

1 Characteristics of the raindrop distributions in RICO shallow cumulus

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7 8 9 10 **Abstract**

11
12 The physical properties of rain spectra are generally modeled using **an** analytical
13 distribution. **It is common for the Gamma distribution and to a lesser extent the**
14 **Lognormal distribution to be used.** The majority of studies **in** the literature focusing on
15 the characterization of **raindrop** distribution are based on deep convective cloud
16 observations, mostly at ground level. This study focuses on shallow cumulus rain
17 **distributions throughout the depth of the cloud layer and subcloud layer** using airborne in
18 situ measurements made with both the PMS-OAP-260X and the PMS-2DP instruments
19 during the RICO field experiment. Sampled spectra analyzed at the scale of LES
20 resolution (100m) are found to be relatively broad, with values of the shape parameter --
21 ν for the Gamma law and σ_g for the Lognormal law -- of the order of 1-3 and 1.5-2,
22 **respectively.** The dependence of the shape parameters on the **main rain variables (number**
23 **concentration, water content, mean volume diameter, sedimentation fluxes and radar**
24 **reflectivity)** is examined, and a parameterization of the shape parameters ν and σ_g **as a**
25 **function of** a power law of the rainwater content and raindrop number concentration is
26 proposed.

27 28 **1. Introduction**

30 Raindrops play a role in the lower troposphere water and energy budgets by carrying
31 water and latent energy from the cloud layer to the subcloud layer and to the surface.
32 Assuming spherical raindrops, the physical properties of the raindrop field can be
33 represented by the raindrop size (or mass) distribution at local scales, i.e. at scales of the
34 order of a few dozen meters. The evolution of the raindrop size distribution depends on
35 the interaction of various processes. In warm clouds, droplet growth is driven by
36 condensation until its collection efficiency with respect to other cloud droplets starts to be
37 significant, i.e. for diameters of the order of 40 μm . For a drop that reaches such a limit,
38 called a precipitation embryo, the drop growth rate is exclusively the result of the
39 collision-coalescence process and is roughly a function of the diameter to the power of
40 six. The transition between these two regimes is highly non linear. The growth of the
41 drops is limited on one hand by the amount of cloud water available. On the other hand,
42 large drop formation is limited by two microphysical processes: collision-induced
43 breakup and spontaneous breakup. The latter occurs for diameters of a value of the order
44 of about 10 mm (Pruppacher and Pitter, 1971). Both breakup processes contribute to a
45 broadening of the raindrop distribution. The effect of collision-coalescence-breakup
46 processes leads to an equilibrium distribution in around one hour (Hu and Srivastava,
47 1995), which corresponds to about twice the lifetime of a shallow cumulus cloud cell. In
48 unsaturated regions, the raindrop spectra evolve as the result of evaporation. In addition
49 to these processes, the sedimentation process redistributes the raindrop sizes in the
50 vertical: because large drops fall faster, the raindrop distribution tends to favor larger
51 drops at lower levels (Milbrandt and Yau, 2005; Seifert, 2008). Thus, assuming a
52 continuous and steady production of rain at cloud top, the rain distribution at a given
53 level is in steady state only if the lifetime of the precipitating event is large enough to
54 counteract the sedimentation size sorting effect. Ultimately, the local raindrop
55 distribution is the result of a coupling between advection, turbulent transport and
56 microphysical processes; collision-coalescence-breakup and sedimentation in cloud in a
57 first stage, evaporation, sedimentation and to a lower extent collision-coalescence-
58 breakup out of the cloud in a second stage.

59 Under some hypotheses, each microphysical main rain variable and process can be
60 directly expressed or parameterized as a function of the integral variables of the rain
61 distribution, mostly moments. The moment of the order p, M_p , is defined following:

62

$$M_p = \int D^p n(D) dD \quad (1)$$

63

64 where D is the particle diameter and $n(D)$ is the volume number density of raindrops with
65 diameter between D and $D+dD$. The raindrop number concentration N_r is the 0th moment
66 of the distribution. The rainwater content q_r is proportional to the 3rd moment of the
67 distribution. Both are prognostic variables in two-moment bulk schemes. In radiative
68 transfer calculation, the extinction is proportional to the 2nd moment. The radar
69 reflectivity, which is an useful quantity for remote sensing measurements, is proportional
70 to the radar reflectivity factor. Assuming Rayleigh scattering, the radar reflectivity factor
71 is the 6th moment of the distribution (Smith et al., 1975). The collection of cloud droplets
72 by raindrops (accretion) is usually parameterized as the product of cloud and rain water
73 contents (Kessler, 1969). The raindrop terminal velocity is roughly proportional to the
74 diameter to the power 0.8. Thus the sedimentation fluxes of the rain concentration and the
75 rainwater content vary as a linear function of the moments 0.8 and 3.8, respectively.
76 Hence, they are roughly dependent on M_1 and M_4 . The evaporation rate is the sum of two
77 linear functions depending roughly on the moments of the order 0.8 and 1.8.

78 Since only a limited number of rainfall integral variables are generally known (e.g. M_0
79 and M_3 in 2-moment bulk schemes, M_6 in remote sensing measurement), a hypothesis on
80 the shape of the distribution is necessary in order to derive the other microphysical
81 properties. Raindrop distributions are generally represented by the exponential law
82 (Marshall and Palmer, 1949), hereafter referred to as MP distribution, or by a Gamma
83 distribution function (Ulbricht, 1983). The latter is expressed as

$$n(D) = N \frac{1}{\Gamma(\nu)} \lambda^\nu D^{\nu-1} \exp(-\lambda D) \quad (2)$$

84 It has 3 independent parameters: the number concentration N , the slope parameter λ and
 85 the shape parameter ν . The Gamma law is a general case of the exponential function
 86 ($\nu=1$). Note that the most common expression used for the shape parameter is $\mu=\nu-1$
 87 rather than ν . The latter is used in this study because it is defined on $]0, +\infty[$, which
 88 permits plots on the logarithmic scale. The slope parameter λ is related to the mean
 89 volume diameter D_v and ν following:

$$\lambda = \frac{1}{D_v} (\nu(\nu+1)(\nu+2))^{1/3} \quad (3)$$

90 In some studies the Lognormal distribution is assessed (Feingold Levin, 1986):

$$n(D) = N \frac{1}{\sqrt{2\pi D \ln \sigma_g}} \exp\left(-\frac{1}{2} \left(\frac{\ln(D/D_g)}{\ln \sigma_g}\right)^2\right) \quad (4)$$

91 where σ_g is the geometric standard deviation and D_g is the mean geometric diameter.

92 The benefit of using these distributions is that each moment of the distribution can be
 93 analytically calculated as a function of the 3 parameters. In a 2-moment bulk scheme, 2
 94 parameters are imposed by the prognostic variables and one remains to be fixed: ν for the
 95 Gamma and σ_g for the Lognormal distribution. Figure 1 shows the moments of the order
 96 1, 2, 4 and 6 as a function of the shape parameters for fixed concentration (M_0) and water
 97 content (M_3). When ν increases, the distribution is narrower: M_p increases with ν for $p <$
 98 3, decreases for $p > 3$ and inversely for the Lognormal law. For $\nu > 10$ or $\sigma_g < 1.1$, each
 99 moment does not vary significantly because the distribution tends to the monodispersed
 100 distribution. Note that, in this study, narrow (broad) refers to spectra with high (low)
 101 value of ν or low (high) value of σ_g and not to high standard deviation values, which also
 102 depend on the mean volume diameter.

