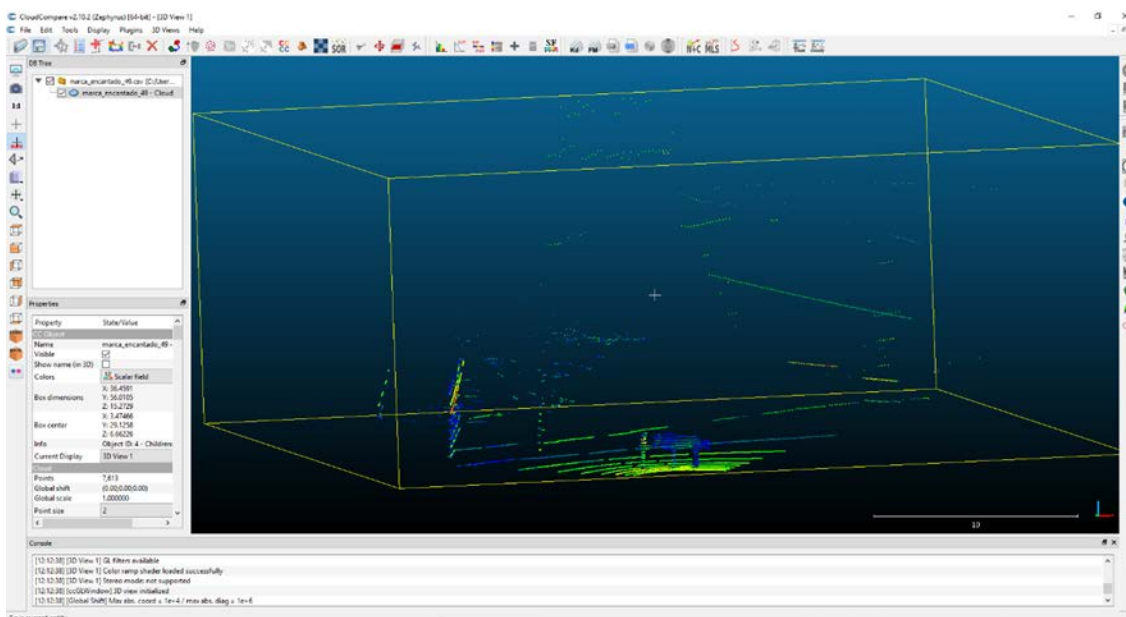


Figure 5. Point cloud from a test scan of a Pura Raza Espanola stallion of the *Yeguada Cartujana del Hierro del Bocado*.

236
237 To be able to work with data in the CloudCompare point cloud processing software, it was
238 necessary to convert the .bag file recorded in RViz into either a comma separated value (.csv)
239 or a .pcd file format. CloudCompare is a free software for processing 3D point clouds and
240 triangulation meshes. It is designed to cope with huge point clouds typically consisting of more
241 than 10 million points and up to 120 million points with 2 gigabytes of memory.

242 As the data were taken in one of the yards of the *Yeguada Cartujana del Hierro del Bocado*,
243 it was necessary to perform manual filtering of the point cloud to facilitate the extraction of the

244 measurements of zoometric variables. The filtering consisted of removing all objects that
245 surrounded the horse from the complete point cloud that made it difficult to take measurements,
246 such as the trees in the yard or the operator steadying the horse. From the point cloud
247 generated and filtered through the CloudCompare program described above, measurements of
248 the racial and functional zoometric variables were manually obtained in cm (Fig. 7).

