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# Impact of integration time steps of rain drop size distribution on their double-moment normalization: a case study in northern Benin

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**Abstract.** This study analyzed the sensitivity of the double-moment normalization of rain drop size distributions (DSDs) to their integration time steps. Six DSD datasets measured at different integration time steps were modeled by the double-moment normalization method. Six couples of reference moments are used for the normalization. Three DSD models (gamma, generalized gamma and log-normal) are used for modeling. Statistical criteria are used to assess the estimation of the moments of the DSDs made by the models. As results, the analysis highlighted the dependence of the shape function parameters on the DSD integration time step. It is also noted that the three DSD models are well adapted to the data used. Nevertheless, the generalized gamma model is recommended because of the independence of the moment estimation errors with respect to the integration time step, which is not the case for the other models.

**Keywords.** Rain DSD, Double-moment normalization, Integration time steps, Unimodal DSD models, Generalized gamma, Weather radars, West Africa rainfall.

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## 1. Introduction

Modeling rain Drop Size Distributions (DSDs) is useful for several applications such as: (i) the quantitative estimation of rainfall by weather radars or by mobile telecommunication links; (ii) predicting the attenuation of satellite signals by rain; (iii) washing of atmospheric particles by rain; (iv) soil erosion by rain; etc. This modeling aims to parameterize the rain DSD by a minimum number of scale parameters that can be related to the meteorological variables observed by sensors—for instance radar reflectivity. Inspired by several previous works, Sempere-Torres *et al.* [1994, 1998] proposed a general single-moment normalization to formulate the DSD in terms of a scaling law. Later, based on the work of Testud *et al.* [2001], Lee *et al.* [2004] proposed a general double-moment normalization approach to the DSD. Using convective and stratiform DSD data, they showed that double-moment normalization better captures the natural variability of the DSDs than single-moment normalization. In terms of physical interpretation, the scaling exponent of the single-moment normalization has a marked dependence on the type of rain *i.e.*, with microphysical processes inducing the raindrop formation, whereas in the double-moment normalization distinction between rain type is not necessary [Lee *et al.*, 2004].

The analysis of the impact of DSD integration time step on their modeling began with the work of Chapon *et al.* [2008]. Recently, using single-moment normalization (scale parameter: rain intensity), Moumouni *et al.* [2021] highlighted the dependence of the shape function parameters of DSD models (gamma and lognormal) on the DSD integration time step. By fitting the individual spectra with unimodal DSD models (gamma and lognormal), they also showed that the overall proportion of well-fitted and very well-fitted spectra increases with the integration time step. The need to integrate DSD measurement in time (or space, with a network of disdrometers) is related to the very small sampling area of disdrometers [Tapiador *et al.*, 2017] which leads to sampling error in individual spectra estimation. The amount of integration is a tradeoff between avoiding random errors due to sampling errors and the risk to smooth up too much the part of DSD variability that is related to microphysical processes (such as convective/stratiform).

For polarimetric radar, several authors [Bringi *et al.*, 2003, 2006, Gorgucci *et al.*, 2002, Gosset *et al.*, 2010, Testud *et al.*, 2001] have shown the importance of the double-moment normalization of the DSD. Indeed, the double-moment normalization of DSD provides an excellent model to reduce the DSD variability which strongly affects radar rain retrieval algorithms since this approach is quite effective in collapsing the DSDs around a mean shape. This work aims to analyze the impact of the DSD integration time step on their double-moment normalization. For this purpose, the double-moment normalization of the DSD approach, proposed by Lee *et al.* [2004], will be used and applied to DSD data measured in northern Benin [Moumouni *et al.*, 2008, 2021]. Thus, the following sections will be devoted successively to the datasets, the method, the results and discussions, and the concluding remark.

## 2. Datasets

The DSDs data used for this study are those sampled in northern Benin, near the town of Djougou (9.66°N, 1.69°E)—using optical Disdrometers: single beam [Löffler-Mang and Joss, 2000, Salles, 1995, Salles *et al.*, 1998] and double beam [Delahaye *et al.*, 2005]—from 2005 to 2007, during the African Monsoon Multidisciplinary Analysis (AMMA) meteorological campaign. The climate of the study area is mainly characterized by a unimodal rainfall regime with April to October as period of the unique rainy season. The annual rainfall is about 1200 mm but is marked by a downward trend on the last decades. The DSDs data have been validated and widely used for several studies [Gosset *et al.*, 2010, Kougbéagbédé *et al.*, 2017, Moumouni, 2009, Moumouni *et al.*, 2008, 2018]. These data are DSDs of rains with an integration time step equal to one minute. They are made up of 93 rainy events which represent a total of: 11,647 spectra when only rain intensities greater than or equal to  $0.1 \text{ mm}\cdot\text{h}^{-1}$  are considered; and 12,342 spectra for a threshold of to  $0.05 \text{ mm}\cdot\text{h}^{-1}$ . The selected rainy events are isolated at the level of each Disdrometer. A rainy event is defined as event with duration at least equal to 15 min and intermittency less than 30 min. The duration of the selected rainy events varies from 15 to 527 min.

