Research Highlights

- An analysis of almost 100,000 one-minute precipitation observations recorded by two types of optical disdrometer, Thies LPM and OTT Parsivel², is presented.
- Disdrometer data processing was made by a custom software developed for R environment which overcome binning differences when calculating particle size distribution statistics, allowing for disdrometer type comparison.
- Thies LPM recorded on average double number of particles than OTT Parsivel², with a greater number of small particles resulting in kinetic energy underestimation.
- Differences between disdrometer type increased with precipitation intensity, with Thies LPM recording nine times higher number of particles than OTT Parsivel², influencing all precipitation variables.

Comparison of precipitation measurements by OTT Parsivel² and Thies LPM optical disdrometers

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Abstract

Optical disdrometers are present weather sensors with the ability of providing detailed information of precipitation such as rain intensity, radar reflectivity or kinetic energy, together with discrete information on the particle size and fall velocity distribution (PSVD) of the hydrometeors. Disdrometers constitute a step forward towards a more complete characterization of precipitation, being useful in several research fields and applications. In this article the performance of two extensively used optical disdrometers, the most recent version of OTT Parsivel² disdrometer and Thies Clima Laser Precipitation Monitor (LPM), is evaluated. During two years four collocated optical disdrometers, two Thies Clima LPM and two OTT Parsivel², collected up to 100,000 minutes of data and up to 30,000 minutes with rain in more than 200 rainfall events, with intensities peaking at 277 mm h⁻¹ in one minute. The analysis of these records show significant differences between both disdrometer types for all integrated precipitation parameters, which can be explained by differences in the raw particle size and velocity distri-

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bution (PSVD) estimated by the two sensors. Thies LPM recorded a larger number particles than Parsivel² and a higher proportion of small particles than OTT Parsivel², resulting in higher rain rates and totals and differences in radar reflectivity and kinetic energy. These differences increased greatly with rainfall intensity. Posible causes of these differences, and their practical consequences, are discussed in order to help researchers and users in the election of the sensor, pointing out at the same time limitations to be addressed in future studies.

Keywords: Optical Disdrometer, Particle-size distribution, Precipitation measurement, Instrumental intercomparison, Rainfall kinetic energy

1 1. Introduction

Disdrometers are devices designed to measure the particle size distribution (PSD), or size and velocity distribution (PSVD), of falling hydrometeors. The PSD describes the statistical distribution of falling particle sizes from the number of particles with a given equi-volume diameter per unit volume of air. The PSVD includes also information about the distribution of the particle fall velocities.

Information on the PSD / PSVD is required for a proper understanding 8 of hydrometeorological regimes (Iguchi et al., 2000; Krajewski et al., 2006; 9 Adirosi et al., 2016), soil erosion (Sempere-Torres et al., 1998; Loik et al., 10 2004; Cruse et al., 2006; Petan et al., 2010; Fernández-Raga et al., 2010; 11 Shuttlewort, 2012; Iserloh et al., 2013; Angulo-Martínez and Barros, 2015; 12 Angulo-Martínez et al., 2016) and other applications such as pollution wash 13 off in urban environments (Kathiravelu et al., 2016; Castro et al., 2010) or 14 interactions of rainfall with crop and forest canopies (Frasson and Krajew-15 ski, 2011; Nanko et al., 2004; Nanko et al., 2013). Rainfall estimation by 16 remote sensing, radar and satellite, also rely on PSD information (Olsen et 17 al., 1978; Atlas et al., 1999; Uijlenhoet and Sempere-Torres, 2006; Tapiador 18

et al., 2017). Disdrometer observations of PSD are also used to derive re-19 lationships between radar reflectivity and rainfall rates (known usually as 20 Z-R relationships), despite the difficulties due to differences in altitude of 21 the measurement–surface vs. cloud base–and the sensing area–a few $\rm cm^2$ vs. 22 km²-(Krajewski et al., 1998; Löffler-Mang and Blahak, 2001; Miriousky et 23 al., 2004; Thurai and Bringi, 2008; Marzano et al., 2010; Jaffrain and Berne, 24 2012; Jameson et al., 2015; Raupach and Berne, 2016; Gires et al., 2016). 25 Many of these studies took place within Precipitation Measurement Missions 26 helping the development of better sensors and algorithms for precipitation 27 detection and quantification; some examples are: Ioannidou et al. (2016) for 28 the Tropical Rainfall Measurement Mission (TRMM), Liao et al. (2014) and 29 Tan et al. (2016) for the Global Precipitation Measurement Mission (GPM), 30 Adirosi et al. (2016) for the Hydrological cycle in the Mediterranean Exper-31 iment (HyMex), or Calheiros and Machado (2014) for the Cloud Processes 32 of the Main Precipitation Systems in Brazil (CHUVA) campaign. 33

In addition, bulk precipitation variables can also be calculated from the PSD (sometimes called the 'PSD moments'), including the rain rate, liquid water content, radar reflectivity, rainfall kinetic energy, among others (Ulbrich, 1983; Testud et al., 2001; Jameson and Kostinski, 1998). As such, disdrometers have been incorporated in operational meteorological networks as present weather sensors and pluviometers.

Current commercial disdrometers are based mainly on two physical princi-40 ples to measure the PSD or the PSVD. The first ones are electro-mechanical 41 impact devices recording the electrical pulses produced by the pressure of 42 falling drops when impacting over a surface. Impact disdrometers such as 43 the Joss and Waldvogel disdrometer (JWD, Joss and Waldvogel, 1967) or 44 piezoelectric force transducers (Jayawardena and Rezaur, 2000) were largely 45 used in the 1980s and 90s. The JWD disdrometer gives good results for light 46 to moderate intensity but underestimates the amount of small size drops dur-47 ing heavy rainfall events, and it cannot detect raindrops smaller than 0.2 mm 48

⁴⁹ of diameter (Tokay et al., 2001). Impact based and pressure disdrometers,
⁵⁰ however, rely on theoretical terminal velocity curves to determine the PSD.

More recent disdrometers are based in optical principles (Hauser et al., 51 1984; Löflfer-Mang and Joss, 2000), either from the occlusion of a laser light 52 beam between an emisor and a receptor device produced by the particle 53 passing through; or based on light scattering measurements from particles 54 passing through the light beam. Both types use an emissary and a receiver 55 of the laser signal generally in a horizontal plane, and the emissary can 56 be punctual or an array of emissaries. Commercial examples of the first 57 type are the particle size and velocity disdrometers Parsivel and Parsivel² 58 by OTT Hydromet, and the Laser Precipitation Monitor (LPM) by Thies 59 Clima. An example of the light scattering principle is the light scatter sensor 60 PWS100 (Campbell Scientific Inc.). Optical disdrometers provide full PSVD 61 measures from the unique light beam horizontal plane ($\sim 1 \text{ cm thick}$) by the 62 amplitude and duration obscuration when particles pass through the beam, 63 respectively. Laser disdrometers are not devoid of detection problems related 64 with the effects of uneven power distribution across the laser beam, wind, 65 splashing, multiple drops appearing at the same time (double detections), 66 edge events ('margin-fallers', or partial detections), as reviewed by several 67 studies (Nespor et al., 2000; Habib and Krajewski, 2001; Tokay et al., 2001; 68 Kruger and Krajewski, 2002; Frasson et al., 2011; Raupach and Berne, 2015). 69

An improvement over laser disdrometers is the two-dimensional video 70 disdrometer (2DVD, Joanneum Research). The 2DVD uses two perpendic-71 ular high-speed line-scan cameras, each with an opposing light source, to 72 record particles from orthogonal angles. The 2DVD provides reliable mea-73 sures of particles fall velocity, size and shape (Kruger and Krajewski, 2002; 74 Schönhuber et al., 2008). Currently this disdrometer is considered a reliable 75 reference for particles larger than 0.3 mm (Tokay et al., 2013; Thurai et al., 76 2017), although its use is mostly restricted to experimentation due to its 77 higher cost and data processing requirements. 78

A bibliography search by the key phrase 'optical AND disdrometer' on 79 publications between 2000 and 2017 in Scopus showed that the two models 80 most currently used are OTT Parsivel (mentioned in 50% of a total of 200 81 documents) and Thies LPM (mentioned in 25%). In some disciplines, both 82 disdrometers have been used interchangeably. This is the case, for instance, of 83 soil erosion studies, where Thies LPM was used for monitoring rainfall char-84 acteristics, most notably the kinetic energy, in relation with splash erosion 85 experiments (Angulo-Martínez et al., 2012; Fernández-Raga et al., 2010), and 86 also in the calibration of the European portable rainfall simulator (Iserloh et 87 al., 2013). Parsivel disdrometers, on the other hand, have been used to deter-88 mine the kinetic energy - rainfall intensity relationship (Petan et al., 2010; 89 Sánchez-Moreno et al., 2012). Both disdrometers were used interchangeably 90 in Slovenia to estimate rainfall parameters, including kinetic energy (Petan 91 et al., 2010; Ciaccioni et al., 2016), and to inter-compare solid precipitation 92 observations in the Tibetan Plateau (Zhang et al., 2015). 93