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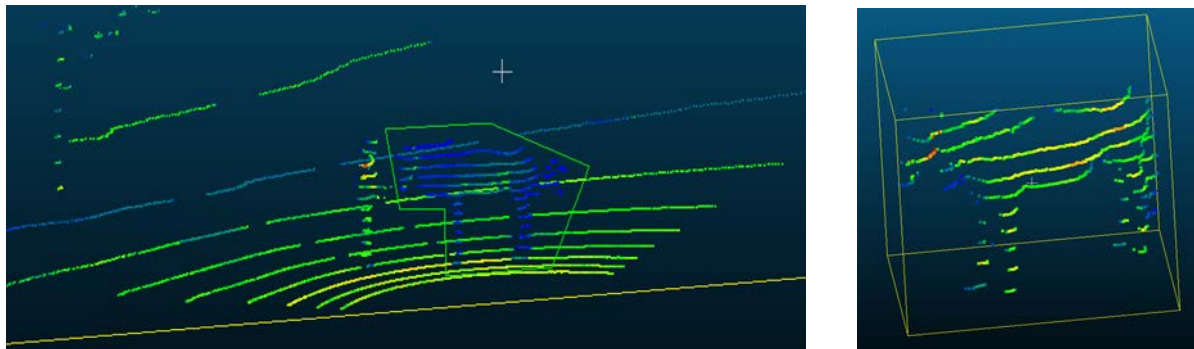


Figure 6. Point cloud filtering.

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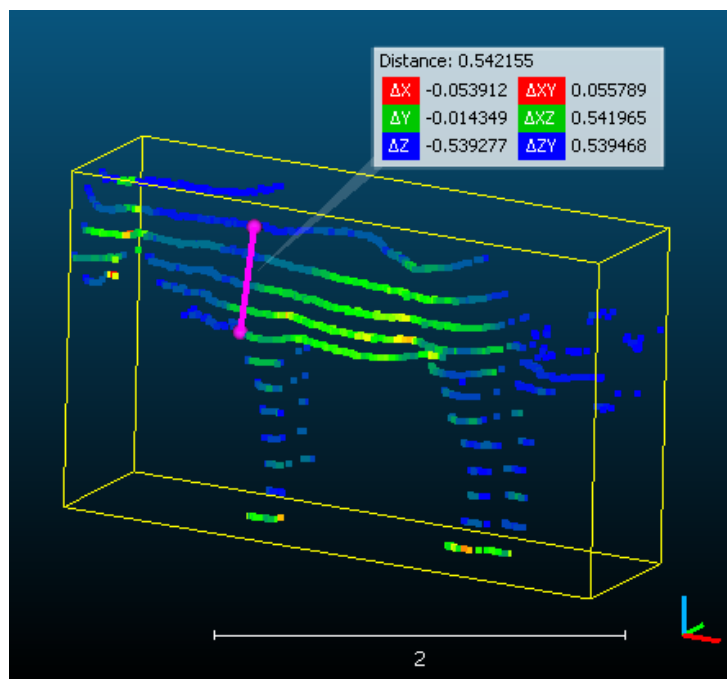


Figure 7. Zoometric measurements of the back length of the Pura Raza Espanola stallion obtained at the *Yeguada Cartujana del Hierro del Bocado*.

252

253 **2.3. Statistical analysis**

254 The relative error formula was used to compare the measurements taken manually and by
255 the sensor to discern the errors made with this latter method for the noncontact measurements
256 of animals:

$$\text{Relative error (\%)} = ((\text{Manually Measurement} - \text{Sensor Measurement}) * 100) / \text{Manual Measurement}$$

257 The accuracy of this method (v^2) was estimated by calculating the percentage of error of the
258 sensor method with respect to the variance (Poly et al., 1967).

$$259 \quad v^2 (\%) = (\text{Variance}_{(\text{Manually Measurement} - \text{Sensor Measurement})} * 100) / \text{Variance}_{\text{Manual Measurement}}$$

260 In addition to the relative error and v^2 , the mean, maximum, and minimum errors were also
261 calculated. The relative error only indicates the deviation over the manual measurements, while
262 v^2 also analyses the variance of these differences between the manual and sensor
263 measurements. Therefore, v^2 is more precise than the relative error, especially when several
264 animals are studied. A low value of v^2 and relative error were desirable. The Pearson
265 correlation coefficient (r) was calculated between the measurements taken manually and by the
266 sensor. Finally, a regression analysis between the coefficient of determination (R^2) and the root-
267 mean-square error (RMSE) between zoometric measurements performed using the
268 conventional procedure (manual) and the LiDAR system (sensor) was also calculated. The
269 procedures were analysed using Statistica 10.0 for Windows package.