From these data, Moumouni *et al.* [2021] computed other rain DSDs datasets of various integration time steps (Table 1). The DSDs duration  $T = L$

**Table 1.** Datasets from a few integration time steps

Data	DATA1	DATA2	DATA3	DATA4	DATA5	DATA6
Integration time steps	$T = 1 \text{ min}$	$T = 2 \text{ min}$	$T = 5 \text{ min}$	$T = 10 \text{ min}$	$T = 15 \text{ min}$	$T = 20 \text{ min}$
Number of spectra	12,342	6152	2437	1198	777	578
Cumulative (mm)	1237.16	1236.89	1235.53	1225.94	1217.20	1216.79

min ( $L = 1, 2, 3, 4, \dots, 20$ ) were calculated within each event, and the remaining spectra whose number is less than  $L$  are not taken into account.

### 3. Methods

Since the pioneering work of Marshall and Palmer [1948] on the rain Drop Size Distribution (DSDs), most of followed studies [Atlas et al., 1999, Cerro et al., 1997, Chen et al., 2019, Tenorio et al., 2012, Tokay and Short, 1996, Uijlenhoet et al., 2003, Ulbrich, 1983, Ulbrich and Atlas, 1998, Wen et al., 2018, Willis, 1984, Zeng et al., 2019, Zhang et al., 2003, etc.] analysed it with  $N(D)$  function defined by spectrum of a given period  $T$  (generally 1 min is used). The diameter of drop named equivalent diameter is the diameter of the sphere having the same volume as that of the measured drop.  $N(D)$  corresponds to the number of raindrops per unit of volume and by interval of diameters and is calculated as follows:

$$N(D_r) = \frac{N_r}{ST\Delta D_r V(D_r)} \quad (1)$$

where  $r$  is the index of the class of equivalent diameter,  $\Delta D_r$  the width of the range of equivalent diameter centred on  $D_r$ . In this study,  $D_r$  and  $\Delta D_r$  are expressed in millimeters.  $S$  is the disdrometer collecting surface area expressed in square meter. For the sampling period  $T$ ,  $N_r$  is the number of drops counted by the disdrometer in each size range.  $V(D_r)$  is the falling speed of the drops of diameter  $D_r$ . For the rain intensity to be proportional to a moment of  $N(D)$  function, the relationship between the drops speed and their diameter is used as suggested by Atlas and Ulbrich [1977]:

$$V(D_r) = 3.78 D_r^{0.67} \text{ (m}\cdot\text{s}^{-1}\text{)}. \quad (2)$$

In all the above studies,  $T = 1 \text{ min} = 60 \text{ s}$  and  $N(D_r)$  is expressed in  $(\text{m}^{-3}\cdot\text{mm}^{-1})$ .

The measured moment of order  $n$  of a rain DSD spectrum (whatever the integration time steps) is defined [Chapon et al., 2008, Lee et al., 2004, Ochou

et al., 2007, Sempere-Torres et al., 1994, Zeng et al., 2019] by

$$M_{\text{mea},n} = \sum_r D_r^n N(D_r) \Delta D_r \quad \text{with } n \geq 0 \text{ and } n \neq \infty. \quad (3)$$

When we assume a DSD model, the theoretical moment of order  $n$  of a rain DSD spectrum (whatever the integration time steps) is defined [Chapon et al., 2008, Lee et al., 2004, Ochou et al., 2007, Sempere-Torres et al., 1994, Zeng et al., 2019] by

$$M_n = \int_0^{+\infty} D^n N(D) dD \quad \text{with } n \geq 0 \text{ and } n \neq \infty. \quad (4)$$

#### 3.1. Double-moment normalization [Lee et al., 2004]

There are several variants of DSD normalization that can be summarized in two approaches: single-moment normalization and double-moment normalization. The single-moment normalization method was generalized by Sempere-Torres et al. [1994]. The double-moment approach was generalized by Lee et al. [2004]. The latter proved that the DSD can be normalized by any two of its reference moments ( $M_i$  and  $M_j$ ,  $i \neq j$ ). This is how they linked the function  $N(D)$  to a the normalizing function which we named the shape function  $F(x)$  and to two parameters  $D_c$  and  $N_c$  (with  $x = D/D_c$ ), such that

$$N(D) = N_c F\left(\frac{D}{D_c}\right). \quad (5)$$

The parameters  $D_c$  and  $N_c$  are defined according to the moments  $M_i$  and  $M_j$  (with  $i \neq j$ )