103 Since the work of Marshall and Palmer (1949) and Best (1950), a large number of studies
 104 have been dedicated to the retrieval of the value of these parameters characteristic of deep
 105 convective events. Most of these studies suggest that rain spectra are narrower than the
 106 MP distribution ($\nu=1$), with ν values roughly in the range 5-10 (Nzeukou et al. 2003;

107 Uijlenhoet et al., 2003) or more (Tokay and Short, 1996) or σ_g values of the order of 1.4
108 (Feingold and Levin, 1986). These studies are based on one-minute surface
109 measurements with the RD-69 disdrometer (Joss and Waldvogel, 1967). Ulbrich and
110 Atlas (1997) airborne 2-D precipitation probe measurements at 6 second-resolution
111 suggest broader spectra, with a mean value of 5 ($\mu=v-1=6$), than the Tokay and Short
112 (1996) mean value of 11, for the same field experiment. By analyzing one-minute
113 resolution spectra derived from video disdrometer measurements at the surface, Brandes
114 et al. (2003) also find broad spectra, with most values falling between the MP value and
115 $v=5$. Van Zanten et al (2005) find narrow drizzle spectra in stratocumulus despite the
116 coarse resolution of 2 minutes, with σ_g values of the order of 1.5-1.8.

117 Studies diverge not only in the magnitude of the shape parameter values but also in their
118 relationship with other variables. Experimental studies show a positive correlation
119 between v and the precipitation flux (Tokay and Short, 1996; Cerro et al., 1997; Nzeukou
120 et al., 2004) and numerical studies point to the narrowing of the spectra with increasing
121 mean volume diameter induced by size sorting (Milbrandt and Yau, 2005; Seifert, 2008).
122 On the other hand, PRECIP98 measurements show a negative correlation between v and
123 the precipitation flux and between v and the mean volume diameter (Zhang et al., 2001).
124 All these studies focused on deep precipitating clouds, stratiform or boundary layer
125 clouds. The lack of convergence between studies suggests a different type of rain spectra
126 according not only to the type of cloud but also to the location in the cloud system, the
127 methodology employed, the temporal and horizontal resolutions, the instruments used,
128 and instrumental biases. Until now, no study has assessed the shape parameter in shallow
129 cumulus convection.

130 In this study, the representation of the rain spectra in shallow cumulus is examined via
131 the values of the shape parameters σ_g and v . The following section describes the data set
132 and gives an insight into the vertical profiles of the measured precipitation fields; the
133 shape parameters analysis results are reported in section 3.

134

135 2. Data set and vertical structure of the precipitation field

136

137 The observations used in this study are derived from in situ shallow precipitating cumulus
138 cloud measurements collected during the RICO field experiment (Rauber et al., 2007;
139 Snodgrass, 2008; Nujiens et al 2009). Two instruments are combined to retrieve the
140 complete raindrop size distribution. The Particle Measuring Systems (PMS) OAP-260-X
141 provides droplet and drizzle size from 5 μm to 635 μm over 63 bins of 10 μm bin width.
142 The PMS-2DP measures the diameter of larger drops over 32 or 64 bins of 200 μm bin
143 width between 100 μm and an upper limit depending on the method used by the NCAR to
144 construct the particle spectra from the PMS-2DP images.

145 The Entire-in method takes into account only particles that fully cross the sampling
146 section and assumes that the diameter is the drop thickness along the diode array
147 (Heymsfield et al., 1978). The sampling volume decreases with drop diameter because
148 the upper limit of the measured diameter is restricted by the thickness of the diodes,
149 which is of the order of 6 mm. The Center-in method also takes into account partially
150 sampled drops by accounting for all particles for which the center is within the sampling
151 section. The diameter of the raindrop is assumed to be the maximum value between its
152 width along the flight path and its thickness. This method increases the 2DP sampling
153 volume and allows larger drops, up to 12700 μm , to be taken into account.

154 Large raindrop diameters are especially subject to be biased due to their non spherical
155 shape (Pruppacher and Beard, 1970; Chandrasekar et al., 1988), to the very low number
156 of such particles and to spurious counts (Heymsfield and Baumgardner 1985; Backer et
157 al., 2009). Thus, the Center-in spectra are used in this study and sensitivity tests are
158 performed according to the method used in section 3.

159 For data processed at 1 Hz, that is a resolution of about 100 m along the flight track, the
160 sampled volume is of the order of 1-4 L and 100-200 L for the PMS-OAP-260X and the
161 PMS-2DP, respectively. This is low compared to the typical value of raindrop number
162 concentration, which is about 0.1-100 L^{-1} . To increase the representativeness of the

163 sample, one can cumulate counts over a larger distance. However, because of the
164 heterogeneity of the raindrops spatial distribution, the shape of the spectra is sensitive to
165 the resolution. An increase of the sample length broadens the spectra.

166 The lower limit of the raindrop spectra D_0 , which corresponds to the separation diameter
167 between cloud droplet and drizzle, is assumed to be 75 μm . Sensitivity tests have shown
168 that the results presented here are not sensitive to this threshold, at least over the range
169 50-100 μm . Spurious counts, which affect both low and high diameters (Backer et al.,
170 2009), are removed in 2DP and OAP-260-X measurements. Similarly to Yuter and Houze
171 (1997), all non-consecutive bins above 1500 μm are set to zero, and the isolated positive
172 bins in OAP-260-X are excluded. Because the moments of the distribution are sensitive
173 to the extremities of the distribution, sensitivity tests are further performed in section 3.
174 Finally, the first bin of the PMS-2DP is removed to avoid overlap with the OAP-260-X
175 measurements.

176 Of the nineteen RICO flights analyzed in this study, thirteen are characterized by
177 significant rainy events (RF01, RF03, RF04, RF05, RF07, RF08, RF10, RF11, RF13,
178 RF14, RF15, RF16, RF19) and six are rejected due to the insignificant number of rain
179 samples (RF02, RF06, RF09, RF12, RF17, RF18). Rain spectra are defined here as
180 samples with rain water content $q_r > 0.010 \text{ g m}^{-3}$. The total number of precipitating
181 samples at 1 Hz resolution is about 21000. During RICO, the NCAR C-130 aircraft flew
182 through the cloud field at different altitudes between about 100 m and 3 km. To
183 distinguish between in-cloud and clear-sky samples, we used data from the Fast Forward
184 Scattering Spectrometer Probe (FFSSP) instrument (Brennguier et al., 1998) that provides
185 the droplet size distribution from ~2 to 50 μm in diameter for the flights RF07, 08 and 11.

186

187 The vertical structure of the main rain variables is represented in Fig. 2. On the upper
188 panel, the first two plots show the number of rain spectra sampled at each level of the
189 lower troposphere, in cloudy air and in clear air, respectively. Because the aircraft was
190 pointing towards the cloud cells, a large part of the rain spectra (almost 60%) were
191 sampled in clouds. The third plot shows the vertical profile of the cloud liquid water

192 content derived from the FFSSP data. The following panels show the profiles of the rain
193 concentration N_r , the rain number concentration flux F_{Nr} , the rain water content q_r , the
194 precipitation flux F_{qr} and the rain mean volume diameter D_v . For each parameter, the two
195 first profiles correspond to in-cloud and clear-sky samples (left and middle column,
196 respectively) for the three flights with FFSSP data. The last profile (left column)
197 corresponds to the statistics of all samples of the thirteen flights. Box plots with 5th, 25th,
198 50th, 75th and 95th percentiles of the distribution are used to indicate the spread of the
199 data. Symbols are mean values for each flight and are superimposed to illustrate the flight
200 to flight variability.