The performance of Parsivel and Thies disdrometers has been compared 94 to other disdrometers such as the 2DVD, the JWD, or by taking a pluviome-95 ter as a reference. Parsivel disdrometers have been evaluated since its first 96 version became commercially available from PM Tech Inc (Sheppard and 97 Joe, 1994; Löffler-Mang and Joss, 2000), with slightly different results de-98 pending on the version of the device analysed (Krajewski et al., 2006; Lanza 90 and Vuerich, 2009; Battaglia et al., 2010; Jaffrain and Berne, 2011; Thurai 100 et al., 2011; Park et al., 2017). In 2005, OTT Hydromet purchased the rights 101 of Parsivel disdrometer and redesigned the instrument. Differences between 102 the PM Tech and the first version of OTT Hydromet Parsivel are described 103 in Tokay et al. (2013), who found important biases in the frequency of small 104 and large drops with respect to a JWD disdrometer. In 2011, OTT Hydromet 105 redesigned the device and presented the Parsivel². This is the current ver-106 sion of the disdrometer, and includes a more homogeneous laser beam and 107 some other modifications that improve its performance (Tokay et al., 2014; 108 Angulo-Martínez and Barros, 2015). The Parsivel² has been compared to 109

other disdrometers. Tokay et al. (2014) compared it with the JWD, and 110 found good agreement in the PSD spectra between both devices for particles 111 sizes larger than 0.5 mm. They also reported systematic underestimation of 112 fall velocities in the Parsivel², for drop diameters of 1.09 mm and higher. 113 Raupach and Berne (2015) and Park et al. (2017) compared the two ver-114 sions of Parsivel with a reference 2DVD, and found that Parsivel², although 115 improving the performance of the first iteration of the disdrometer, still had 116 important biases that resulted in underestimation of small drops and overes-117 timation of large drops, especially during high intensity rains. 118

Thies LPM, on the other hand, became commercially available in 2005 119 from Adolf Thies GmbH & Co. Early analysis of the performance of the Thies 120 disdrometer for detecting different hydrometeors was presented by Bloemink 121 and Lanzinger (2005) at the WMO Technical Conference on Meteorological 122 and Environmental instruments and methods of observations (TECO-2006, 123 Geneva, Switzerland), while an evaluation of its capacity for measuring rain-124 fall intensities and amounts was presented in the same conference one year 125 later (Lanzinger et al., 2006). Since then, this disdrometer has been used 126 worldwide with several firmware updates. Frasson et al. (2011) evaluated 127 the performance of four collocated Thies disdrometers and found that sys-128 tematic biases existed between the devices, and attributed them to miscalcu-129 lation of the disdrometer's sensing area. Lanzinger et al. (2006) found that 130 three LPMs measured higher rainfall amounts than a collocated reference 131 rain gauge, especially during higher intensities, and also reported system-132 atic biases between the three disdrometers. Upton and Brawn (2008) also 133 found discrepancies in the velocity records by three collocated Thies, while 134 the number of particles and their sizes were more consistent. 135

There number of studies inter-comparing Thies and Parsivel disdrometers, however, is very reduced. Brawn and Upton (2008) evaluated the parameters of fitted gamma distributions to the PSD data, and found substantial differences between Thies and Parsivel. Upton and Brawn (2008) found that

Parsivel tended to underestimate the number of small drops (up to three 140 times less for the two lowest size bins) with respect to Thies, while it tended 141 to over-estimate the number of drops larger than 2 mm. They also reported 142 an underestimation of particle fall velocity in comparison with Thies and 143 with the theoretical terminal velocity, especially for midsize drops (1 mm -144 3 mm), and underestimation of total rainfal volume by Parsivel with respect 145 to Thies. These studies were based on the earlier version of the Parsivel dis-146 drometer, and no study up to date has focused on comparing the Thies LPM 147 and the Parsivel². Such a study, however, is highly needed if measurements 148 made with these two disdrometers are to be compared. 149

The objective of this study is to compare the measurements recorded by 150 Thies LPM and OTT Parsivel² optical disdrometers, with the goal of provid-151 ing a quantitative assessment of both sensors and highlighting the associated 152 uncertainties. Measurements of PSVD and integrated rainfall variables as 153 rain rate, kinetic energy, reflectivity and number of drops per volume of air 154 under natural rainfall events are compared, either at the one-minute, the 155 event and the whole season scales. Some technical problems that arise from 156 the different binning of the PSVD matrix by the two devices, which hinder 157 the comparison between their measurements, are dealt with. In the following 158 section a description of the two sensor types and the sampling site is given, 159 together with details of the data processing. Section 3 analyses the results 160 obtained, which are discussed in section 4. Section 5 concludes. 161

¹⁶² 2. Data and Methods

¹⁶³ 2.1. Sampling site and instrumentation

Rainfall characteristics under natural conditions were monitored at Aula Dei Experimental Station (EEAD-CSIC) in the central Ebro valley, NE Spain (41°43'30"N, 0°48'39"W, 230 m.a.s.l.). The experimental site is located in a research farm located on a flat river terrace, classified as having a cold semiarid climate (BSk, Köppen-Geiger). The average annual precipitation was
344.4 mm in the period 1990-2017 (recorded at the Aula Dei meteorological
station which belongs to the network of the Spanish national weather agency,
AEMET) with equinoctial rainfalls (monthly maxima in May, 44 mm, and
October, 39.3 mm; and minima in July, 16.2 mm, and December, 21.7 mm).

Four disdrometers, two Thies Clima LPM and two OTT Parsivel², were 173 operated in continuous record during the period between 17/06/2013 and 174 21/07/2015. Two disdrometers of both types were placed in two masts (Mast-175 1 and Mast-2), which were located 1.5 m apart from each other (Figure 1). 176 Each mast consisted in a pole with two arms 0.5 m apart from each other 177 where two devices, one of each model, were installed. The four sensors were 178 oriented in the same N-S direction. One-minute rainfall PSVD observations 179 were recorded automatically during the period, and rainfall episodes were 180 identified according to the following criteria: a rainfall episode started when 181 rainfall was continuously recorded by at least two disdrometers of different 182 type during at least 10 minutes; and two rainfall episodes were delimited 183 by, at least, one entire hour without rain in at least two disdrometers of 184 different type. Observations corresponding to solid or mixed precipitation 185 were disregarded, as were those with internal error or bad quality flags. 186

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[FIGURE 1: Sampling site with four collocated disdrometers]

Both optical disdrometers, Thies Clima LPM and OTT Parsivel², are based on the same measurement principle. Their external structure is formed by two heads that connect the sheet of laser light through which falling drops are measured. Drop diameter and fall velocity are determined from the obscurations amplitude and duration in the path of an infrared laser beam, between a light emitting diode and a receiver, within a sampling area of approximately 50 cm² (Donnadieu et al. 1969; Löffler-Mang and Joss, 2000).

Raindrops are assumed spherical for sizes less than 1 mm in diameter, and 195 therefore the size parameter is the equivalent diameter for raindrops below 196 this size. For larger raindrops, a correction for oblateness is made, and 197 the size parameter is interpreted as an equi-volume sphere diameter. The 198 laser signal is processed by a proprietary software, and the size (equi-volume 199 particle diameter) and velocity of each particle is determined. The meteor 200 type (drizzle, rain, hail, or snow) is determined based on typical size and 201 velocities, and weather codes (SYNOP and METAR) are generated. A PSVD 202 matrix counting the number $N_{i,j}$ of particles for given size (i) and velocity (j) 203 classes is recorded at desired intervals, usually one minute. Several integrated 204 variables are also computed and stored at the same intervals. These include 205 the number of particles detected (NP, \min^{-1}) , the particle density (ND,206 $m^{-3} mm^{-1}$), the rainfall amount (P, mm) and intensity (R, mm h^{-1}), the 207 radar reflectivity (Z, dB mm⁶ m⁻³), visibility (m) and kinetic energy (J m⁻²) 208 mm-1). 209

This operational principle in subject to a number of potential sources of 210 bias, as reviewed by Frasson et al. (2011). One of such sources of bias is the 211 uneven power distribution across the laser beam, or variations of this power 212 with time. Also, the geometry of the laser beam limits the estimation of 213 fall velocity to the vertical component, producing biased measures when the 214 particles fall with a different trajectory or angle due to wind or eddy drag 215 (Salles and Poesen, 1999). Other source of biased measurements is due to the 216 ocurrence of coincident particles, which are perceived as just one single drop 217 by the sensor. Similarly, the event of one drop falling at the edge of the laser 218 beam ('margin faller'), therefore being only partially observed, leads to biased 219 measurements. Both sensors mention in their technical data some correction 220 for edge-detection (margin fallers) and coincident particles, although there is 221 little information on how these two events are identified and treated. More 222 details of both instruments and the measurement technique, along with the 223 assumptions used to determine the size and velocity of hydrometeors, can 224 be found in Löffler-Mang and Joss (2000), Battaglia et al. (2010), Tapiador 225

et al. (2010), Frasson, et al. (2011) Jaffrain and Berne (2011), Tokay et al. (2013 and 2014), Raupach and Berne (2015), and in their respective technical manuals.