$$D_c = \left(\frac{M_j}{M_i}\right)^{\frac{1}{j-i}} \quad (6)$$

and

$$N_c = M_i^{\frac{j+1}{j-i}} M_j^{-\frac{i+1}{j-i}}. \quad (7)$$

When  $N(D)$  is expressed in  $(\text{m}^{-3}\cdot\text{mm}^{-1})$ , we note that the moments of order  $n$   $M_{\text{mea},n}$  or  $M_n$  are expressed in  $(\text{m}^{-3}\cdot\text{mm}^n)$  and therefore the parameters  $D_c$  and  $N_c$  are expressed respectively in (mm)

and ( $\text{m}^{-3} \cdot \text{mm}^{-1}$ ). Thus, the parameters  $D_c$  and  $N_c$  stand for the average size and the concentration of the drops in the DSD spectrum, respectively. When we introduce the expression (5) in (4), we get

$$M_n = C_n N_c D_c^{n+1} \quad (8)$$

with

$$C_n = \int_0^{+\infty} x^n F(x) dx. \quad (9)$$

Substituting (6) and (7) in (8) we obtained

$$M_n = C_n M_i^{\frac{j-n}{j-i}} M_j^{-\frac{i-n}{j-i}}. \quad (10)$$

The self-consistency of the constant  $C_n$  are obtained by successively setting  $n = i$  and  $n = j$  in expression (10). Thus, this leads to  $C_i = 1$  and  $C_j = 1$ .

The choice of the two reference moments depends on the use of DSD modeling. In the following, to facilitate the understanding of this modeling, we have chosen two reference moments of successive order by posing  $i = k$  and  $j = k + 1$ , with  $k \geq 0$  and  $k \neq \infty$ . Thus, the expressions (6), (7) and (10) become

$$D_c = \frac{M_{k+1}}{M_k} \quad (11)$$

$$N_c = \frac{M_k^{k+2}}{M_{k+1}^{k+1}} \quad (12)$$

and

$$M_n = C_n M_k^{k-n+1} M_{k+1}^{n-k}. \quad (13)$$

The self-consistency of the constant  $C_n$  corresponds in this case to:  $C_k = 1$  and  $C_{k+1} = 1$ .

### 3.2. Case of the gamma DSD model

The gamma DSD model proposed by Tokay and Short [1996] or Maki *et al.* [2001] is written

$$N(D) = N_G D^\mu \exp(-\Lambda_G D) \quad (14)$$

$N_G$ ,  $\Lambda_G$  and  $\mu$  are the model parameters:  $N_G$  is the intercept parameter;  $\Lambda_G$  is the slope parameter; and  $\mu$  is the shape parameter. Its theoretical moment of order  $n$ , calculated with formula (4) is

$$M_n = N_G \frac{\Gamma(\mu + n + 1)}{\Lambda_G^{(\mu + n + 1)}}. \quad (15)$$

In this expression,  $\Gamma$  is the mathematical function gamma. By calculating the moments  $M_k$  and  $M_{k+1}$  with expression (15) and replacing them in expressions (11) and (12), we find

$$\Lambda_G = \frac{\mu + k + 1}{D_c} \quad (16)$$

and

$$N_G = \frac{N_c (\mu + k + 1)^{(\mu + k + 1)}}{D_c^\mu \Gamma(\mu + k + 1)}. \quad (17)$$

Substituting (16) and (17) into (14), and taking into account the expression (5), one obtains the expression the shape function of the gamma DSD model

$$F_G(x; \mu) = \frac{(\mu + k + 1)^{(\mu + k + 1)}}{\Gamma(\mu + k + 1)} x^\mu \exp[-(\mu + k + 1)x]. \quad (18)$$

By calculating (9) with the expression (18), this leads to the following relationship.

$$C_{G,n} = \frac{(\mu + k + 1)^{(k-n)} \Gamma(\mu + n + 1)}{\Gamma(\mu + k + 1)}. \quad (19)$$

### 3.3. Case of the generalized gamma DSD model

Similarly to the gamma DSD model (14), the generalized gamma DSD model can be written in this form

$$N(D) = N_{GG} D^\nu \exp(-\Lambda_{GG} D^\varepsilon) \quad (20)$$

$N_{GG}$ ,  $\Lambda_{GG}$ ,  $\nu$  and  $\varepsilon$  are the model parameters:  $N_{GG}$  is the intercept parameter;  $\Lambda_{GG}$  is the slope parameter; and  $\nu$  and  $\varepsilon$  are the shape parameters. Its theoretical moment of order  $n$ , calculated with formula (4) is

$$M_n = N_{GG} \frac{\Gamma[(\nu + n + 1)/\varepsilon]}{\varepsilon \Lambda_{GG}^{[(\nu + n + 1)/\varepsilon]}}. \quad (21)$$