201 Values are averaged over the rain fraction at the corresponding level. Hence, these
202 profiles are not directly comparable to profiles averaged over the whole domain or over
203 the projected cloud fraction. Above the cloud base, some rain falls in clear sky. This
204 feature may be due to the wind shear, to turbulent motions, or to the fact that raindrops
205 have a longer lifespan than cloud droplets. Such a pattern was reproduced by the Dutch
206 atmospheric model DALES LES simulations of shallow cumulus even without shear (not
207 shown). However, in LES simulations, a large part of the rain mass falls in clear sky,
208 which is not suggested here by the q_r and the F_{qr} profiles. Finally, Fig. 2 reveals that all
209 rain quantities are larger in clouds mainly due to evaporation that occurs in clear sky.

210 The profiles of q_r and F_{qr} do not show a particular trend with altitude. While evaporation
211 leads to a decrease of their domain-average value, here values are averaged over the rain
212 fraction, which decreases with height, compensating for the effects of evaporation. In
213 contrast, the rain number concentration N_r (and the rain concentration flux) and the mean
214 volume diameter D_v decrease and increase with decreasing altitude, respectively. All
215 processes (collection, evaporation, sedimentation) contribute to a decrease of the number
216 concentration and of the rain concentration flux, which is consistent with these
217 observations. The dispersion of the mean volume diameter is small, in particular in the
218 subcloud layer, and it exhibits the same trend in cloudy air and in clear air, suggesting
219 that its evolution is mainly driven by size sorting. The trends shown here are similar to
220 those observed in drizzling stratocumulus clouds (Wood et al., 2005), except that here

221 rain concentration and mean volume diameter also vary above the lifting condensation
222 level, both in and outside the cloud.

223 In comparison to the results of van Zanten et al. (2010) (their Figure 8), the profiles show
224 the same trends, with a pronounced increase of N_r with the altitude while q_r remains more
225 or less constant. However both profiles reveal higher values with median values of N_r and
226 q_r ranging from 1 to 100 L^{-1} and from 0.1 to 0.3 gm^{-3} , respectively. These differences
227 come from the cases selected here: 9 precipitating cases have been added and 3 cases
228 with a very low precipitation amount have been removed. It follows that the statistics are
229 shifted to larger values as reflected by the flight average values. Note that the profiles
230 presented here are closer to the simulations of the LES models reported in van Zanten et
231 al. (2010). As shown by the box plots, the scatter of the rain variables is large, especially
232 for the rain water content that cover about two orders of magnitude. This scatter is due to
233 the large heterogeneity of the rain field inside a given cloud system and to the differences
234 in the microphysical and macrophysical properties of the sampled cloud systems. In
235 boundary layer clouds, the strength of the precipitation production depends on both the
236 cloud droplet concentration and liquid water path or cloud depth (Geoffroy et al., 2008;
237 Jiang et al., 2010, Burnet and Brenguier, 2010), that both vary among the different flight
238 cases. However note that for the profiles of N_r , F_{nr} and D_v , both box plots and flight
239 averages follow the same pronounced vertical trend reflecting the consistency of the
240 observations.

241 Some studies have examined the relationship between the slope parameter λ and the
242 shape parameter v for remote sensing retrieval of the rain distribution characteristics,
243 mainly the precipitation flux, from radar measurements (Zhang et al., 2001; Chang et al.,
244 2009). Atlas and Ulbricht (2006) suggest that there is no universal relationship that would
245 describe all types of storm spectra accurately. The RICO measurements encompass a
246 large range of rain microphysical properties and confirm this fact. Indeed, assuming that
247 the Gamma distribution gives an accurate representation of the rain spectra, λ depends on
248 v and D_v (Eq. 3). Because the profile of D_v varies significantly with height, it follows that
249 the λ - v relationship depends necessarily on the altitude.

250 This study is restricted to the estimation of the shape parameter of both Lognormal and
251 Gamma laws assuming that N_r and q_r are known, as is the case in a simulation using a 2-
252 moment bulk microphysics scheme. Figure 3 shows the space parameter of N_r and q_r for
253 all RICO spectra at 1 Hz resolution. The reported values cover a large range of rain
254 properties from drizzle, with about 50 % of the drop concentration values greater than 5
255 L^{-1} and 10% greater than 50 L^{-1} , and intense precipitating events with samples having
256 high local rainwater content between 1 and 10 $g\ m^{-3}$. The mean volume diameter ranges
257 from 100 μm to about 1 mm near the surface. Most of the measurements are performed
258 inside clouds or close to clouds rather than in clear sky. As a result, the statistics are
259 slightly biased toward initial stages of precipitation formation. Nevertheless, as attested
260 by Figure 3, the data set covers a large range of values, hence we assume in the following
261 that it is representative of rain spectra in shallow cumulus.

262

263 3. Shape parameters analysis results

264

265 In this section, the ability of the Lognormal and the Gamma distributions to represent
266 shallow cumulus drop spectra is evaluated. The method used is the one detailed in
267 Geoffroy et al. (2010) (hereafter G10) for cloud droplet spectra analysis. The raindrop
268 spectra are assumed to be described by an analytical distribution. For each moment
269 representative of a physical process M_1 , M_2 , M_4 and M_6 , the shape parameter is calculated
270 numerically by a minimization of the distance between the measured moment and the
271 analytical moment. This method is similar to the commonly used method of moments
272 (Waldvogel 1974; Ulbrich 1983) applied with M_0 , M_3 and a 3rd moment that is the one to
273 parameterize. It has the benefit of providing the exact value to use to represent a
274 considered moment and avoiding negative values for v , as it can be found by analytical
275 calculation, for instance in Zhang et al. (2001). Some studies (Ulbrich and Atlas, 1998;
276 van Zanten et al., 2005) consider truncated functions. However, the assumed distributions
277 are not truncated when used in models or for remote sensing parameter retrieval in order
278 to avoid too many complex calculations. Moreover, the use of complete distributions

279 allows analytical integrations. For these reasons, this study is limited to complete
280 functions. Moreover, such truncations do not significantly modify the results.

281

282 Figure 4 shows the shape parameters for each moment (M_1 , M_2 , M_4 and M_6) estimation as
283 a function of the considered moment. The number of samples in each moment class is
284 represented in the lower row. The value of v is represented on a log-scale because of the
285 strong dependence of moments with $\log(v)$ (Fig. 1). According to G10, the circles and
286 triangles are the shape parameter values that minimize, in each moment class, the
287 arithmetic and the geometric standard deviation of the absolute and relative errors,
288 respectively.

289

290 For each minimization, there is a strong scatter of the shape parameter. The values of
291 v range roughly from 1 to 10. As a general trend, we observe that spectra become
292 narrower, as shown by the increase of σ and the decrease of v for both percentiles and the
293 mean values, as the value of the considered moment increases. This trend is especially
294 pronounced for the M_1 and the M_2 minimizations. The M_6 minimization gives narrower
295 spectra on average, especially for the Lognormal model, because of the highest
296 dissymmetry of this function. However, high order moments are sensitive to the presence
297 of large drops. When spurious counts are not cleaned, broader spectra are obtained for the
298 M_6 minimization. Despite the large scatter observed in the shape parameters and the
299 dependence of the results on the chosen moment, data are merged together in order to
300 derive a trade-off value of the shape parameters and to determine a single law
301 representative of all processes.

302

303 The trade-off values v^* and σ_g^* , of the Gamma and the Lognormal law, respectively, are
304 calculated by averaging the 80 optimum shape parameter values in each bin following
305 G10 for the different resolutions 1, 0.5, 0.2, and 0.05 Hz (i.e., a distance of the order of
306 100, 200, 500 and 2000 m, respectively). The results are summarized in Table 1.