There are slight hardware variations between the two instruments, as well as differences in how the raw data are treated and converted into the outputted variables. Since these differences may have a impact on the final records, we review the relevant characteristics of each device in the following paragraphs.

234 Thies Clima Laser Precipitation Monitor

The Laser Precipitation Monitor (LPM) uses a 780 nm laser beam which 235 is 228 mm long, 20 mm wide, and 0.75 mm thick on average, resulting in 236 a sampling area of 45.6 cm^2 . Geometric deviations from this standard are 237 reported by the manufacturer for each particular disdrometer, and for in-238 stance the sampling areas of the two devices used on the experiment were 230 46.65314 and 49.04051 cm². It records particles starting from 0.16 mm of 240 diameter, and precipitation starting from 0.005 mm h^{-1} . The Thies tech-241 nical documentation indicates that that the size and velocity measurements 242 are 'checked for plausibility' to prevent issues such as edge events, implying 243 that some particles are filtered out, although the details of this procedure 244 are not specified. From the raw particle data several bulk variables ('PSVD 245 moments') are integrated internally by the device's firmware. Drop diame-246 ters and velocities are then grouped into 22 and 20 classes ranging between 247 0.125 mm up to 9 mm and 0 m s⁻¹ up to 12 m s⁻¹, respectively (see Table 248 6), and the number of particles recorded at each size and velocity pair bin 249 is stored. The bulk variables computed by the Thies LPM does not include 250 the kinetic energy. In addition, several status flags are provided in the data 251 telegrams informing about voltage oscillations, sensor temperature, and an 252 evaluation of the measurement quality. 253

[TABLE 1]

255 OTT Parsivel² disdrometer

The Parsivel disdrometers used in this study belong to the second gener-256 ation manufactured by OTT Hydromet Inc (Parsivel²). The Parsivel² uses a 257 780 nm laser beam which is 180 mm long, 30 mm wide, and 1 mm thick on 258 average, with no indication about deviations from these values from the man-259 ufacturer. The sampling area for the two Parsivel disdrometers was therefore 260 54 cm^2 . The Parsivel² records particles starting from 0.2 mm of diameter, 261 and precipitation starting from 0.001 mm h^{-1} . The measured particles are 262 stored in drop diameter and fall velocity bins in a 32 x 32 matrix with uneven 263 intervals starting at 0 mm diameter up to 26 mm and from 0 m s⁻¹ up to 264 22.4 m s^{-1} (Table 6). The first two size categories, which correspond to sizes 265 of less than 0.25 mm, are left empty by the manufacturer because of the low 266 signal-to-noise ratio. The Parsivel², similarly to the Thies, also provides a 267 sensor status flag and several control variables in its data telegram. 268

According to Battaglia et al. (2010), particles up to 1 mm are assumed 269 spherical, and between 1 and 5 mm they are assumed as horizontally-oriented 270 oblate spheroids with axial ratio linearly varying from 1 to 0.7, with this ratio 271 being kept constant at 0.7 for larger sizes. The Parsivel technical documen-272 tation mentions that the device filters out edge events, although the exact 273 details of this procedure are not given. Battaglia et al. (2010) mention that 274 the newest Parsivel units include two extra photo-diodes at the edge of the 275 laser beam to detect and remove the edge events, but the manufacturer pro-276 vides no information about this. Independently to filtering our edge events, 277 Löffler-Mang and Joss (2000) indicate that a correction of the effective sam-278 pling area is used depending on the particle size. Some sources (Tokay et 279 al., 2013) also refer that a correction to the fall velocity is applied to drop 280 sizes between 1 and 5 mm, although once again there is not more information 281 on this correction. Parsivel² disdrometers external structure differs from the 282

Thies LPM in incorporating a splash protection shield above the laser heads, which aims at minimising the effect of splashed drops that interfere with a high velocity with the laser beam and result in biased measurements.

286 2.2. Processing disdrometer data

One minute disdrometer data telegrams were stored in an industrial 287 miniature PC (Matrix 504 Artila Inc). The PC included custom software 288 to collect, pre-process and send data telegrams to a central server. Time 289 synchronisation was performed once per day using the Network Time Proto-290 col (NTP), allowing bias correction of the internal disdrometer clocks that 291 have a tendency to drift. Direct reading of the data telegrams generated 292 by the disdrometers resulted in one-minute time series of the variables of 293 interest for this study: PSVD matrices $(N_{i,j})$, bulk variables (P, R, NP, ND,294 Z, E), SYNOP codes, and status and error flags. An exception were Thies 295 disdrometers, which do not compute the kinetic energy, E. Parsivel, on the 296 other hand, gives the kinetic energy expressed in J, so it was divided by the 297 sampling area and the rainfall amount to obtain E. 298

In addition to the bulk variables computed by the internal software of the devices, the bulk variables were computed again from the PSVD matrices, using the following expressions:

$$P = \frac{4}{3}\pi \sum_{i,j} \left(\frac{1}{A_i} N_{i,j} \left(\frac{D_i}{2} \right)^3 \right)$$
(1)

$$R = \frac{P}{\Delta t} \tag{2}$$

$$NP = \sum_{i,j} N_{i,j} \tag{3}$$

$$ND = \frac{1}{R\Delta t} \sum_{i,j} \left(\frac{1}{A_i} \frac{N_{i,j}}{V_j} \right)$$
(4)

$$Z = \log\left(\frac{1}{\Delta t} \sum_{i,j} \left(\frac{1}{A_i} N_{i,j} \frac{D_i^6}{V_j}\right)\right)$$
(5)

$$E = \frac{4}{3} \pi \frac{\rho}{P} \sum_{i,j} \left(\frac{1}{A_i} N_{i,j} \left(\frac{D_i}{2} \right)^3 \frac{V_j^2}{2} \right)$$
(6)

where ρ is the density of water (1000 kg m⁻³), D_i is the mean diameter of class *i*, V_j is the mean velocity of velocity class *j*, and Δt is the sampling frequency (s). The effective sampling area, A_i (m⁻²) depends on the particle size, since in order to be correctly sensed the particles need to be inside the light beam in its entirety, so:

$$A_i = A\left(1 - \frac{D_i}{2w}\right) \tag{7}$$

where A is the sampling area of the disdrometer and w is the width of the laser beam. As it can be seen, the effective sampling area gets reduced as the drop size increases, and the magnitude of the correction applied is inversely proportional to w.

This allowed, on one hand, obtaining E for Thies disdrometers, but also 311 permitted to apply a number of corrections that simplified the comparison 312 between the two types of disdrometer. Thus, we ignored the particle counts 313 in the first size bin of Thies disdrometers and the counts in the size bins 314 larger than 8 mm, so the two disdrometer types were measuring the same 315 range of drop sizes (0.25 to 8 mm). We also applied a filter to remove highly 316 unlikely drop size and velocity combinations, as done in many studies (e.g., 317 Tokay et al., 2001; Jaffrain and Berne, 2011; Tokay et al., 2013; Raupach et 318

al., 2015). In order to do that, each size and velocity bin was compared with
the terminal fall velocity model of Beard (1976), and the bins for which a
difference larger than 50% existed with the theoretical model were excluded.

In order to compare PSVD data between disdrometer types, the 10th, 322 50th and 90th percentiles of the particle size (D10, D50, D90) and velocity 323 (V10, V50, V90) were computed (Table 2). One problem that arises when 324 percentiles are computed from binned data is that the resulting percentiles 325 may be biased depending on the binning structure of the data. If all the 326 particles recorded in one bin are assigned the mean value of the bin (the 327 easiest option), different bin configurations will lead to different computed 328 percentiles, even if the raw data before binning were the identical. When 329 data from different binning structures are compared, as it is the case here 330 between Thies and Parsivel disdrometers, an interpolation scheme needs to 331 be used for distributing the range of values within each bin across all the 332 particles corresponding to that bin. Here we used a random distribution 333 over the range of values in the bin following a linear probability distribution 334 constructed by fitting a line between two points determined as the average 335 of the number of particles in the bin and the corresponding values on the 336 neighbouring bins. Given the high number of particles detected, the random 337 component of this scheme has a negligible effect on the results. Once all 338 the number of particles by minute were assigned particle size and velocity 339 values, the percentiles were calculated, allowing for a comparison between 340 disdrometers. 341

In addition to one-minute data, the mean (m) and maximum (M) values of some of these variables (Rm, RM, KEm, KEM, Em, EM, NPm) were computed for each rainfall event. A summary of the variables analysed is provided on Table 2.

[TABLE 2]

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All data processing, including reading the raw telegrams, computing the integrated variables (erosivity for Thies LPM and size and velocity percentiles), and plotting, was performed using a custom package for the R environment, disdRo (Beguería et al., 2017).