By calculating the moments  $M_k$  and  $M_{k+1}$  with expression (21) and replacing them in expressions (11) and (12), we find

$$\Lambda_{GG} = \left\{ \frac{\Gamma[(\nu + k + 2)/\varepsilon]}{D_c \Gamma[(\nu + k + 1)/\varepsilon]} \right\}^\varepsilon \quad (22)$$

and

$$N_{GG} = \frac{\varepsilon N_c}{D_c^\nu \Gamma[(\nu + k + 1)/\varepsilon]} \left\{ \frac{\Gamma[(\nu + k + 2)/\varepsilon]}{\Gamma[(\nu + k + 1)/\varepsilon]} \right\}^{(\nu + k + 1)}. \quad (23)$$

Substituting (22) and (23) into (20), and taking into account the expression (5), leads to express the shape function of the generalized gamma DSD model as:

$$F_{GG}(x; \nu, \varepsilon) = \frac{\varepsilon}{\Gamma[(\nu + k + 1)/\varepsilon]} \times \left\{ \frac{\Gamma[(\nu + k + 2)/\varepsilon]}{\Gamma[(\nu + k + 1)/\varepsilon]} \right\}^{(\nu + k + 1)} \times x^\nu \exp \left\{ - \left[ \frac{\Gamma[(\nu + k + 2)/\varepsilon]}{\Gamma[(\nu + k + 1)/\varepsilon]} \right]^\varepsilon x^\varepsilon \right\}. \quad (24)$$

By calculating (9) with the expression (24), one obtains

$$C_{GG,n} = \left\{ \frac{\Gamma[(\nu + k + 2)/\varepsilon]}{\Gamma[(\nu + k + 1)/\varepsilon]} \right\}^{(k-n)} \frac{\Gamma[(\nu + n + 1)/\varepsilon]}{\Gamma[(\nu + k + 1)/\varepsilon]}. \quad (25)$$

- *Comparison with the generalized gamma of Lee et al. [2004]*

The generalized gamma DSD model proposed by Lee et al. [2004] is written as follows:

$$N(D) = M_0 \frac{c\lambda}{\Gamma(\mu)} (\lambda D)^{c\mu-1} \exp(-(\lambda D)^c). \quad (26)$$

By comparing this expression to the expression (20), we notice that  $c = \varepsilon$  and  $\mu = (\nu + 1)/c$ . By replacing these relations in the expression of the constant  $C_{GG,2,n}$  established by Lee et al. [2004] and setting  $i = k$  and  $j = k + 1$ , we find the expression of the constant  $C_{GG,n}$  (25). So models (20) and (26) are similar.

### 3.4. Case of the lognormal DSD model

The lognormal DSD model proposed by Feingold and Levin [1986] is written

$$N(D) = \frac{N_T}{\sqrt{2\pi}D \ln \sigma} \exp \left[ -\frac{\ln^2(D/D_g)}{2 \ln^2 \sigma} \right] \quad (27)$$

$N_T$ ,  $D_g$  and  $\sigma$  are the parameters of the model:  $N_T$  the total drops number per unit volume ( $N_T = M_0$ );  $D_g$  is the geometric mean of drops diameter ( $\ln D_g = \frac{\ln D}{\ln \sigma}$ ); and  $\sigma$  geometric standard deviation ( $\ln^2 \sigma = \frac{(\ln D - \ln D_g)^2}{\ln^2 D}$ ).  $\ln$  is natural logarithm. Its theoretical moment of order  $n$ , calculated with formula (4) is

$$M_n = N_T D_g^n \exp \left( \frac{1}{2} n^2 \ln^2 \sigma \right). \quad (28)$$

By calculating the moments  $M_k$  and  $M_{k+1}$  with expression (28) and replacing them in expressions (11) and (12), we find

$$D_g = D_c \exp[-0.5(2k+1) \ln^2 \sigma] \quad (29)$$

and

$$N_T = N_c D_c \exp[0.5k(k+1) \ln^2 \sigma]. \quad (30)$$

When we introduce (29) and (30) into (27) and take into account the expression (5), we obtain the expression the shape function of the lognormal DSD model

$$F_L(x; \sigma) = \frac{\exp[0.5k(k+1) \ln^2 \sigma]}{\sqrt{2\pi}x \ln \sigma} \times \exp \left\{ -\frac{[\ln x + 0.5(2k+1) \ln^2 \sigma]^2}{2 \ln^2 \sigma} \right\}. \quad (31)$$

By calculating (9) with the expression (31), we obtain

$$C_{L,n} = \exp \{ 0.5[k^2 + n^2 + k - n(2k+1)] \ln^2 \sigma \}. \quad (32)$$

### 3.5. Implementation of double-moment normalization

For each rain DSD integration time step, the double-moment normalization is implemented successively as follows:

- (1) Calculation of the sample moments of order  $n$  with (3).
- (2) Choice of two reference moments of order  $k$  and  $k + 1$ , then the calculation of the parameters  $D_c$  and  $N_c$  and of the average sample of  $C_n$  with the relation (8).
- (3) Estimation of the parameters of each shape function. The constant  $C_n$  being used at the end of the modeling to estimate the moments, these sample will be used to estimate the parameters of the shape functions.
  - (3.1) For the gamma DSD model, the parameter  $\mu$  is estimated by making a least-squares adjustment by comparing  $C_{G,n}$  (19) with the sample of  $C_n$  calculated in the previous step.
  - (3.2) For the generalized gamma DSD model, the parameters  $\nu$  and  $\varepsilon$  is estimated by making a least-squares adjustment by comparing  $C_{GG,n}$  (25) with the sample of  $C_n$  calculated in the previous step.
  - (3.3) For the lognormal DSD model, the parameter  $\sigma$  is estimated by making a least-squares adjustment by comparing  $C_{L,n}$  (32) with the sample of  $C_n$  calculated in the previous step.
- (4) Calculation of the normalized spectra with the expression (5) then representation of these spectra and of the shape functions (gamma, generalized gamma and lognormal).
- (5) Model evaluation: comparison of the estimated moments  $M_{\text{est},n}$  (8) with the measured moments  $M_{\text{mea},n}$  (3). In expression (8), the constant  $C_n$  is estimated for each model from the estimated values of the parameters of the shape functions:  $C_{G,n}$  for the gamma model;  $C_{GG,n}$  for the generalized gamma model; and  $C_{L,n}$  for the lognormal model. For the comparison, the following statistical criteria were used:

- standard deviation of fractional error:

$$\text{SDFE} = \sqrt{E \left[ \left( \frac{M_{\text{est},n} - M_{\text{mea},n}}{M_{\text{mea},n}} \right)^2 \right]} \quad (33)$$

- mean relative error:

$$\text{MRE} = E \left[ \left( \frac{M_{\text{est},n} - M_{\text{mea},n}}{M_{\text{mea},n}} \right) \right]. \quad (34)$$

In these expressions,  $E[\cdot]$  denotes the sample mean.

## 4. Results and discussion

### 4.1. Results

The rain DSDs observed in northern Benin are modeled using the method above described. Data from the six integration time steps (Table 1) are used separately. For each integration time step, six different couples of moments ( $M_k, M_{k+1}$ ) are used for normalization. The statistics of the parameters  $D_c$  and  $N_c$ , for each integration time step and for each couple of moments, are presented in Table A1 of the appendix. The average values of the constant  $C_n$ , for each integration time step and for each couple of moments, are presented in Table A2 of the appendix. The variation of these statistics between the couples of moments is obvious from expressions (11) and (12). Nevertheless, for a couple of moments, the variation of these statistics as a function of the integration time step is not obvious. This raises several questions: (i) are these statistics significantly different? (ii) if yes, what is this due to? These questions deserve to be addressed in our future work.

For the couple of moments ( $M_3, M_4$ ), the result of the adjustment of the constant  $C_n$ , for each model, is represented in Figure 1. For the same couple of moments, the normalized spectra and the shape functions are represented in Figure 2. These figures show that each considered model fits well the DSDs observed in northern Benin. Figure 2 shows that: (i) the gamma and generalized gamma models seem to better model the mean distribution in the left tail; (ii) the three models seem to model the mean distribution in the same way in the intermediate region; (iii) the lognormal model seem to better model the mean distribution in the right tail. This description deserves to be deepened by first clarifying the importance of this mean distribution for the DSD modeling. Indeed, in this study, we adjusted the mean of the constant  $C_n$  instead of the mean distribution, because it is very

useful for the estimation of the other moments from the reference moments (relation 10 and 13). Figure 1 shows that the constant  $C_n$  is better fitted by the generalized gamma model.

The importance of a DSD model being its ability to estimate from the reference moments its other moments, the three proposed models were evaluated. The result of this evaluation for the couple of moments ( $M_3, M_4$ ) is shown in Figure 3. This result for the other couples of moments is shown in Figures A1 to A5, in the appendix. The SDFE indicates, for each couple of moments ( $M_k, M_{k+1}$ ), whatever the integration time step, that the three models estimate the moments of order between  $k - 0.5$  and  $k + 1.5$  with almost null error. Nevertheless, this error becomes very significant outside this interval. The MRE indicates, for all couples of moments, that the estimate with the generalized gamma model is better than that of the other two models. Moreover, with the generalized gamma model, the MRE is almost independent of the integration time step, contrary to the case of the other two models.

The impact of the integration time step on the modeling is analyzed. The estimated parameters, for each integration time step and for each couple of moments, are presented in Table 2. For the couple of moments ( $M_3, M_4$ ), these parameters are represented as a function of the integration time step in Figure 4. The relationships established for the six couples of moments are presented in Table 3. The parameters of the shape functions are therefore dependent on the integration time step, whatever the model and the pair of moments.