307

308 A value of 3.2 for v^* and 1.63 for σ_g^* is obtained from the cleaned spectra (noted E2).
309 The broadness of the spectra increases when **when the resolution decreases**, as expected,
310 because of the high heterogeneity of the rain field. At the scale of the cloud cell,
311 distributions are close to the MP distribution ($v=1$). Table 1 also shows the arithmetic and
312 geometrical means of each ensemble of shape parameter values. The geometric mean of
313 the Gamma law shape parameter v and the arithmetic mean of the Lognormal law shape
314 parameter σ_g are close to the trade-off values v^* and σ_g^* , respectively. These results
315 suggest that such methods of averaging (geometric mean for the Gamma and arithmetic
316 mean for the Lognormal) are adequate for estimating the shape parameter. Moreover, this
317 **result** is consistent with the logarithmic and **the** linear dependency of the moments for the
318 Gamma and the Lognormal laws, respectively. The arithmetic mean, generally used in
319 studies to retrieve the characteristic v value of the rain distributions, has significantly
320 higher values.

321 To gain insight into the errors associated with the spurious count for both large and small
322 drops and those associated with a lack of statistical representation, sensitivity tests to the
323 tail of the rain spectra were performed. Without removing the spurious count, the Entire-
324 in method (E) and Center-in method (C) give similar results, which suggest a low
325 contribution of the drops larger than 6 mm, with a v^* value of the order of 2. This value
326 should give a lower boundary for v^* . Truncations under 300 μm in diameter (i.e. use of
327 only 2DP measurements), above 1.5 mm and both show that the shape parameter value is
328 mostly sensitive to the presence of the smallest drops. Spectra obtained are narrower,
329 with an extreme value of v^* equal to 9 at 1 Hz, which should give a comfortable
330 estimation of its upper bound. Such truncated spectra are close to the Joss and Waldvogel
331 (1967) disdrometer range. The 0.05 Hz value of v^* is close to that derived from most
332 previous studies.

333 **The data for the shape parameter v are reported on Figure 5a-f as function of N_r , D_v and**
334 **q_r , in order to examine the sensitivity of this shape parameters to variables prognosticated**

335 in 2-moment bulk schemes. Only M_1 and M_4 moment values are presented here because
336 they are the most important with respect to the parameterization purpose, especially for
337 the sedimentation and the evaporation processes. The largest scatter in the 6th box plot of
338 Fig. 5c,d corresponds to the transition between the OAP-200-X and the 2DP
339 measurements marked by an important decrease in the size resolution (from 10 to 200
340 μm). Measurements show a clear negative trend as a function of q_r , as already depicted in
341 Fig. 4. In contrast no obvious trend is observed for N_r and D_v over the whole range. For
342 both lowest and largest D_v values, v is large (median values > 5), corresponding to
343 narrow size distributions. The broadest spectra correspond to large concentration values
344 greater than about 4 L^{-1} and intermediate mean volume diameter values from about 200 to
345 $400 \mu\text{m}$, but with a large dispersion as reflected by the 25th -75th percentile interval that
346 could reach an order of magnitude.

347 At the early stage of the rain formation, samples are characterized by high concentration
348 values, especially in the upper part of the cloud as attested by the figure 2, low D_v values
349 and narrow spectra. As drops growth by collision-coalescence and are mixed by
350 turbulence, that is for high rainwater content samples, the size spectra broadens and the
351 mean volume diameter reaches intermediate values while the concentration slightly
352 decreases but still remains relatively high. As a result, the flight average concentration
353 values are larger than 10 L^{-1} above 1500 m as indicated by Fig. 2. Consequently, spectra
354 with large concentration may be young narrow spectra characterized by low mean
355 volume diameter, or on the opposite aged broad spectra with a large amount of rain. This
356 explains the large scatter of v for large concentration values. The vertical profiles of Fig.
357 5g,h. show an increase of v with decreasing altitude. This trend is more pronounced in the
358 subcloud layer. It is consistent with experimental studies that show narrower distributions
359 at the surface than in clouds (Tokay and short, 1996; Ulbrich and Atlas, 1998) and with
360 1-D numerical studies focusing on the effect of size sorting (Milbrandt and Yau, 2005;
361 Seifert, 2008).

362 The shape parameters retrieved here differ from those reported in previous studies that
363 focused on deep convective events for similar spatial and horizontal resolutions. These

364 discrepancies are likely due to differences in rain characteristics specific to the cloud
365 regime. In shallow cumulus the mean volume diameters are lower and the rain number
366 concentrations are larger than in deeper clouds. They can also be partially attributed to
367 instrumental limits, averaging procedures and the location of the samples. As in G10,
368 ACE-2 stratocumulus case measurements were also analyzed. However, they have not
369 been included here because the particle counter used during ACE-2 has an upper
370 boundary too low (350 μm) to cover the complete range of raindrop diameters. Indeed,
371 the drop number in the last bin was often non zero indicating that the spectra were
372 truncated. However the results obtained by analyzing the ACE-2 data set are in
373 agreement with the RICO spectra typical of drizzle (Fig. 5 a-f), i.e. with σ_g^* values of the
374 order of 1.5 and v^* values of the order of 5. Moreover, these values are quantitatively
375 consistent with van Zanten et al. (2005) DYCOMS-II stratocumulus two-minute averaged
376 spectra.

377 Because samples are mainly in clouds or close to clouds, trade-off values derived in this
378 study may be more representative of the first stages of rain development than of subcloud
379 layer rain spectra. However, because these large drops reach the ground and are not
380 subject to complete evaporation, it may be more important to represent the drop size
381 distribution in the upper levels of the cloud layer in order to accurately represent the
382 raindrop growth and evaporation. If raindrops are size-sorted during their fall and spectra
383 narrower than predicted, it will lead to an overestimation of the fall velocity. However,
384 evaporation of a large raindrop is low because its lifetime in subsaturated air is short. A 2
385 mm drop falling in an 80% relative humidity environment covers a distance of 2 km in 4
386 minutes and loses only 3 % of its mass. In contrast, a 200 μm drop in the same conditions
387 evaporates completely after 11 minutes and after a distance of about 700 m. Thus, the
388 predicted amount of rain that evaporates and the amount of precipitation that reaches the
389 ground would not be considerably biased.

390 As for cloud droplet spectra (G10), the shape parameter is mostly sensitive to the water
391 content as shown by Figure 5e,f. However the size sorting process also modulates the
392 drop spectral width. For samples with low q_r , spectra are predominantly narrow (low $1/v$)

393 whatever the value of N_r . For samples with large q_r , the spectra are predominantly broad
 394 for large N_r and narrow for small N_r due to size sorting. Thus we parameterize the shape
 395 parameter as a function of a power law of q_r and N_r . Figure 6 a, b shows scatterplots of v
 396 and σ_g as a function of $(N_r q_r)^{0.25}$ and $(N_r q_r)^{0.1}$, respectively, for the 4 moments and the
 397 values that minimize both absolute and relative errors in each bin. The percentile
 398 intervals indicate that the data dispersion increases as $(N_r q_r)$ increases, especially for
 399 moments M_1 and M_2 . This is consistent with Figure 5a-d that reveals that the spread of v
 400 is larger for large values of N_r while it remains constant over the q_r range.

401 For each law, the resulting 80 optimum parameters are fitted which leads to the following
 402 expressions:

$$v^p = 18 / (N_r q_r)^{0.25},$$

$$\sigma_{g^p} = 1 + 0.30 [(N_r q_r)^{0.1}]$$
(5)

403 where q_r is expressed in g m^{-3} and N_r in m^{-3} .