251 2.3. Comparison of disdrometer measurements

Prior to any analysis, minute observations with low-quality or bad sensor 352 status flags were removed from the comparison dataset. Minutes with miss-353 ing data, precipitation below 0.1 mm h^{-1} or less than 10 particles detected 354 in any of the four disdrometers were also removed. This way, only minutes 355 with good quality data in the four devices were considered in the analysis. 356 The comparison was made primarily on the bulk variables computed from 357 the PSVD matrix stored in the one-minute telegrams outputted by the four 358 disdrometers, by applying equations 1 to 6. We also compared the bulk vari-359 ables calculated by the internal firmware of the devices, in order to check the 360 impact of the effective sampling area correction and the removal of unlikely 361 size-velocity bins. 362

Kernel density and violin plots, i.e. non-parametric graphical estima-363 tions of the probability density functions of the variables, were used as a 364 preliminary analysis tool. A formal comparison between the two disdrom-365 eter types was performed using a Gamma generalised linear mixed model 366 (Gamma GLMM), with the bulk variables listed in Table 2 as response 367 variables. Mixed models allow incorporating both fixed-effects and random-368 effects in the regression analysis (Pinheiro and Bates, 2000). The fixed-effects 369 describe the values of the response variable in terms of explanatory variables 370 that are considered to be non-random, whereas random-effects are treated 371 as arising from random causes, such as those associated with individual ex-372 perimental units sampled from the population. Hence, mixed models are 373 particularly suited to experimental settings where measurements are made 374 on groups of related, and possibly nested, experimental units. If the group-375

ing factor was ignored when modelling grouped data, the random (group) 376 effects would be incorporated to the error term, leading to an inflated esti-377 mate of within-group variability. This allowed us to assess for differences in 378 the response variables as a function of the disdrometer type (fixed factor), 379 while controlling for possible differences due to the location of the two masts 380 (random factor). Since the explanatory variable is a dichotomic variable 381 (the disdrometer type), this configuration is equivalent to a random-effects 382 Analysis of Variance (ANOVA). A Gamma distribution was used to model 383 the response variables, since this distribution is best suited to positive data 384 with variance increasing with the mean, as it is the case of the disdrometric 385 variables analysed here. This model configuration can be described as: 386

$$y_{i} \sim \text{Gamma}(\theta_{i}, \nu)$$

$$\theta_{i} = \nu/\mu_{i}$$

$$g(\mu_{i}) = \mu + \beta_{t(i)} + \alpha_{m(i)} + \epsilon$$

$$\beta_{t(j)} \sim \mathcal{N}(0, \sigma_{\beta}^{2})$$

$$\alpha_{m(i)} \sim \mathcal{N}(0, \sigma_{m(i)}^{2})$$

$$\epsilon \sim \mathcal{N}(0, \sigma^{2})$$

(8)

where y_i is the *i*th observation of the response variable Y; ν is a shape 387 parameter; θ_i is a scale parameter, which can be expressed in terms of ν 388 and a mean value corresponding to the *i*th observation μ_i ; μ is a global 389 mean; $\beta_{t(i)}$ is a parameter accounting for the effect of the disdrometer type 390 corresponding to observation i, t(i); and $\alpha_{m(i)}$ is a parameter accounting for 391 the location (mast) corresponding to observation i, m(i). In our case, we 392 counted with four disdrometers grouped into t(i) = (T, P) disdrometer types 393 (Thies and Parsivel, respectively), and located in m(j) = (1, 2) masts, and 394 we set $\beta_1 = \alpha_1 = 0$. For the link function $g(\mu_i)$ we used an identity link, 395 $g(\mu_i) = \mu_i$, except for R, Z, E and NP for which a log link, $g(\mu_i) = \log \mu_i$, 396 was used. 397

The model in eq. (8) was fitted by generalized least squares (GLS), using the function lme from the R library lme4 (Pinheiro and Bates, 2011). A random sample of N=1000 records, corresponding to 250 minutes, was used in the analysis, in order to avoid size effects affecting negatively the statistical significance tests (Type I error inflation; see, e.g., Lin et al., 2013).

403 **3. Results**

A summary report on the data acquired by the four disdrometers is re-404 ported on Table 3. Almost 100,000 minutes of data were obtained from 405 each device. Missing values due to technical issues (power supply failures 406 and device hangouts, data communication problems) were found in all dis-407 drometers, although they were more prevalent on one of the Parsivels (P2), 408 resulting in a significantly lower number of records by this device. The num-409 ber of errors, as reported by the status flags of the devices, was low, albeit 410 larger in Parsivel than in Thies devices. Some records were discarded due to 411 quality issues, either based on the quality flat reported by Thies (only data 412 with quality flags above 90% were accepted), or on non-consistent data in the 413 telegram (saturation of the PSVD bins or excessively large intensity values) 414 in the Parsivels. Since Parsivel does not report the data quality, no quality 415 threshold could be used. Around 31% of the minutes recorded rain hydrome-416 tors in both Thies, while this percentage was lower for Parsivel (27.5%) in 417 P1; the value of P2 was even lower, but can not be considered since this 418 device recorded a significantly reduced number of minutes due to technical 419 issues). The larger amount of minutes with rainfall in Thies disdrometers 420 can be attributed to their highest sensitivity, since they are able to records 421 smaller raindrops (more on this later). 422

All types of precipitation events occurring in the sampling site were represented, with the majority of observations corresponding with autumn rains, as corresponds to the climatology of the area. Rain rates varied between $_{426}$ 0.014 mm h⁻¹ and 277 mm h⁻¹. The minimum precipitation rates were between 0.014 and 0.020 mm h⁻¹, with no differences between devices. The absolute maximum precipitation rates measured during the experiment depended on the disdrometer type, with Thies being the ones recording the highest values.

As mentioned in section 2.1, only the common minutes were selected from the complete dataset, defined as those having high quality data and detection of rainfall particles in each of the four disdrometers. This led to a total of 434 46,636 records, corresponding to 11,659 minutes belonging to 157 rainfall 435 episodes.

436

[TABLE 3]

When considering only the records for which data of the four disdrometers 437 existed, the total accumulated precipitation as measured by the disdrome-438 ters internal software was 244.9 mm (T1), 254.5 mm (T2), 220.4 mm (P1), 439 and 228.1 mm (P2). This values were slightly different to those calculated 440 from the PSVD data, which were slightly lower at 240.1 mm (T1), 253.0 mm 441 (T2), 218.6 mm (P1), and 222.0 mm (P2). A graphical comparison of the 442 cumulative time series for the computed and internal precipitation is pro-443 vided in Figure 2. Some discrepancies in total precipitation were therefore 444 found between the devices, with the two Thies LPM devices recording more 445 precipitation than the Parsivel ones. Between locations, mast 2 tended to 446 record larger precipitation in both devices, although the magnitude of this 447 difference was much lower than the difference between disdrometer types. 448

⁴⁴⁹ Differences were also found with respect to cumulative kinetic energy, for ⁴⁵⁰ which larger values were also found for Thies (2100 and 2101 J m⁻² mm-1) ⁴⁵¹ than for Parsivel (1749 and 1829). This corresponds to values obtained from ⁴⁵² the PSVD data, since Thies disdrometers do not calculate the kinetic energy ⁴⁵³ internally. Unlike with P, for E there were important differences between the values measured by the Parsivel² disdrometers (2100 and 2181) and those calculated from the PSVD, reported above.

This result suggests that differences between devices could be done, to a certain extent at least, to Thies LPM devices being more sensitive in the lower range of the PSVD spectrum, although this hypothesis requires further analysis, as done in the following sections.

460 3.1. Example events

475

Two events, representative of low and high precipitation intensity rates, 461 were selected to illustrate the differences between disdrometer outputs. Time 462 series of some bulk variables are shown in Figures 3 and 4. In both events, 463 Thies devices consistently reported higher rainfall intensitity and cumulative 464 precipitation. This is related to a larger number of rain particles detected, as 465 shown by the number density (which factors out the different rain intensities). 466 There were differences, too, in the median particle size, which was much 467 larger in the Parsivel devices. Interestingly, it seems that these differences 468 (larger number of drops in Thies, but larger mean size in Parsivel) somehow 469 cancelled out for radar reflectivity and kinetic energy, which depend both on 470 the number of drops, their size and velocity. 471

These differences were most evident in the high intensity event, and were also higher if no corrections for unlikely drops and effective sampling area were performed (Supplementary material, Figures A.1 and A.2).