### 4.2. Discussion

In West Africa, gamma and lognormal DSD models are used more to adjust measured DSDs [Nzeukou *et al.*, 2004, Ochou *et al.*, 2007, Sauvageot and La-caux, 1995, Moumouni *et al.*, 2008, 2021]. This study showed that these models are well adapted for the analyzed DSDs, but the generalized gamma DSD model is better because of its flexibility [Lee *et al.*, 2004].

The choice of reference moments for DSD normalization must be made according to the applications. In this article, we have analyzed the case of consecutive moment order, because it allows us to simplify the equations. The expressions of the shape

**Table 2.** Parameters of the shape functions of the models. They are estimated for each integration time step and for each couple of moments

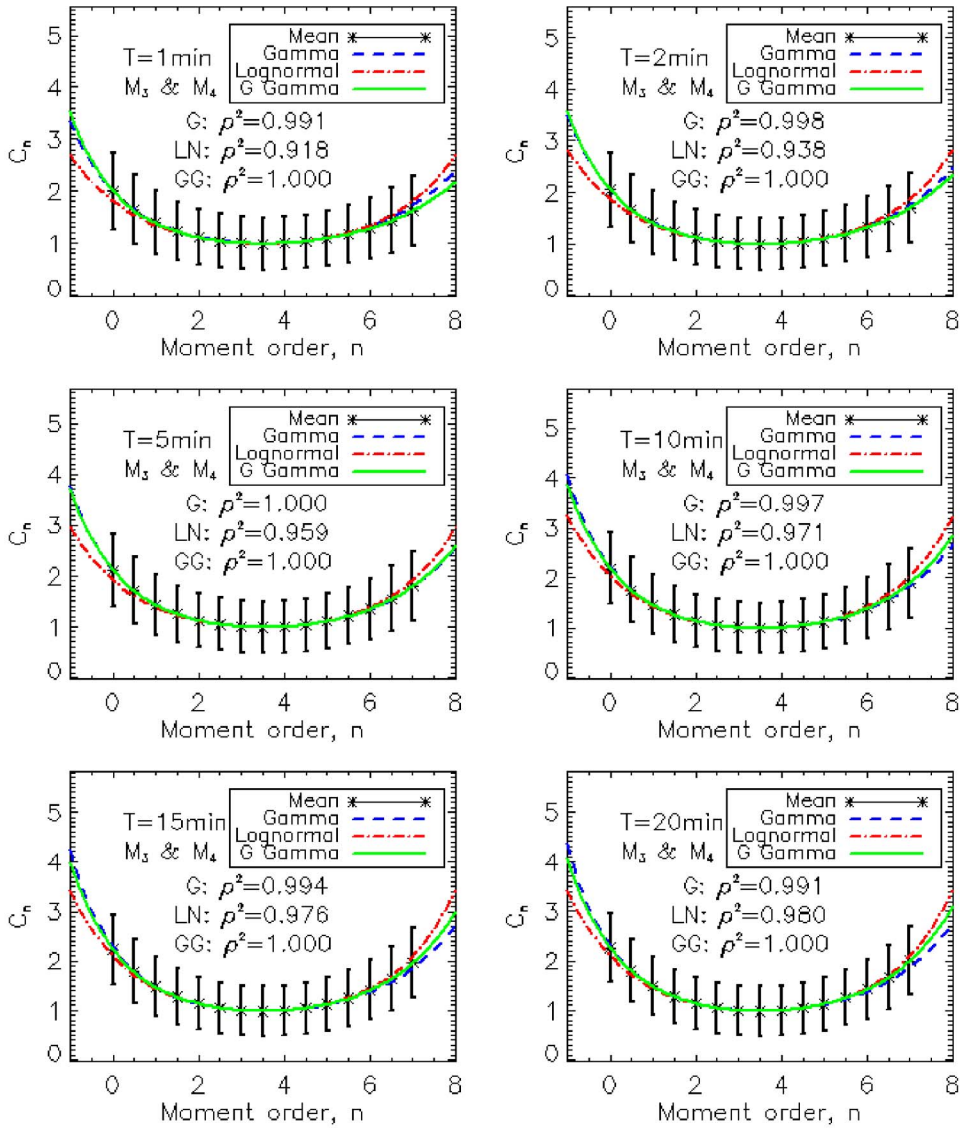
$T$ (min)	$(M_0, M_1)$				$(M_1, M_2)$				$(M_2, M_3)$			
	$\mu$	$\nu$	$\varepsilon$	$\sigma$	$\mu$	$\nu$	$\varepsilon$	$\sigma$	$\mu$	$\nu$	$\varepsilon$	$\sigma$
1	6.040	3.520	1.470	1.400	6.190	2.650	1.680	1.380	6.550	3.680	1.400	1.360
2	5.750	4.360	1.220	1.410	5.780	3.620	1.340	1.390	5.920	4.610	1.160	1.370
3	5.560	5.000	1.080	1.410	5.510	4.250	1.180	1.390	5.550	5.290	1.030	1.380
4	5.440	5.440	1.000	1.420	5.360	5.040	1.040	1.400	5.340	5.620	0.970	1.380
5	5.350	5.910	0.930	1.420	5.230	5.480	0.970	1.400	5.150	6.140	0.900	1.390
6	5.250	6.460	0.860	1.420	5.090	5.790	0.920	1.410	4.970	6.630	0.840	1.390
7	5.150	6.660	0.830	1.430	4.980	6.390	0.850	1.410	4.850	6.740	0.820	1.390
8	5.100	7.050	0.790	1.430	4.920	6.780	0.810	1.410	4.750	7.160	0.780	1.400
9	5.070	7.380	0.760	1.430	4.860	7.090	0.780	1.410	4.670	7.640	0.740	1.400
10	4.960	7.630	0.730	1.430	4.720	7.460	0.740	1.420	4.510	7.750	0.720	1.400
11	4.910	7.990	0.700	1.430	4.670	7.990	0.700	1.420	4.440	7.990	0.700	1.410
12	4.840	8.190	0.680	1.440	4.580	8.190	0.680	1.420	4.340	8.190	0.680	1.410
13	4.830	7.730	0.710	1.440	4.590	7.710	0.710	1.420	4.360	7.710	0.710	1.410
14	4.780	8.430	0.660	1.440	4.520	8.630	0.650	1.420	4.270	8.450	0.660	1.410
15	4.720	8.680	0.640	1.440	4.450	8.890	0.630	1.430	4.180	8.700	0.640	1.410
16	4.700	9.010	0.620	1.440	4.430	9.250	0.610	1.430	4.160	8.660	0.640	1.410
17	4.630	9.070	0.610	1.440	4.340	9.310	0.600	1.430	4.050	8.900	0.620	1.420
18	4.620	9.250	0.600	1.450	4.330	9.730	0.580	1.430	4.030	9.070	0.610	1.420
19	4.550	9.980	0.560	1.450	4.230	10.530	0.540	1.430	3.900	10.030	0.560	1.420
20	4.590	9.610	0.580	1.450	4.290	10.140	0.560	1.430	3.990	9.230	0.600	1.420