404 In order to compare the accuracy of each analytical distribution to represent the rain
 405 spectra, relative and absolute errors between measured and theoretical moments are
 406 calculated. Table 2 summarizes the offsets and standard deviations of the absolute and
 407 relative errors over the whole range of moment values calculated for the gamma and the
 408 Lognormal distribution, with trade-off and parameterized values. Both laws give similar
 409 results. The parameterized expressions improve the results in terms of both bias and
 410 standard deviation.

411

412 **Conclusion**

413

414 In situ measurements of rain collected during the RICO experiment were analyzed in
 415 order to validate the commonly used analytical representation of raindrop size
 416 distribution and quantify their broadness for shallow cumulus clouds. Data from the
 417 PMS-OAP-260-X and the PMS-2DP, were combined to retrieve the complete raindrop

418 size distribution. Thirteen flights with significant rain events have been selected. The
419 aircraft sampling strategy provides a comprehensive set of raindrop spectra typical of
420 trades shallow cumulus clouds by flying at different levels in the lower troposphere. First,
421 the vertical profiles of the microphysical rain variables were examined. It is shown that
422 the rain number concentration and the mean volume diameter decrease and increase with
423 decreasing altitude, respectively, whereas the rain water content remains more or less
424 constant. Both box plots with percentiles of the distribution of the observations and flight
425 average values follow the same pronounced vertical trend reflecting the consistency of
426 the observations.

427 Next, the broadness of the size distribution was studied by analyzing the relationship
428 between a considered moment of the size distribution and the two main rain variables
429 used in microphysical schemes: the rain mixing ratio and the rain number concentration.
430 For each moment representative of a physical process M_1 , M_2 , M_4 and M_6 , the shape
431 parameter is calculated numerically by minimizing the distance between the measured
432 moment and the derived analytical moment. For a given spectra, there is generally not a
433 single value of the shape parameter that accurately represents each moment
434 simultaneously. As a general trend, we observe that spectra become narrower as the value
435 of the considered moment increases. Nevertheless, a constant trade-off value is proposed
436 for both the Gamma law and the Lognormal law. On the ensemble, spectra are found to
437 be broad at the scale of a LES simulation (~ 100 m), with trade-off values v^* of the order
438 of 3.2 and σ_g^* of the order of 1.63. At a coarser scale, distributions tend to be broader,
439 with values of the shape parameter close to the MP value, which reflects the
440 heterogeneity of the raindrop field. Given the differences in the altitude of the samples, as
441 well as instrumental issues, these results are consistent with studies of the literature
442 focusing on deep convective events. Sensitivity tests to the extreme values of the drop
443 sizes suggest that the contribution of the smallest drops to the broadness of the
444 distribution is important. The Lognormal and the Gamma laws give similar results.
445 However, the Gamma law allows analytical integration -- for instance, the integration of
446 the sedimentation flux using Roger et al (1993) parameterization of the terminal velocity.

447 As a second step, the dependency of the shape parameter as a function of the variables
448 prognosticated by a LES microphysical scheme was explored. Measurements show a
449 clear negative trend as function of the rainwater content, but no obvious trend as function
450 of the drop concentration neither of the mean volume diameter. These results are
451 consistent with the microphysical processes involved. Indeed, at the early stage of the
452 rain formation samples are characterized by high concentration values, low mean volume
453 diameter values and narrow spectra. As drops growth by collision-coalescence, rain
454 becomes more intense and the size spectra broaden. Finally, the rain spectra tend to be
455 narrower near the surface due to size sorting. In order to take into account this behavior, a
456 parameterization as a function of a power law of ($q_r N_r$) that improves the representation
457 of the rain spectra was developed for the LES scale. However, LES simulations of
458 precipitating shallow cumulus clouds showed that a change of ν from 1 to 11 impacts the
459 mean LWP of about 20% after 2 to 6 hours of simulations (not shown). These tests also
460 suggested that the use of the tradeoff value should be sufficient to represent the
461 magnitude of the precipitation rate in shallow cumulus clouds. Questions remain for deep
462 convection. Indeed a variable shape parameter may impact significantly the results in
463 heavily precipitating clouds (Shipway and Hill, 2012). Moreover, the measurements of
464 raindrop spectra are somehow limited by statistics issues due to the low number of
465 raindrops and by instrumental biases. These measurements are important for
466 reconstructing rain history in the lower troposphere and subsequently for constraining
467 rain formation — the main source of uncertainty in precipitation calculation — at the
468 scale of the cloud system. The results presented here highlight needs to improve particle
469 measurements over the whole spectrum range as well as to provide such data at all stages
470 of rain development.

471

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476

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633

		1 Hz (~100m)	0.5Hz (~200m)	0.2 Hz (~500m)	0.05 Hz (~2000m)
E2	v^*	3,2	2,7	2,2	1,6
E2	$\langle V \rangle_{\text{geom}}$	3,5	3,0	2,5	1,8
E2	$\langle V \rangle_{\text{arith}}$	6,7	5,5	4,4	3,2
E2	σ_g^*	1,63	1,67	1,72	1,81
E2	$\langle \sigma_g \rangle_{\text{geom}}$	1,59	1,63	1,68	1,76
E2	$\langle \sigma_g \rangle_{\text{arith}}$	1,62	1,66	1,71	1,79
E	v^*	2,4	1,9	1,5	1,0
C	v^*	2,2	1,8	1,3	0,9
E2 <1500	v^*	3,3	2,8	2,3	1,8
E2 >300	v^*	8,0	7,6	6,9	5,9
E2 300-1500	v^*	9,0	8,6	8,1	7,3

634

Table 1

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636

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638

639

640

Values of v^* , σ_g^* , the arithmetic mean v_{arith} , σ_{arith} , and geometric mean v_{geom} σ_{geom} of the ensemble of shape parameter values. And values of v^* for spectra reconstructed using the Center-in (C) method, Entire-in method (E), spectra truncated above 1500 μm (<1500), under 300 μm (>300) and both (300-1500). All values are given for four resolutions: 1, 0.5, 0.2 and 0.05 Hz.

	M_1	M_2	M_5	M_6
Lognormal, $\sigma_g^*=1.63$	$\mu_{log} \quad \sigma_{log}$			
	1.08 1.32	1.05 ± 1.27	1.0 ± 1.41	1.42 ± 3.21
	$\mu_{abs} \pm \sigma_{abs}$			
	0.7 ± 3.0 ($\mu\text{m cm}^{-3}$)	319 ± 1232 ($\mu\text{m}^2 \text{cm}^{-3}$)	-2.4 ± 16.8 ($10^8 \mu\text{m}^5 \text{cm}^{-3}$)	-4 ± 184 ($10^{14} \mu\text{m}^6 \text{cm}^{-3}$)
Gamma, $\nu_j^*=3.2$	$\mu_{log} \quad \sigma_{log}$			
	1.06 ± 1.32	1.06 ± 1.27	0.93 ± 1.41	0.87 ± 3.21
	$\mu_{abs} \pm \sigma_{abs}$			
	0.7 ± 2.9 ($\mu\text{m cm}^{-3}$)	328 ± 1247 ($\mu\text{m}^2 \text{cm}^{-3}$)	-3 ± 18 ($10^8 \mu\text{m}^5 \text{cm}^{-3}$)	19 ± 184 ($10^{14} \mu\text{m}^6 \text{cm}^{-3}$)
Lognormal, σ_g^p	$\mu_{log} \quad \sigma_{log}$			
	1.07 1.27	1.05 1.22	1.0 1.32	1.38 2.62
	$\mu_{abs} \pm \sigma_{abs}$			
	0.1 ± 2.2 ($\mu\text{m cm}^{-3}$)	122 ± 622 ($\mu\text{m}^2 \text{cm}^{-3}$)	0.3 ± 8.4 ($10^8 \mu\text{m}^5 \text{cm}^{-3}$)	69 ± 827 ($10^{14} \mu\text{m}^6 \text{cm}^{-3}$)
Gamma, ν^p	$\mu_{log} \quad \sigma_{log}$			
	1.02 ± 1.26	1.04 ± 1.22	0.95 ± 1.32	0.91 ± 2.55
	$\mu_{abs} \pm \sigma_{abs}$			
	-0.3 ± 2.7 ($\mu\text{m cm}^{-3}$)	74 ± 570 ($\mu\text{m}^2 \text{cm}^{-3}$)	1 ± 9 ($10^8 \mu\text{m}^5 \text{cm}^{-3}$)	-1 ± 122 ($10^{14} \mu\text{m}^6 \text{cm}^{-3}$)