[FIGURES 3 and 4]

The PSVD plots (Figures 5 and 6), depicting the number of drops detected for each combination of drop size and velocity classes during the event by each disdrometer, help explain the differences found. A first and evident

difference is that Thies disdrometers had a much wider distribution of the 479 PSVD spectra than Parsivel ones. The terminal velocity of raindrops as a 480 function of their size according to Beard (1978), also depicted in the figure, 481 was used to filter out unlikely combinations of size and velocity. Combina-482 tions which differ by more than 50% with the theoretical fall velocity are 483 represented in the figure with a 50% transparency. Although a majority of 484 particles were found to lie in a region close to the theoretical line, Thies de-485 vices had a much larger number of particles far from the theoretical model, 486 both in the high and low intensity events. Particularly, a large number of very 487 small particles at much higher velocities than expected was very prominent, 488 as were the drops with a large diameter but a fairly low velocity. Typically, 489 the first case (small, fast raindrops) are attributed to edge events (partial 490 recognition or larger drops falling in the edge of the laser beam), or splashed 491 particles, while the second case are interpreted as double detections (two or 492 more simultaneous drops). Both effects tend to increase with the precipita-493 tion intensity, as the anomalous events become more frequent. 494

The frequency of anomalous raindrops was much lower in the Parsivel 495 output, for which the vast majority of cases fell within the theoretical model 496 limits. This can be attributed to a number of facts. From pure geometrical 497 considerations, a larger prevalence of edge events can be expected from Thies, 498 since its laser beam has a reduced width (20 mm) with respect to Parsivel 490 (30 mm), so the proportion of edge events with respect to the number of 500 particles detected is higher. Other reasons such as a larger proneness to 501 splashing or differences in the internal processing of the data (that, as stated 502 by the manufacturers, includes some filtering of anomalous data), may also 503 help explain this differences. 504

⁵⁰⁵ Finally, and interestingly, an underestimation of drop velocities with re-⁵⁰⁶ spect to the theoretical model could be found in Parsivel devices, most no-⁵⁰⁷ tably in the high intensity event and for particles larger than 1 mm. A formal analysis of these differences, considering the whole data set, is presented in the following section.

511 3.2. Integrated variables, minute scale

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521

When the whole dataset was analysed, differences between disdrometers 512 were also evident, as shown by the exploratory kernel density plots (Figure 513 7). This was further confirmed by the Gamma GLMM analysis (Table 4). 514 The coefficients reported in the Table for the fixed effects correspond to β_T 515 and β_P when μ is set to zero in equation 8, and can be interpreted as the 516 mean values of the response variables for each disdrometer type, when other 517 factors (the mast, in this case) are accounted for. The table includes also the 518 p-values corresponding to these coefficients, as well as the residual and mast 519 standard deviation (σ and $\sigma_{m(i)}$, respectively). 520

[FIGURE 7 and TABLE 4]

The analysis yielded significant differences between disdrometer types for 522 all the response variables analysed, while the random effect (the mast) had 523 a negligible effect as shown by its small variance with respect to the random 524 error (residual). There were substantial differences in the number of particles 525 detected, NP, and in the PSVD percentiles. Thus, Thies disdrometers had 526 a lower coefficient for NP (230 vs 194), indicating a tendency to detect a 527 higher number of particles (an increase of circa 20%). This also had much 528 lower coefficients for D10 and D50 (0.59 vs 0.74 for the median drop size, 520 i.e. a decrease of circa 20%), as well as for V10 and V50 (2.4 vs 2.9, i.e. an 530 18% difference). The magnitude of the difference was lower for the highest 531 percentiles (D90 and V90), where Thies even had a higher coefficient for 532 velocity, indicating a larger spread of velocities compared to Parsivel. 533

These differences in the number of particles and in the PSVD were translated to the bulk variables, which also showed significant differences in all cases. The magnitude of the effect, i.e. the mean differences between the two disdrometer types, were high for the particle density (21,600 vs 15,920, a 36% increment) and kinetic energy (11.09 vs 9.66, i.e. a 15% difference), while they were smaller (albeit significant) for R and Z (12% and 7% difference, respectively).

The differences found in the PSVD percentiles allows for a better un-541 derstanding of the differences in the integrated variables, since the particle 542 size and velocity have contrasting effects on R, ND, Z, and E. In general, 543 a higher number of particles implies increasing values of all these variables, 544 which favours Thies devices since it tended to detect a higher number of par-545 ticles. The particle size, on the other hand, has a similar effect of increasing 546 all the variables for which it is relevant (R, Z and E). Since the particle 547 size was in general higher in Parsivel devices, this effect partially cancels out 548 the effect of the increasing number of particles. Particle velocity, which was 549 in general higher in Parsivel (except for the largest drops), has a positive 550 effect in E, but a negative effect on Z, which further explains the differences 551 found. The particle density (ND), finally, is not affected by the drop size and 552 is negatively affected by fall velocity, and that the reason why this variable 553 showed the highest difference between both disdrometers. 554

555 3.3. Integrated variables, event scale

Although one of the benefits of the optical disdrometers is their ability to provide large amounts of information at very fine temporal scales (as oneminute data analysed here), very frequently these data data are aggregated over larger time periods or rainfall events for practical issues. For instance, it is typical the computation of kinetic energy totals for rainfall events, for instance for soil erosion applications. When considering the same variables at the event level, looking at the mean and maximum values over the event, ⁵⁶³ similar results were found (Figure 8 and Table 5).

[FIGURE 8 and TABLE 5]

Again, significant fixed effects were found for all response variables, while the random effect was marginal in all cases. The average number of particles during the events was much larger for Thies, and the median drop size and velocity was lower. There were also differences, although of smaller size, in the rest of integrated variables.

570 3.4. Effect of PSVD data correction

564

The effect that the data correction scheme may have on the integrated 571 variables merits some analysis, since it modifies the PSVD distribution. Here 572 we applied a filter that consisted on eliminated the unlikely drops, which was 573 aimed at eliminating edge events and double detections, while a correction 574 for the sensing area as a function of the drop size was applied to compensate 575 the loss of mass. The results showed in the previous sections were all based 576 on the corrected data, but in order to determine the effect of this correction 577 on the computed variables, the analysis was replicated without applying the 578 filtering and the correction. 579

The results are shown in the Supplementary material, in Table A.1 and 580 Figure 7. A comparison with the results shown in the previous section re-581 veals the same general pattern, but with stronger effects. For instance, the 582 coefficient for the number of particles NP was 62% higher in Thies than in 583 Parsivel. Interestingly, the effect of the correction on the particle size per-584 centiles had a different sign on Thies, for which D50 increased from 0.53 585 (without correction) to 0.60 (with correction), while on Parsivel it decreased 586 from 0.80 to 0.74. For the median particle velocity (V50), the coefficient re-587 mained very similar before and after correction for Thies, while for Parsivel 588

it increased from 2.88 to 3.09 (7%). The relative magnitude of the differences between Thies and Parsivel disdrometers was 88% for ND, 12% for R, 15% for E and 7% for Z, i.e. much higher than after filtering and correction for ND but similar for the other three variables.

⁵⁹³ 3.5. Effect of rainfall intensity

Data were divided by intensity ranges in order to test if the effect of 594 the disdrometer type changed with different rain intensities. As the rainfall 595 intensity increases, it is expected to find more and bigger drops, which may 596 in turn modify the differences found between disdrometer types. Data were 597 thus divided in three intensity groups: low intensity (from 0.1 mm h^{-1} up to 598 2 mm h^{-1}), medium intensity (from 2 mm h^{-1} up to 10 mm h^{-1}) and high 590 intensity (more than 10 mm h^{-1}). Model coefficients for the three intensity 600 ranges are given in Table 6, and kernel density plots can be found in the 601 Supplementary material (Figures A.4, A.5 and A.6). 602

603

[TABLE 6]

The same effects described above were found at different rainfall inten-604 sities. The magnitude of the effects, however, tended to increase with the 605 intensity. Thus, the relative difference between the coefficients of NP ranged 606 between 7% (146 vs 136) for low rainfall intensity, 27% for medium intensity 607 and 65 % for high intensity, while the median particle size ranged between 608 16%, 28% and 200%. Equally large were the relative differences between the 609 coefficients of ND, which varied between 33%, 67% and up to 292%, while for 610 the remaining variables the increase of the effect with the rainfall intensity 611 was less pronounced. 612

613 4. Discussion

Optical disdrometers are commercially affordable sensors able to provide a thorough description of precipitation, and they are being increasingly used by national weather services as present weather sensors and even rain gauges requiring low maintenance. Besides their use in operational networks, optical disdrometers provide information on precipitation drop spectra that has applications in different fields, and they are being increasingly used in research.