$T$ (min)	$(M_3, M_4)$				$(M_4, M_5)$				$(M_5, M_6)$			
	$\mu$	$\nu$	$\varepsilon$	$\sigma$	$\mu$	$\nu$	$\varepsilon$	$\sigma$	$\mu$	$\nu$	$\varepsilon$	$\sigma$
1	5.930	3.050	1.570	1.370	6.170	3.920	1.340	1.380	6.600	3.500	1.460	1.350
2	5.650	3.990	1.270	1.380	5.750	4.900	1.110	1.390	5.970	4.480	1.190	1.370
3	5.480	4.640	1.120	1.380	5.490	5.490	1.000	1.390	5.610	5.020	1.070	1.380
4	5.380	5.200	1.020	1.390	5.330	5.960	0.930	1.400	5.380	5.470	0.990	1.380
5	5.220	5.750	0.940	1.390	5.190	6.320	0.880	1.400	5.190	5.960	0.920	1.390
6	5.110	6.300	0.870	1.400	5.060	6.720	0.830	1.410	5.010	6.530	0.850	1.390
7	5.030	6.620	0.830	1.400	4.950	6.960	0.800	1.410	4.880	6.750	0.820	1.390
8	4.990	7.160	0.780	1.400	4.890	7.410	0.760	1.410	4.770	7.150	0.780	1.400
9	4.940	7.500	0.750	1.400	4.830	7.620	0.740	1.410	4.700	7.640	0.740	1.400
10	4.810	7.750	0.720	1.410	4.690	7.890	0.710	1.420	4.550	7.750	0.720	1.400
11	4.780	8.320	0.680	1.410	4.640	8.140	0.690	1.420	4.460	8.300	0.680	1.410
12	4.700	8.360	0.670	1.410	4.560	8.350	0.670	1.420	4.370	8.530	0.660	1.410
13	4.690	8.030	0.690	1.410	4.560	7.860	0.700	1.420	4.390	8.200	0.680	1.410
14	4.650	8.810	0.640	1.410	4.500	8.620	0.650	1.420	4.300	8.610	0.650	1.410
15	4.580	8.880	0.630	1.420	4.430	8.890	0.630	1.420	4.210	8.870	0.630	1.410
16	4.580	9.460	0.600	1.420	4.420	8.680	0.640	1.420	4.170	9.010	0.620	1.410
17	4.490	9.520	0.590	1.420	4.310	9.100	0.610	1.430	4.070	9.280	0.600	1.420
18	4.500	9.950	0.570	1.420	4.310	9.520	0.590	1.430	4.040	9.680	0.580	1.420
19	4.410	10.520	0.540	1.420	4.220	9.820	0.570	1.430	3.930	10.220	0.550	1.420
20	4.470	10.120	0.560	1.420	4.290	9.260	0.600	1.430	4.010	9.860	0.570	1.420

 $\mu$ : gamma,  $(\nu, \varepsilon)$ : generalized gamma,  $\sigma$ : lognormal.