642

Table 2

643

Values of the geometric mean μ_{log} and the geometric standard deviation σ_{log} of the Log errors and the

644

arithmetic mean μ_{abs} and the arithmetic standard deviation σ_{abs} of the absolute errors of calculated for M_1 ,

645

 M_2 , M_4 , M_6 , for the Lognormal and the Gamma parametric functions, when using the constant trade-off

646

tuning parameters values, σ_g^* and ν^* and the parameterized value as a function of $N_{t,q}$, σ_g^p and ν^p .

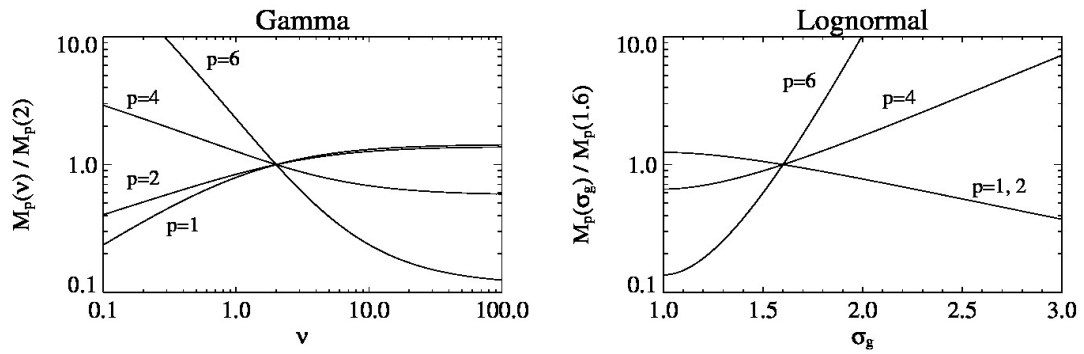
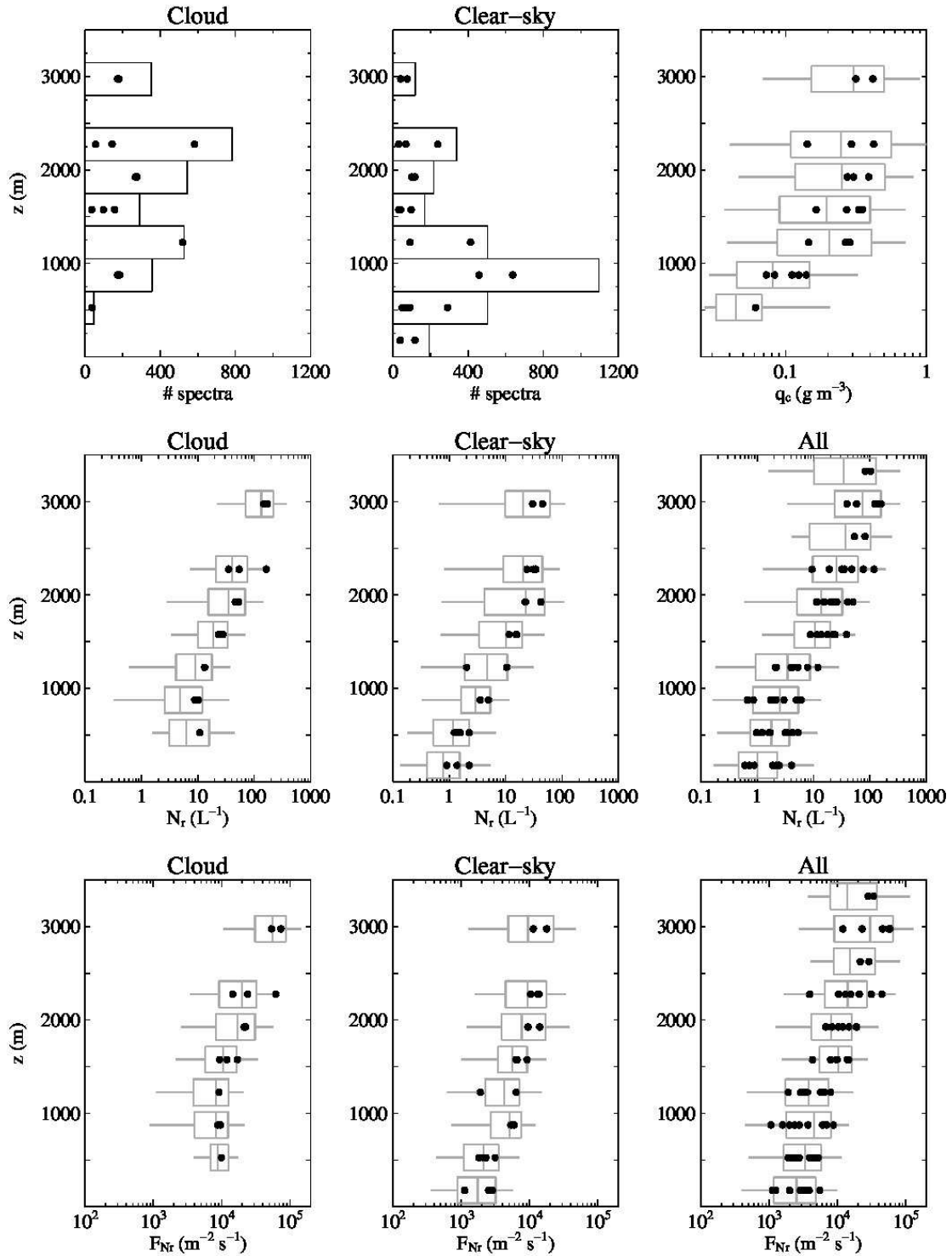


Figure 1: Relationship between the moments of the order $p=1, 2, 4, 6$ and the shape parameter for the Gamma function (left) and the Lognormal function (right). Each moment is normalized by the value corresponding to $v = 2$ for the Gamma function and $\sigma_g=1.6$ for the Lognormal function.