270

271 **3. Results and discussion**

272 In general, morphology is considered more important in horse breeding than in other livestock
273 species due to the horse's relationship with sports performance and meat production (Preisinger
274 et al., 1991). To obtain data for the genetic evaluations, all the PRE horses registered in the PRE
275 studbook are tested morphologically. Although manual measurements in equids are the most
276 precise, they often involve a risk for the assessor (as wild, young, and stressed animals could
277 cause them injury) and consuming excessive time and money. In the last decade, 71,349 PRE
278 horses have been tested with manual measurements and linear morphological traits with the
279 consequent expense of time and money that this entails. In addition, only PRE over two years
280 old could be tested, since younger foals are often incapable of standing still while the
281 veterinarian takes the zoometric measurements. This problem also prevents studies of growth in
282 the breed. Therefore, a new methodology, that could be applied quickly, systematically and
283 without touching the animal (even with animals in the field) would be of great benefit to PRE
284 breeding programmes. Therefore, the zoometric measurements in this paper were obtained
285 twice, once manually and once using the Velodyne LiDAR sensor, as shown in Table 2.

286
287

Table 2. Zoometric measurements, relative error, v^2 error and Pearson correlations in five Pura Raza Española horses from Estrepe Cartujana obtained using the conventional procedure (manual) and the LiDAR system (sensor)

M	Stallions 1			Stallions 2			Stallions 3			Mare 1			Mare 2			Global v^2	r
	m	s	r.e.	m	s	r.e.	m	s	r.e.	m	s	r.e.	m	s	r.e.		
HW	156	160	-2.56	159	149	6.29	164	156	4.88	160	149	6.88	163	150	7.98	1.14	-0.32
HWI	151	141	6.62	150	148	1.33	153	155	-1.31	149	146	2.01	153	150	1.96	3.59	0.60
HCr	158	145	8.23	159	155	2.52	159	140	11.95	160	154	3.75	162	161	0.62	23.17	0.77
LB	156	150	3.85	155	155	0.00	159	153	3.77	159	158	0.63	170	156	8.24	0.86	0.39
LH	60	63	-5.00	58	57	1.72	59	62	-5.08	61	57	6.56	62	60	3.23	0.52	0.06
WH	23	29	-26.09	23	22	4.35	24	23	4.17	21	17	19.05	23	25	-8.70	3.92	0.64
LN	74	78	-5.41	73	70	4.11	79	73	7.59	78	78	0.00	78	75	3.85	0.64	0.25
LS	62	53	14.52	62	54	12.9	62	67	-8.06	66	73	10.61	69	70	-1.45	0.98	0.71
LFA	54	54	0.00	47	56	19.15	49	40	18.37	44	37	15.91	47	51	8.51	1.07	0.48
DSD	72	54	25.00	71	54	23.94	74	62	16.22	74	66	10.81	77	67	12.99	3.58	0.90
LCr	56	50	10.71	54	54	0.00	52	51	1.92	52	54	3.85	53	55	3.77	1.86	-0.45
LF	53	44	16.98	50	42	16.00	51	48	5.88	48	41	14.58	45	42	6.67	0.86	0.53
DCr	52	48	7.69	49	39	20.41	52	53	1.92	52	35	32.69	53	49	7.55	17.70	0.47
LG	51	64	25.49	48	49	2.08	50	50	0	57	62	8.77	54	55	1.85	2.32	0.62
LBu	47	45	4.26	45	41	8.89	48	47	2.08	42	39	7.14	45	47	4.44	0.25	0.79
AS	57	66	15.79	57	86	50.88	56	62	10.71	58	83	43.1	57	65	14.04	230.60	0.66
ACr	14	25	78.57	18	10	44.44	18	23	27.78	17	21	23.53	18	20	11.11	4.17	-0.55