Thies Clima LPM and OTT Parsivel² are among the most common, state-620 of-the-art, optical disdrometers. Despite being based on the same functioning 621 principle and having similar characteristics in terms of sensibility and range 622 of particle detection, there are substantial differences between them that 623 may affect differently their records. We have stressed the differences in the 624 higher and (more important) lower particle size detection ranges of the two 625 devices, with Thies having a lower detection threshold that may induce bias 626 in the records of the two disdrometer types. Filtering the PSVD matrix 627 to a common detection range, as done here, allows for a fair comparison 628 between the outputs of the disdrometers, and should be recommended for 629 any study that aims at presenting general results. However, as we have seen 630 here, despite applying the same detection thresholds to the data outputted 631 by the two disdrometers, significant differences were found both at the level 632 of PSVD spectra (particle size and velocity percentiles) and on the bulk 633 variables (PSVD moments). 634

There are a number of factors that may help explain the differences found. Geometrical differences between the laser beams are highly relevant, since they greatly influence the probability of bias-inducing effects such as edge events ('margin fallers') and double detections. A larger sampling area, for instance, implies a higher chance of double detections. At this respect, the larger sampling area of Parsivel (54 cm²) over Thies devices (45.6 cm² on average) implies that Parsivel disdrometer should be more affected by double

detections. Double detections, i.e. time-overlapping drops, may be sensed 642 just as one single drop (hence causing a loss of mass which may translate 643 to a reduced precipitation record); or as a much larger drop at an unusu-644 ally low velocity. Since these unusual particles are often discarded from the 645 PSVD matrix, this may result in another source of mass loss, which may or 646 not be partially solved by the sampling area correction (more on this later). 647 Although this would require further research, for instance with the help of 648 numerical simulations as in the work by Raasch and Umhauer (1984), we 649 suspect that the tendency towards a lower number of particles detected and 650 lower precipitation amounts found on Parsivel devices may have a relation-651 ship with this effect. 652

But geometrical effects are not restricted to this. Since the effective sam-653 pling area of optical disdrometers depends on the particle size, not only the 654 total area but also the width of the laser beam plays an important role as a 655 source of bias. In particular, the proportion of edge events (i.e. particles that 656 are sensed only partially due to falling at the edge of the laser beam) over 657 the total number of particle detections of the same diameter class is inversely 658 proportional to the width of the beam. The smaller width of the laser beam 659 on Thies (20 mm) over Parsivel (30 mm) plays against the former, which 660 should be more prone to be affected by edge events. This becomes more 661 relevant for the higher particle bins. For 5 mm particles, for instance, the ef-662 fective with gets reduced to 15 mm for Thies, i.e. a reduction of 25%, while 663 for Parsivel this reduction amounts to 16.6%. Edge events result in partially 664 sensed particles, implying a mass loss and an over-estimation of fall velocity. 665 The high prevalence of over-accelerated, small particles in the PSVD spectra 666 of Thies disdrometers may be related to this effect, although again further 667 analysis is required in order to confirm this hypothesis. At this respect, the 668 Thies manufacturer checks and reports on each device the deviations due to 669 fabrication tolerances from the theoretical geometrical properties of the laser 670 beam, whereas this information is not given for Parsivel. 671

In order to overcome this problems, we applied a correction scheme which 672 is similar to the ones found in other studies (e.g. Löffler-Mang and Joss, 2000; 673 Battaglia et al., 2010; Raupach and Berne, 2015). The scheme consists on 674 two parts: the first implies removing highly unlikely particle counts, i.e. those 675 with velocities that are far from the theoretical fall velocity corresponding to 676 their size. These unlikely particles are most possibly caused by edge events 677 and double detections, so they are removed from the PSVD data. This causes 678 a loss of mass, and this loss of mass is uneven since it increases with the par-679 ticle size (due to the geometric effect explained above), so the second part of 680 the scheme consists on correcting the effective sampling area used in calculat-681 ing the bulk variable from the PSVD (equation 7). The correction, however, 682 is not guaranteed to restitute all the mass loss, and careful calibration is 683 required in order to match the filtering of unlikely particles (which depends 684 on the threshold used for particle removal) with the effective area correction. 685 Here we used a threshold corresponding with a difference higher than 50%686 with respect to the theoretical fall velocity matched to a factor or 1/2 of the 687 drop diameter for the area correction, but other combinations are possible. 688 Again, numerical simulation should help in determining the best correction 689 parameters, which in turn should consider the different beam geometries. 690

Our results showed differences between the two disdrometer types, which 691 were not totally removed by the correction scheme (although they were par-692 tially diminished with respected to the un-corrected records). Differences in 693 the in the internal treatment of the data by the two devices, which is not pub-694 lic, may also help explain this differences. Both manufacturers indicate that 695 some treatment of unlikely detections is performed internally, but very little 696 detail is given. From the examination of the raw PSVD matrices, it seems 697 that the correction applied by Thies, if any, is very subtle, while the output 698 of Parsivel seems to be much more affected by corrections. The technical lit-699 erature, also, gives more detail in the case of the Parsivel, for which at least 700 a correction for the effective sampling area is reported (Löffler-Mang and 701 Joss, 2000). The exact nature of these corrections, however, is not known, or 702

even if they are applied to the integrated variables only, or also to the PSVD
data. This uncertainty makes it difficult to implement an effective correction
scheme that makes the outputs of the two disdrometer comparable.

The external structure of the devices also plays an important role and 706 may lead to incorrect drop detections due to turbulence (see, for instance, 707 Constantinescu et al. 2007, for a review of turbulence induced errors on 708 pluviometers) and splashing (particles intercepted by the enclosure of the 709 devices which break and splash away in smaller but accelerated drops, see 710 Kathiravelu et al., 2016). It seems that the Thies disdrometer is more prone 711 to having splashed drops interfering with the laser beam, since it contains 712 larger flat surfaces susceptible of splashing particles in the direction of the 713 sensor. The Parsivel units, on the other hand, do not have flat surfaces 714 and include a splash protection shield that seems to effectively reduce the 715 risk of splashing. These morphological differences may also affect differently 716 in case of wind, since the turbulences generated may be very different on 717 both devices, and may also be a cause of systematic bias between the two 718 disdrometers. A future study using high speed video and a wind-tunnel setup 719 could help examine the occurrence and magnitude of these effects, which are 720 poorly quantified up to now. 721

Finally, we also detected a tendency towards underestimating the velocity 722 of falling particles in the case of the Parsivel units, especially in the range 723 between 1 and 3 mm. This have been shown previously, and according to 724 Tokay et al. (2014) this issue is known to the Parsivel manufacturer who 725 mentioned that it is in process of being fixed. However, at least the units 726 tested, still suffered from the same problem. Underestimation of the fall 727 velocity may have a substantial influence on the bulk variables computed 728 from the PSVD data, since the velocity intervenes in several of the equations. 729 Systematic underestimation of fall velocity has an effect of increasing ND730 and Z, while it decreases E. 731

Differences in the number of particles detected, and biases in the estimation of particle size and velocity, result in complex biases in the integrated variables. This is due to the different effect that these factors have on their computation, since depending on the case there are linear or inverse relationships involved. This stressed the relevance of not only an unbiased estimation of the PSVD by the disdrometers, but also of any filtering and correction scheme applied to the PSVD data during post-processing.

739 5. Conclusions

The two types of disdrometer analysed showed different PSVD spectra for 740 the same rainfall events, while the differences between two devices of the same 741 type were much smaller and compatible with random differences. In particu-742 lar, Thies devices recorded a much larger number of drops than Parsivel², but 743 also a much larger spread of the PSVD spectra, with a significant amount of 744 drops with unexpected combinations of size and velocity, most notably small 745 drops with excessively high velocities, compatible with edge events ('margin 746 fallers'). Parsivel² devices, on the contrary, recorded less drops and a PSVD 747 spectra which was much closer to the theoretical model. They also had a 748 tendency towards underestimating drop velocity with respect to both Thies 749 and a theoretical fall model. 750

Differences in the PSVD spectra resulted in significant discrepancies be-751 tween both disdrometers in all bulk precipitation parameters such as rain 752 intensity and amount, particle density, radar reflectivity, or kinetic energy. 753 These differences were found when these variables were computed by the in-754 ternal firmware of the devices, but also when they were computed by us from 755 the PSVD data. When the PSVD data were filtered by considering only par-756 ticles with diameters between 0.25 and 8 mm and by removing unlikely drop 757 size and velocity pairs, and a correction for the effective sampling area was 758 used, the magnitude of the differences was reduced although the tendency 759

remained. In all cases, the differences increased with precipitation intensity,
as did the variance between devices of the same type, in agreement with the
expectation and with previous studies.

The differences found may be explained by hardware or software differ-763 ences. Geometrical differences on the laser beams of the two devices translate 764 to different prevalence of bias-inducing effects such as edge events and double 765 detections, while differences the external design may also have a large influ-766 ence on the drop splash. The manufacturers of both disdrometers indicate 767 that corrections have been implemented to prevent or reduce the magnitude 768 of this effects, but the exact procedures are not documented. Different so-769 lutions can be adopted to limit undesired effects, both at the hardware and 770 the software level, and inspection of the resulting PVSD spectra during the 771 same rainfall events suggests that the level of correction is higher in the case 772 of Parsivel than in the case of Thies. Wether these differences are (total or 773 partially) due to hardware and design differences, or they are caused by hard-774 ware or software filtering and correction of the PSVD data, is still a question 775 with no clear answer. Since some crucial aspects of the internal functioning 776 of both devices are hidden from the final user, it is very difficult to design a 777 data treatment process that would enable making the records of Thies and 778 Parsivel disdrometers compatible and comparable across studies. 779

780 6. Acknowledgments

This work has been supported by the research projects CGL2011-24185, CGL2014-52135-C3-1-R and CGL2017-83866-C3-3-R, financed by the Spanish Ministerio de Economía, Industria y Competitividad (MINECO) and EU-FEDER. The work of M. Angulo-Martínez was supported by a postdoctoral grant by MINECO. We declare that we have no competing interests.