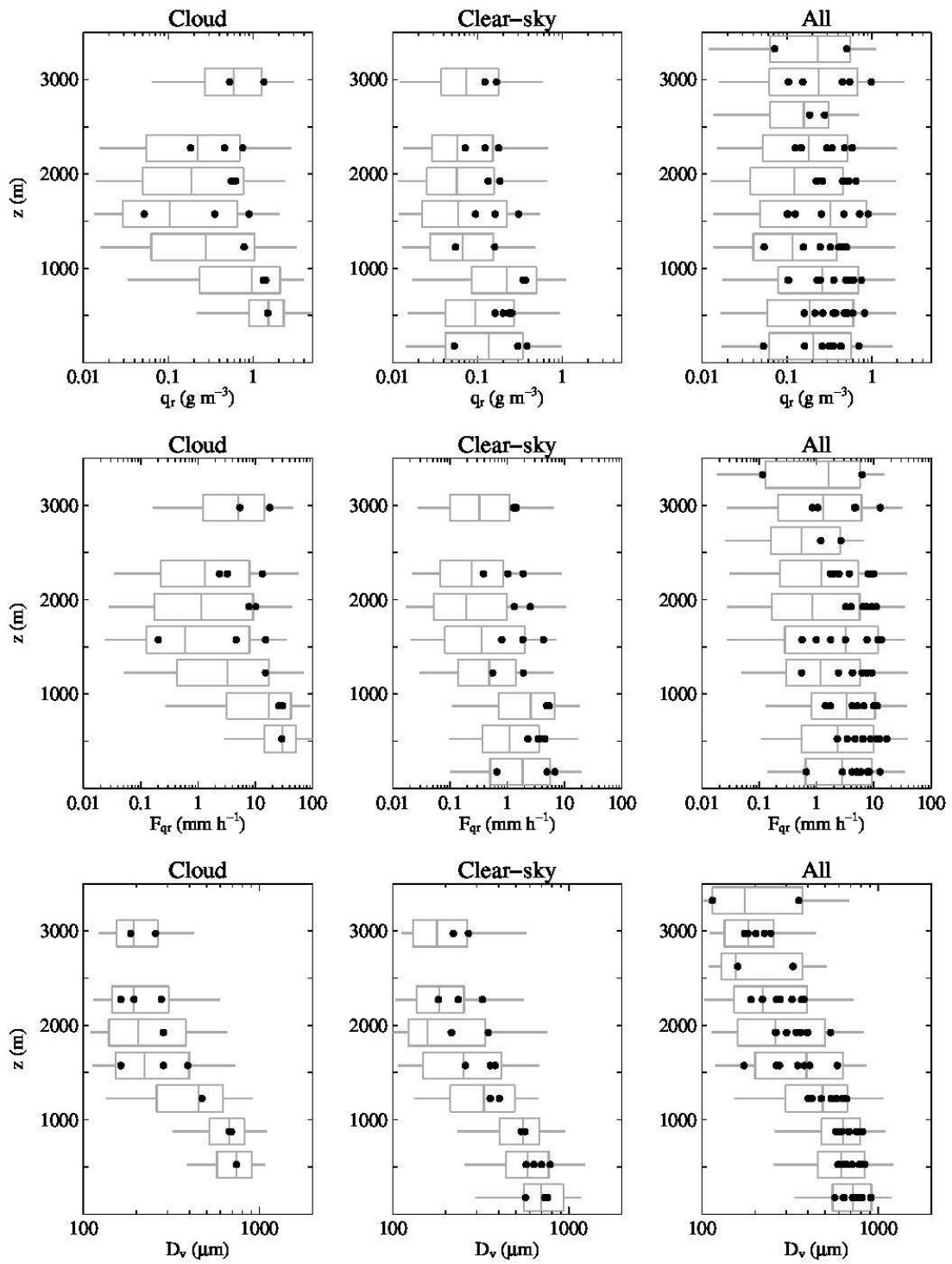


Figure 2

Total number of rain sampled spectra in cloud region (top row, left) and in clear sky (top row, center). Vertical profile of statistical distribution of cloud water content q_c (top row,

right) sampled at 1 Hz for flights with Fast-FSSP measurements available and vertical profile of statistical distribution of the rain variables sampled at 1 Hz for the rain concentration N_r , the rain concentration flux F_{Nr} , the rain water content q_r , the precipitation flux F_{qr} and the rain mean volume diameter D_v , in the cloud region (left) and in the clear sky region (middle), for flights with Fast-FSSP available, and in all regions (right). The boxplots denote the 5th, 25th, 50th, 75th and 95th percentiles of the variable distribution in every 300 m layer. Full circles are mean values for each flight.

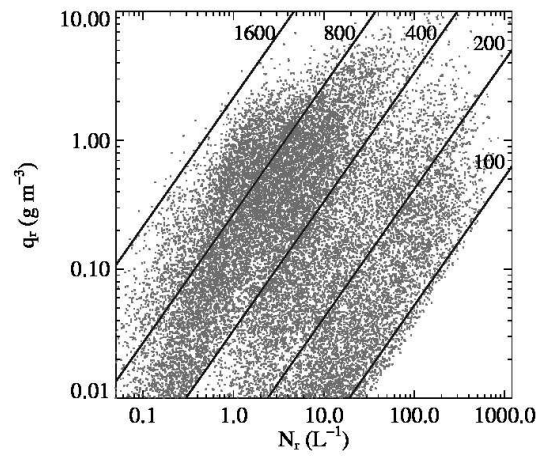


Figure 3:

Scatter-plot of the drop number concentration, N_r , and the rain water content, q_r , for drop spectra sampled at 1 Hz.

Lines represent constant mean volume diameters for $D_v = 1600, 800, 400, 200, 100\ \mu m$.

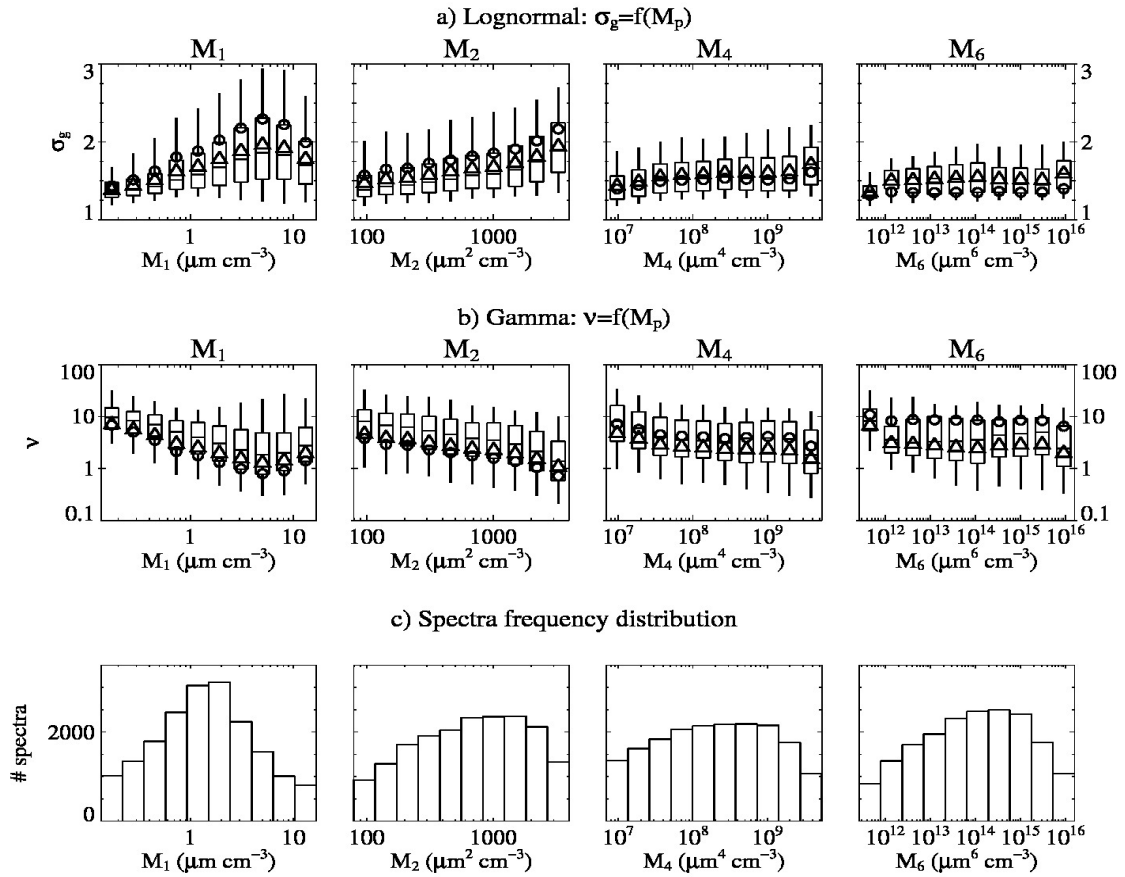


Figure 4 Statistical distribution of the shape parameter values as a function, from left to right, of the M_1 , M_2 , M_4 , and M_6 moment values. The X-axis is divided into 10 classes on a Logscale. The boxplots denote the 5th, 25th, 50th, 75th and 95th percentiles of the shape parameter distribution in each class. The circles and triangles denote the tuning parameter value that minimizes the standard deviation of the absolute error and the geometric standard deviation of the Log error in each class, respectively. The top and second rows are for the Lognormal function and the Gamma function, respectively. The third row shows the number of sampled spectra in each moment class.