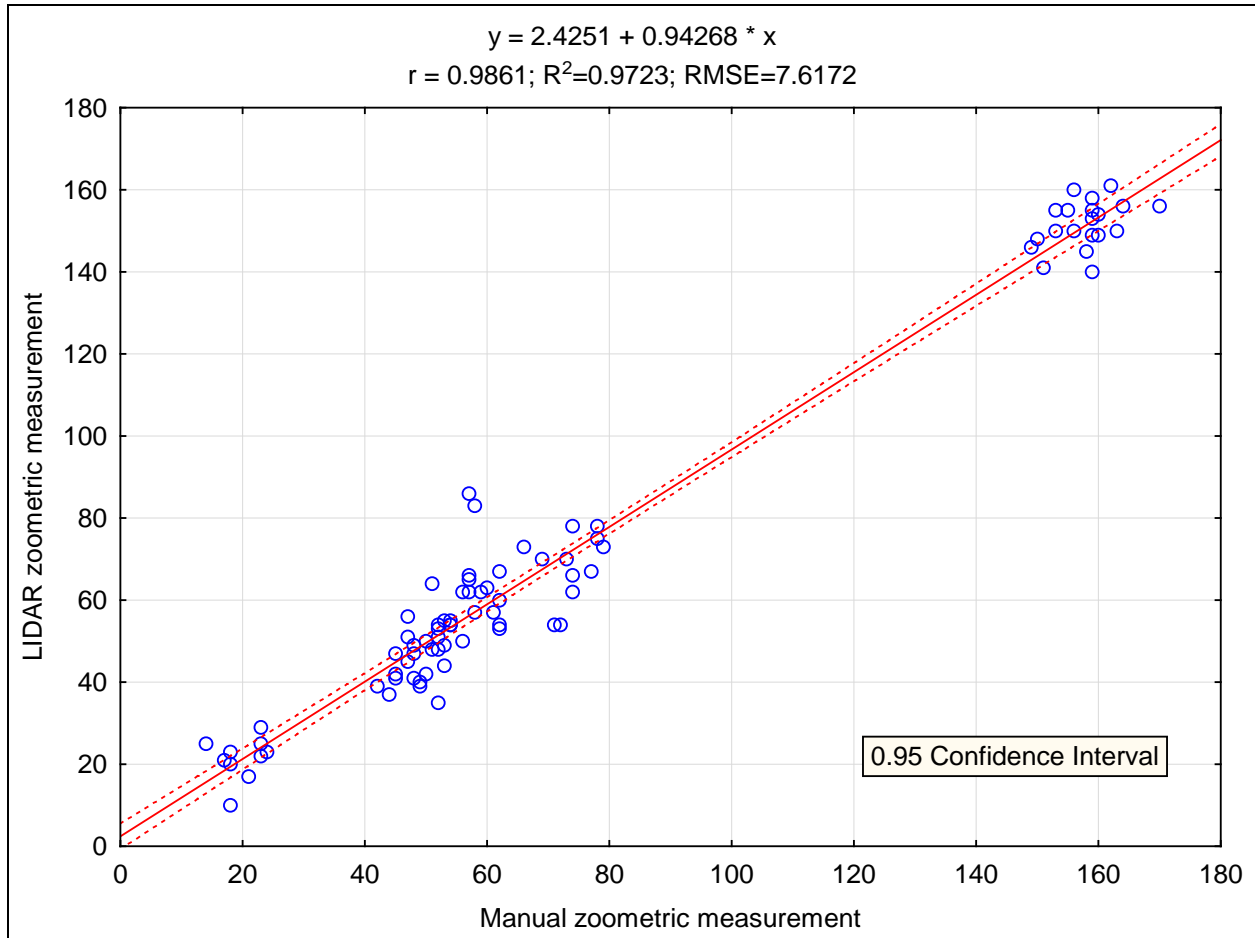
288 Measurements (M); conventional procedure (m) and the LiDAR system (s); relative error (r.e.); Pearson correlations coefficient (r); MHeight at withers¹ (HW);
 289 Height at croup¹ (HCr); Length of head¹ (LH); Width of head¹ (WH); Length of neck¹ (LN); Length of shoulder¹ (LS); Length of forearm¹ (LFA); Length of body¹
 290 (LB); Length of croup¹ (LCr); Length of femur¹ (LF); Length of gaskin¹ (LG); Depth of croup¹ (DCr); Length of buttock¹ (LBu); Dorso-sternum diameter¹ (DSD);
 291 Angle of shoulder² (AS); Angle of croup² (ACr). ¹Centimeters ²Degrees