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1071 [TABLES AND FIGURES]

Size bins (mm)		Velocity bins (m s ^{-1})		
Thies	Parsivel	Thies	Parsivel	
	$0.000-0.125^{a}$		0.0 - 0.1	
0.125 – 0.250	$0.125 - 0.250^{a}$	0.0 - 0.2	0.1 – 0.2	
0.250 - 0.375	0.250 - 0.375	0.2 - 0.4	0.2 - 0.3	
0.375 – 0.500	0.375 – 0.500	0.4 - 0.6	0.3 - 0.4	
0.500 - 0.750	0.500 – 0.625	0.6 - 0.8	0.4 – 0.5	
0.750 - 1.000	0.625 – 0.750	0.8 - 1.0	0.5 - 0.6	
1.000 - 1.250	0.750 - 0.875	1.0 - 1.4	0.6 – 0.7	
1.250 - 1.500	0.875 - 1.000	1.4 - 1.8	0.7 – 0.8	
1.500 - 1.750	1.000 - 1.125	1.8 - 2.2	0.8 – 0.9	
1.750 - 2.000	1.125 - 1.250	2.2 - 2.6	0.9 - 1.25	
2.000 - 2.500	1.250 - 1.500	2.6 - 3.0	1.03 - 1.2	
2.500 - 3.000	1.500 - 1.750	3.0 - 3.4	1.2 - 1.4	
3.000 - 3.500	1.750 - 2.000	3.4 - 4.2	1.4 - 1.6	
3.500 - 4.000	2.000 - 2.250	4.2 - 5.0	1.6 - 1.8	
4.000 - 4.500	2.250 - 2.575	5.0 - 5.8	1.8 - 2.05	
4.500 - 5.000	2.575 - 3.000	5.8 - 6.6	2.05 - 2.4	
5.000 - 5.500	3.000 - 3.500	6.6 - 7.4	2.4 – 2.8	
5.500 - 6.000	3.500 - 4.000	7.4 - 8.2	2.8 – 3.2	
6.000 - 6.500	4.000 - 4.500	8.2 - 9.0	3.2 - 3.6	
6.500 - 7.000	4.500 - 5.125	9.0 - 10.0	3.6 - 4.1	
7.000 - 7.500	5.125 - 6.000	> 10.0	4.1 - 4.8	
7.500 - 8.000	6.000 - 7.000		4.8 - 5.6	
> 8.000	7.000 - 8.000		5.6 - 6.4	
	8.000 - 9.000		6.4 – 7.2	
	9.000 - 10.250		7.2 - 8.2	
	10.250 - 12.000		8.2 - 9.6	
	12.000 - 14.000		9.6 - 11.2	
	14.000 - 16.000		11.2 - 12.8	
	16.000 - 18.000		12.8 - 14.4	
	18.000 - 20.000		14.4 - 16.4	
	20.000 - 23.000		16.4 - 19.2	
	23.000 - 26.000		19.2 - 21.4	

Table 1: Classification or particles according to equivolume diameter (D) and fall velocity (V) bins by disdrometer type.

 $^{\rm a}$ Left empty by the manufacturer.

Variables	Units	Acronym
Rain rate, mean and max	${\rm mm}~{\rm h}^{-1}$	R, Rm, RM
Precipitation accumulated	mm	Р
Number of particles	\min^{-1}	NP, NPm
Particle density	$m^-3 mm^-1$	ND, NDm
Radar reflectivity	dBZ	Z
Kinetic energy	$\rm J~m^{-2}~mm^{-1}$	E, Em, EM
10th PSD percentile	mm	D10
50th PSD percentile	mm	D50
90th PSD percentile	mm	D90
Mean PSD	mm	Dm
10th PVD percentile	${\rm m~s^{-1}}$	V10
50th PVD percentile	${\rm m~s^{-1}}$	V50
90th PVD percentile	${\rm m~s^{-1}}$	V90
Mean PVD	mm	Vm

Table 2: Disdrometer evaluated variables. M and m stand for maximum and mean, respectively.

Table 3: Disdrometer data summary. Number of minutes recorded, errors, minutes with rain (SYNOP codes 61, 63 and 65), and high quality minutes; percentage of records in each season, and by rain intensity ranges; and maximum rain intensity.

	T1	T2	P1	P2
Total minutes	98,861	99,290	92,029	74,608
Error flags	20	33	240	103
Rain minutes	$30,\!359$	$30,\!507$	$25,\!299$	18,376
% rain minutes	30.7	30.7	27.5	24.6
High quality rain minutes	$25,\!357$	$25,\!688$	$23,\!895$	$18,\!376$
Common, high quality, rain minutes	$11,\!659$	$11,\!659$	$11,\!659$	$11,\!659$
% rain minutes in winter	27.7	27.7	28.7	33.7
% rain minutes in spring	26.6	26.1	25.3	10.9
% rain minutes in summer	11.1	11.1	11.1	11.9
% rain minutes in autumn	34.6	35.2	35.0	43.5
% minutes 0.1-2 mm h ⁻¹	84.6	83.6	86.8	85.8
% minutes 2-5 mm h ⁻¹	11.9	12.4	10.4	11.1
% minutes 5-10 mm h^{-1}	2.3	2.8	1.9	2.0
% minutes 10-25 mm h ⁻¹	0.75	0.8	0.7	0.59
$\% \text{ minutes } >25 \text{ mm h}^{-1}$	0.43	0.46	0.3	0.49
Lowest $R \pmod{\mathrm{h}^{-1}}$	0.018	0.020	0.015	0.014
Highest $R \pmod{h^{-1}}$	251	277	170	169

Table 4: Gamma Generalized Linear Mixed-Effects Model coefficients for one-minute records (random sample size of N = 1000). Refer to Table 2 for a list of acronyms of response variables.

Response	Fixed effects				Random effects	
variable	Thies		Parsivel		Mast	Residual
	coeff	p-value	coeff	p-value	std. dev.	std. dev.
NP	230.1	${<}2\times10^{-16}$	193.8	${<}2\times10^{-16}$	0.000	0.8719
D10	0.3374	${<}2\times10^{-16}$	0.4772	${<}2\times10^{-16}$	3.614×10^{-3}	0.1730
D50	0.5956	${<}2\times10^{-16}$	0.7420	${<}2\times10^{-16}$	1.488×10^{-3}	0.1899
D90	1.012	${<}2\times10^{-16}$	1.026	${<}2\times10^{-16}$	0.000	0.209
V10	1.316	${<}2\times10^{-16}$	1.793	${<}2\times10^{-16}$	1.716×10^{-2}	0.2097
V50	2.399	${<}2\times10^{-16}$	2.875	${<}2\times10^{-16}$	2.450×10^{-2}	0.1646
V90	3.818	${<}2\times10^{-16}$	3.608	${<}2\times10^{-16}$	1.200×10^{-2}	0.1445
R	1.440	1.659×10^{-7}	1.254	${<}2\times10^{-16}$	2.292×10^{-8}	1.467
ND	$21,\!600$	${<}2\times10^{-16}$	$15,\!920$	${<}2\times10^{-16}$	0.000	0.578
Z	24.55	${<}2\times10^{-16}$	23.23	${<}2\times10^{-16}$	0.000	0.2828
E	11.09	${<}2\times10^{-16}$	9.660	${<}2\times10^{-16}$	2.099×10^{-8}	0.4912

Table 5: Gamma Generalized Linear Mixed-Effects Models coefficients for event totals (sample size N = 624). Refer to Table 2 for a list of variable acronyms.

Response	Fixed effects				Random effects	
variable	Thies		Parsivel		Mast	Residual
	coeff	p-value	coeff	p-value	std. dev.	std. dev.
NP	167.5	${<}2\times10^{-16}$	146.3	${<}2\times10^{-16}$	0.000	0.8463
D10m	0.3448	${<}2\times10^{-16}$	0.4909	${<}2\times10^{-16}$	3.073×10^{-3}	0.1629
D50m	0.6061	${<}2\times10^{-16}$	0.7560	${<}2\times10^{-16}$	0.000	0.1564
D90m	0.9971	${<}2\times10^{-16}$	1.027	${<}2\times10^{-16}$	0.000	0.1566
V10m	1.351	${<}2\times10^{-16}$	1.826	${<}2\times10^{-16}$	2.027×10^{-2}	0.2036
V50m	2.465	${<}2\times10^{-16}$	2.876	${<}2\times10^{-16}$	2.607×10^{-2}	0.1375
V90m	3.791	${<}2\times10^{-16}$	3.597	${<}2\times10^{-16}$	1.907×10^{-2}	0.1114
Rm	1.051	${<}2\times10^{-16}$	0.9615	${<}2\times10^{-16}$	0.000	1.063
RM	3.351	${<}2\times10^{-16}$	3.430	${<}2\times10^{-16}$	6.788×10^{-8}	1.584
NDm	20,780	${<}2\times10^{-16}$	$15,\!930$	${<}2\times10^{-16}$	9.283×10^{-5}	0.4714
Em	11.03	${<}2\times10^{-16}$	9.505	${<}2\times10^{-16}$	1.867×10^{-7}	0.3792
Zm	22.75	${<}2\times10^{-16}$	21.55	${<}2\times10^{-16}$	1.872×10^{-7}	0.2068