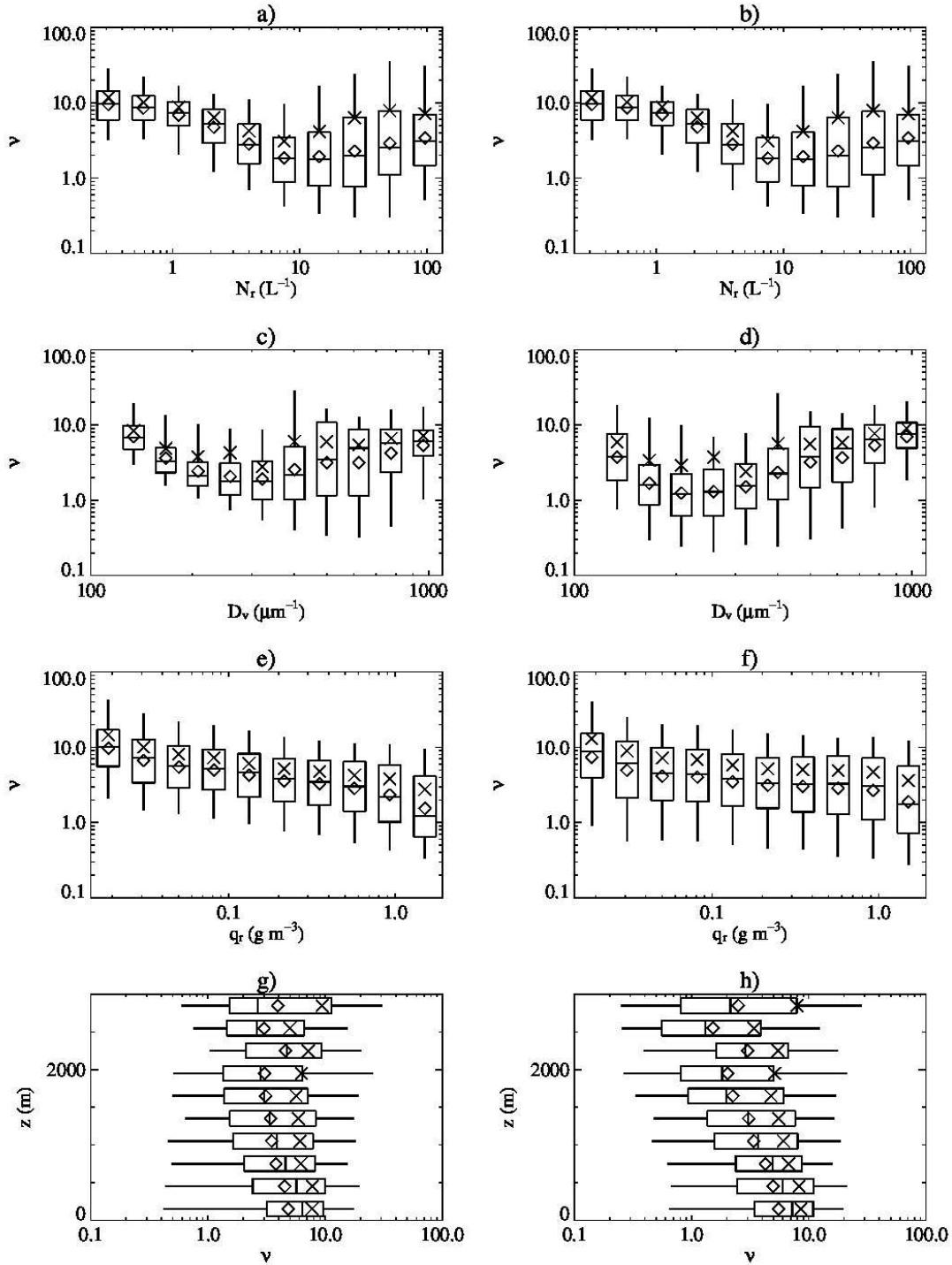


Figure 5 Statistical distribution of the shape parameter values as a function of rain number concentration N_r (upper line), rain mean volume diameter D_v (2nd line) and rainwater

content (3rd line) and profile of the statistical distribution of the shape parameter values (lower line) for the M_1 minimization (left row) and the M_4 minimization (right row). The boxplots denote the 5th, 25th, 50th, 75th and 95th percentiles of the shape parameter distribution in each class. The diamonds and crosses denote the arithmetic mean and the geometric mean in each class, respectively.

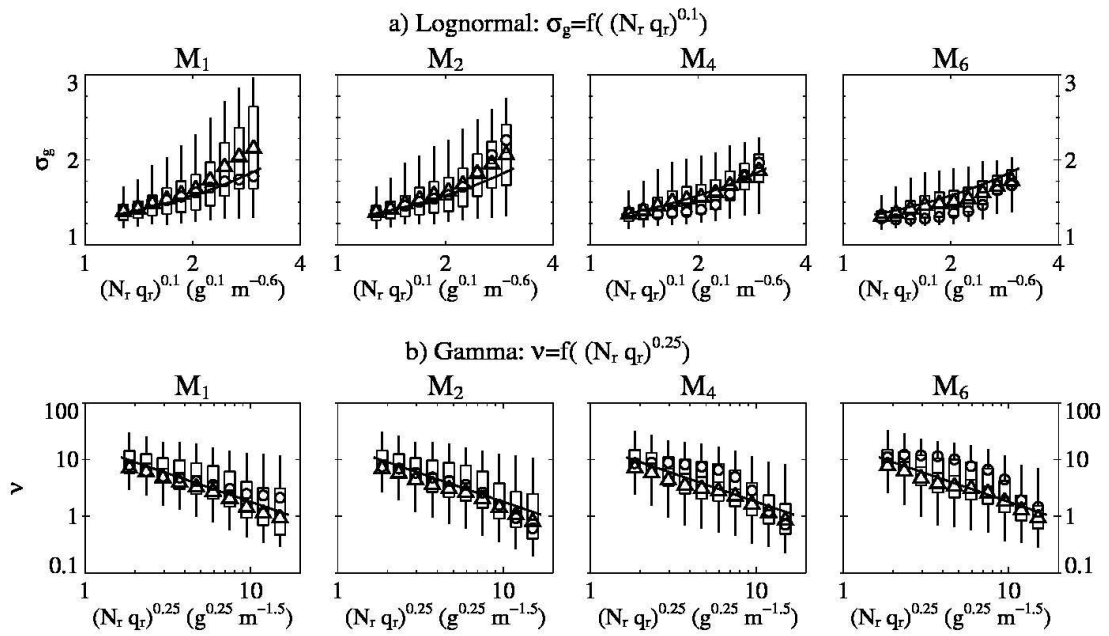


Figure 6: same as Fig. 4 but plotted as a function of a power law of $q_r N_r$. The thick lines represent the proposed parameterizations for the variable shape parameter.