292 Size and body conformation are critically important traits in nearly all horses breeds, which
293 are presumably subject to a strict process of selection over time (Brooks et al., 2010). The
294 zoometric measurements recorded during this study were similar to those obtained previously in
295 the same breed (Molina et al., 1999; Gómez et al., 2009). Although the bay PRE horses are the
296 tallest in this study, the height of stallion 2 was between that of stallions 1 and 3.

297 The measurements were taken both in mares and stallions to measure the possible
298 difference in accuracy of the measurement at the horse's mane; in mares, the mane is usually
299 cut, and in stallions it is usually left long (Fig. 1 and 4). In our results, stallions had a higher
300 relative error than mares (10.83 and 8.80, respectively) and a higher average v^2 value.

301 Among the three stallions, the relative error ranged between 0 (HC, LFA, LCr and LC) and
302 78.57 (Acr). A total of 7.84% of the zoometric measurements studied had a relative error of 0;
303 for 52.94% had an error > 0 and <10 ; for 21.57% had an error ≥ 10 and <20 ; for 11.76% had an
304 error ≥ 20 and <30 ; and for 5.88% had an error ≥ 30 . For the two mares studied, the relative
305 error ranged between 0 (LN) and 43.10 (AS). A total of 2.94% of the zoometric measurements
306 studied had a relative error of 0; for 64.71% had an error > 0 and <10 ; for 23.53% had an error
307 ≥ 10 and <20 ; for 2.94% had an error ≥ 20 and <30 ; and for 5.88% had an error ≥ 30 (Table 2).
308 The mean relative error was 10.83% and 8.00% without angle measurements. A total of 82.5%
309 of the studied features had an accuracy (v^2) of less than 5%; only one of the angle
310 measurements (AS) had a higher v^2 value. The Pearson correlation between the manual and
311 sensor measurements ranged from 0.90 (DSD) to
312 -0.55 (Acr) or -0.45 (LCr) if the angles were no taken into account (Table 3). The angles are, in
313 general, very difficult to measure also in the traditional system and in the linear morphological
314 assessments since they are influenced by the horse's posture (Sánchez et al., 2013). We
315 suggest that perhaps this problem could be solved with the use of LiDAR or a similar technique,
316 since the scan could be performed when the animal is perfectly poised, and in real life, the
317 horse is always moving to some degree while zoometric measurements are being taken. The R^2

318 and RMSE estimated by comparing the zoometric measurements obtained using the
319 conventional procedure (manual) and the LiDAR system (sensor) were 0.97 and 7.62,
320 respectively (Figure 8).



321
322 **Figure 8.** A scatter plot with regression analysis between the coefficient of determination (R^2) and root-
323 mean-square error (RMSE) comparing zoometric measurements obtained for five Pura Raza Española
324 horses from Estrepe Cartujana performed using the conventional procedure (manual) and the LiDAR
325 system (sensor)

326
327 Thus, our results differ from those reported for different body measures on Mediterranean
328 buffaloes (Negretti et al., 2007), correlating traditional measures with predicted ones using
329 image analysis and obtaining coefficients ranging between 0.91 and 0.99 depending on the
330 measured trait. Moreover, Pallottino et al. (2015) correlated traditional measurements with those
331 obtained with a dual web-camera system ($r = 0.998$) in Lipizzan horses. In any case, It is