Response	Fixed effects				Random effects		
variable		Thies	Parsivel		Mast	Residual	
	coeff	p-value	coeff	p-value	std. dev.	std. dev.	
Low rainfa	ll intensi	ty $(0.1 < I < 2 \text{ mm})$	$n h^{-1}$):				
NP	145.8	$< 2 \times 10^{-16}$	136.1	${<}2\times10^{-16}$	1.132×10^{-7}	0.7129	
D10	0.3481	$< 2 \times 10^{-16}$	0.4723	$< 2 \times 10^{-16}$	3.795×10^{-3}	0.1758	
D50	0.5975	${<}2\times10^{-16}$	0.7109	${<}2\times10^{-16}$	3.160×10^{-3}	0.1765	
D90	0.9440	$< 2 \times 10^{-16}$	0.9503	$< 2 \times 10^{-16}$	0.000	0.1650	
V10	1.365	$< 2 \times 10^{-16}$	1.762	${<}2\times10^{-16}$	2.189×10^{-2}	0.2156	
V50	2.416	$< 2 \times 10^{-16}$	2.768	$< 2 \times 10^{-16}$	2.189×10^{-2}	0.2156	
V90	3.639	${<}2\times10^{-16}$	3.425	${<}2\times10^{-16}$	1.145×10^{-2}	0.1202	
R	0.6675	1.659×10^{-7}	0.6202	$< 2 \times 10^{-16}$	0.000	0.6570	
ND	24,840	$< 2 \times 10^{-16}$	18,710	${<}2\times10^{-16}$	9.824×10^{-3}	0.5478	
Z	21.44	$< 2 \times 10^{-16}$	20.45	${<}2\times10^{-16}$	0.000	0.2281	
E	9.434	${<}2\times10^{-16}$	7.953	${<}2\times10^{-16}$	1.113×10^{-2}	0.4108	
Medium ra	ainfall int	ensity $(2 < I < 10)$	$mm h^{-1}$):			
NP	519.2	$< 2 \times 10^{-16}$	408.1	$< 2 \times 10^{-16}$	3.144×10^{-9}	0.4014	
D10	0.3122	$< 2 \times 10^{-16}$	0.4944	$< 2 \times 10^{-16}$	1.681×10^{-3}	0.1232	
D50	0.5936	${<}2\times10^{-16}$	0.8246	${<}2\times10^{-16}$	7.793×10^{-4}	0.1592	
D90	1.525	$< 2 \times 10^{-16}$	1.772	$< 2 \times 10^{-16}$	1.203×10^{-10}	0.1268	
V10	1.177	$< 2 \times 10^{-16}$	1.893	$< 2 \times 10^{-16}$	8.798×10^{-3}	0.1666	
V50	2.420	${<}2\times10^{-16}$	3.133	${<}2\times10^{-16}$	2.348×10^{-2}	0.1587	
V90	4.488	${<}2\times10^{-16}$	4.147	${<}2\times10^{-16}$	3.325×10^{-2}	9.908×10^{-2}	
R	4.048	1.659×10^{-7}	3.596	$< 2 \times 10^{-16}$	1.145×10^{-2}	0.1202	
ND	13,730	$< 2 \times 10^{-16}$	8,228	$< 2 \times 10^{-16}$	6.932×10^{-3}	0.3899	
Z	34.26	${<}2\times10^{-16}$	32.22	${<}2\times10^{-16}$	7.137×10^{-3}	0.1092	
E	15.09	${<}2\times10^{-16}$	13.95	${<}2\times10^{-16}$	7.105×10^{-3}	0.3521	
High rainfa	all intensi	ities (I>10 mm	h^{-1}):				
NP	1367.0	$< 2 \times 10^{-16}$	829.7	$< 2 \times 10^{-16}$	9.263×10^{-9}	0.3532	
D10	0.287	${<}2\times10^{-16}$	0.5391	${<}2\times10^{-16}$	0.000	0.1866	
D50	0.510	$< 2 \times 10^{-16}$	1.030	$< 2 \times 10^{-16}$	0.000	0.2777	
D90	1.525	$< 2 \times 10^{-16}$	1.772	$< 2 \times 10^{-16}$	1.645×10^{-2}	0.1560	
V10	1.015	$< 2 \times 10^{-16}$	2.047	$< 2 \times 10^{-16}$	0.000	0.2213	
V50	2.012	${<}2\times10^{-16}$	3.529	${<}2\times10^{-16}$	0.000	0.2672	
V90	4.992	$< 2 \times 10^{-16}$	4.467	$<\!2 imes 10^{-16}$	0.000	0.1196	
R	15.94	1.659×10^{-7}	14.33	$<\!2 imes 10^{-16}$	2.374×10^{-2}	0.2910	
ND	10,370	$< 2 \times 10^{-16}$	3,54 3 7	$< 2 \times 10^{-16}$	0.000	0.428	
Ζ	43.05	${<}2\times10^{-16}$	40.88	${<}2\times10^{-16}$	9.882×10^{-3}	8.927×10^{-2}	
E	19.84	$< 2 \times 10^{-16}$	20.81	$<\!\!2 \times 10^{-16}$	5.844×10^{-9}	0.3198	

Table 6: Gamma Generalized Linear Mixed-Effects Model coefficients for minutes with varying rainfall intensities.



Figure 1: Sampling site with four collocated disdrometers: two Parsivel² (P1 and P2, with serial numbers 304555 and 304553); and two Thies (T1 and T2, with serial numbers 0436 and 0655).



Figure 2: Accumulated precipitation (R, mm) and kinetic energy $(E, J m^{-2} mm^{-1})$ during the two years experiment (only the minutes with data on the four disdrometers are used).



Figure 3: Time series of disdrometer bulk variables during a high-intensity event (E365, 25/11/2014).



Figure 4: Time series of disdrometer bulk variables during a low-intensity event (E455, 23/02/2015).



Figure 5: Particle size and velocity density (PSVD) plots of a high-intensity event (E365, 25/11/2014). The color scale indicates the number of particles for each size and velocity class (NP). Deviations larger than 50% from the theoretical terminal velocity model (Beard, 1976; red line) are indicated with a 50% transparency.



Figure 6: Particle size and velocity density (PSVD) plots of a low-intensity event (E455, 23/02/2015). Legend as in Figure 5.



Figure 7: Kernel density plots for one-minute records.



Figure 8: Violin plots for events means and maxima. Refer to Table 2 for a list of acronyms of the variables.



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Figure A.1: Time series of disdrometer bulk variables during a high-intensity event (E365), with no corrections of the PSVD data.



Figure A.2: Time series of disdrometer bulk variables during a low-intensity event (E455), with no corrections of the PSVD data.



Figure A.3: Kernel density plots for one-minute records, with no corrections of the PSVD data.



Figure A.4: Kernel density plots for low rainfall intensities $(0.1 < I < 2 \text{ mm h}^{-1})$.



Figure A.5: Kernel density plots for medium rainfall intensities $(2 < I < 10 \text{ mm h}^{-1})$.



Figure A.6: Kernel density plots for high rainfall intensities (I>10 mm h^{-1}).

Variable	Fixed effects				Random effects	
	Thies		Parsivel		Mast	Residual
	coeff	p-value	coeff	p-value	std. dev.	std. dev.
NP	311	${<}2\times10^{-16}$	192	${<}2\times10^{-16}$	1.130×10^{-8}	1.027
D10	0.2409	${<}2\times10^{-16}$	0.5010	${<}2\times10^{-16}$	8.726×10^{-4}	0.2493
D50	0.5302	${<}2\times10^{-16}$	0.8040	${<}2\times10^{-16}$	0.000	0.2420
D90	1.126	${<}2\times10^{-16}$	1.254	${<}2\times10^{-16}$	0.000	0.2320
V10	1.199	${<}2\times10^{-16}$	1.972	${<}2\times10^{-16}$	3.062×10^{-2}	0.2420
V50	2.392	${<}2\times10^{-16}$	3.085	${<}2\times10^{-16}$	0.000	0.1760
V90	4.215	$< 2 \times 10^{-16}$	4.203	$< 2 \times 10^{-16}$	0.000	0.1641
R	1.326	1.130×10^{-4}	1.183	8.77×10^{-11}	0.000	1.660
ND	$33,\!370$	$< 2 \times 10^{-16}$	17,750	$< 2 \times 10^{-16}$	1.246×10^{-7}	0.6232
Z	24.00	${<}2\times10^{-16}$	22.45	${<}2\times10^{-16}$	0.000	0.2968
Ε	10.370	$< 2 \times 10^{-16}$	8.968	${<}2\times10^{-16}$	0.000	0.4733

Table A.1: Gamma Generalized Linear Mixed-Effects Model coefficients for one-minute records, with no corrections of the PSVD data (N = 1000).