332 important to highlight that the Pearson correlation coefficients are not the best way to detect the
333 differences between two methods of measurement, since we are not as interested in whether
334 they increase or decrease in the same direction. The main interest is to ensure that the
335 difference between the two measures is as small as possible. This could be better estimated
336 with the relative error. For the body size measurements of sheep, a low-cost dual web camera
337 was used, yielding a mean relative error of 5% (Menesatti et al., 2014). In addition, 16 adult
338 dairy cows were studied with pairs of pictures, taken with a portable instrument incorporating
339 two synchronized cameras, with photogrammetric measurements having an accuracy within 0.5
340 cm (Gaudioso et al., 2014). The use of a structured light depth camera for three-dimensional
341 body measurements of dairy cows in free-stall barns has also been evaluated (Pezzuolo et al.,
342 2018), obtaining coefficients of determination of $R^2 > 0.84$ and deviations lower than 6% with
343 respect to manual measurements. As in our case, lower performances were obtained for the
344 angle measurements (back slope: $R^2 = 0.12$). In our study, the AS measurement had a very
345 high v^2 value, while ACr had a low v^2 value in both genders. This is the first study in which 17
346 zoometric measurements were analysed, with the majority of the previous studies having fewer
347 traits analysed (ranging from 3 to 8 body measurements (Huang et al., 2018; Pezzuolo et al.,
348 2018)). Taking accurate manual measurement data as validation criteria, LiDAR methodology
349 was used in another study to measure three live cattle, and the experimental results showed
350 that the final deviations were close to 2 mm and within approximately 2% (Huang et al., 2018).
351 The reproducibility of these traits in the linear morphological system ranged from 0.86 to 0.99 in
352 PRE (Sánchez et al., 2013). In studying another group of zoometric measurements, Gaudioso
353 et al. (2014) found a standard deviation of 2.0 cm. It should be noted that we have not been able
354 to study the precision of these 17 zoomometric measurements since replications were not
355 carried out.

356 Once the methodology proposed in this project has been presented and the results
357 obtained from this process have been analysed, the final development of two conclusions can

358 be presented, one from a generic point of view and another from a specific point of view, in
359 addition to a roadmap for future developments.

360 **4. Conclusions**

361 A general methodology has been created that establishes a data evaluation process, so that in
362 the future, applications can be developed in which we can encompass the entire process. The
363 use of a LiDAR system has simplified the taking of measurements, giving rise to a method that
364 can obtain optimal results in monitoring the growth of the livestock. This methodology specifies
365 a clear, precise and logical process.

366 In theory, this method can be extrapolated to any equine breed, as well as to other livestock
367 species. Although it is a pilot study and its accuracy must be improved, it could help in
368 preventing excess spending and stressing animals with the current manual zoometric
369 measurement. More specifically, the following conclusions can be drawn based on the results
370 obtained from implementing the methodology. The approach carried out in this work verifies the
371 viability of the noncontact measurement of large livestock. Knowledge of the scalable growth
372 and animal welfare at the location where the animals are housed will improve the quality of the
373 animals and genetic breeding. However, as can be seen in the results section, due to the
374 constant movement of the horses and the limitation of single-frame analysis, it is not possible to
375 fully analyse the three-dimensional image of the horse. The geometric accuracy of these
376 sensors is closely related to the quality of their external orientation. This external orientation is
377 what we will obtain from the integrated navigation systems (GNSS and INS). The laser scanner
378 measures only the vector oriented from the opening of the laser system to an object point
379 normally on the ground, the horse. Three-dimensional points can only be calculated if at any
380 moment the position and orientation of the LiDAR sensor with respect to a coordinate system is
381 known. A future study will try to improve these drawbacks by integrating two synchronized point

382 clouds through two LiDARs positioned on each side of the animal and performing repeated
383 measurements with the conventional procedure (manual) and the LiDAR system (sensor).

384 For this problem, in the near future this system will continue to be optimized by activating the
385 LiDAR sensor and an integrated navigation system that will allow the sensor to move so that the
386 horse can be rotated while it is being scanned. This will further allow a complete reconstruction
387 of the horse in 3D for greater accuracy in obtaining the zoometric variables.

388

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