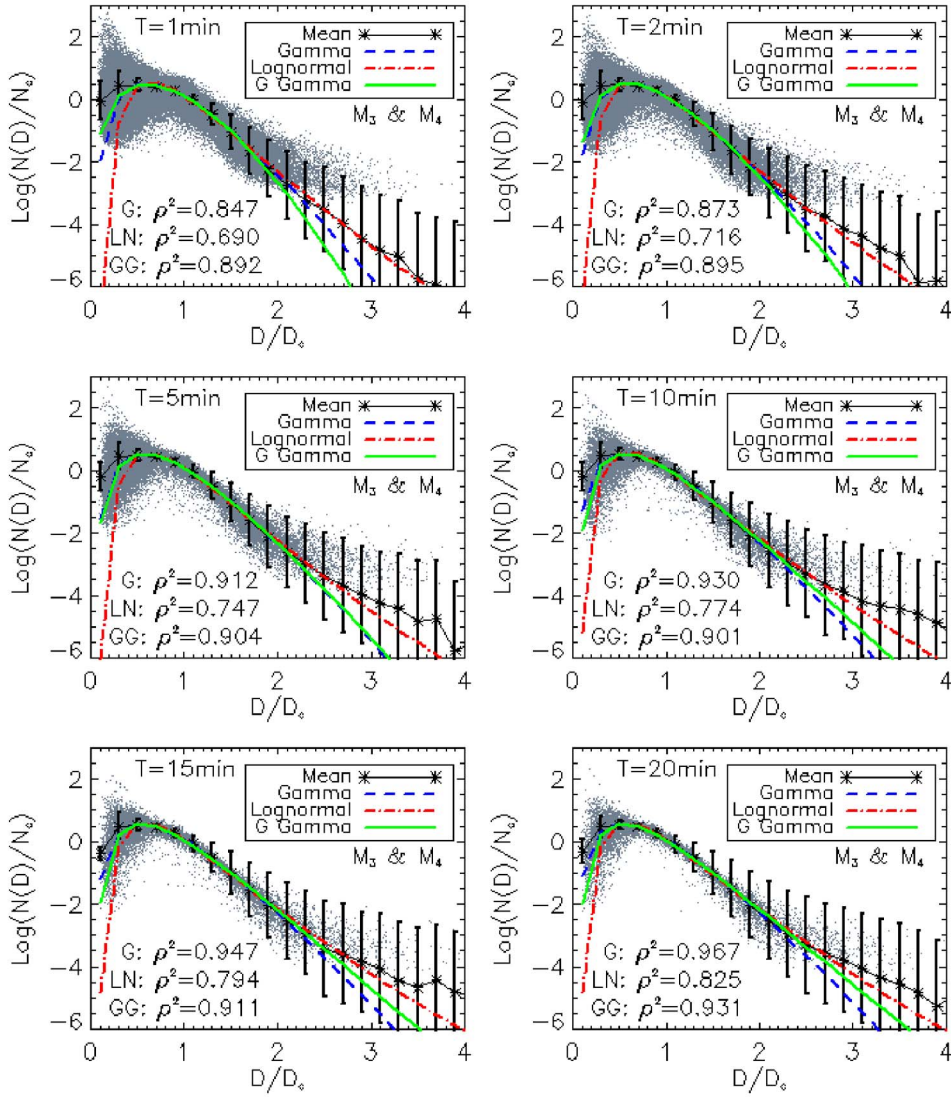




**Figure 1.** Fit of the constant  $C_n$  for each model (G is gamma: blue, GG is generalized gamma (noted G Gamma): green and LN is lognormal: red). The asterisk represents the sample mean and the error bar represents the standard deviation.  $\rho^2$  is the square of the linear Pearson correlation coefficient, between the sample mean and the model. Case of the couple of moments ( $M_3$ ,  $M_4$ ).

function or the constant  $C_n$ , established for the three DSD models used, will be used for any application requiring the use of consecutive moment order. For other combinations of moments, one can refer to Lee et al. [2004]. For retrieval of rain intensity from radar measurements, these authors suggest the combination of moments  $M_6$  (close to the reflectivity, variable measured by the radar) and  $M_i$  ( $i$  less than 6).

Weather radar measurements are usually based on a sequence of quasi-horizontal scans (PPIs, for Plan Position Indicators) each providing a quasi-instantaneous snapshot of the rain, with a typical sampling rate of 1 to 5 min. Depending on the overall scanning strategy, several subsequent PPIs may be used to provide a time averaged rainrate every 10 to 15 min. The parameters of the rainfall estimation

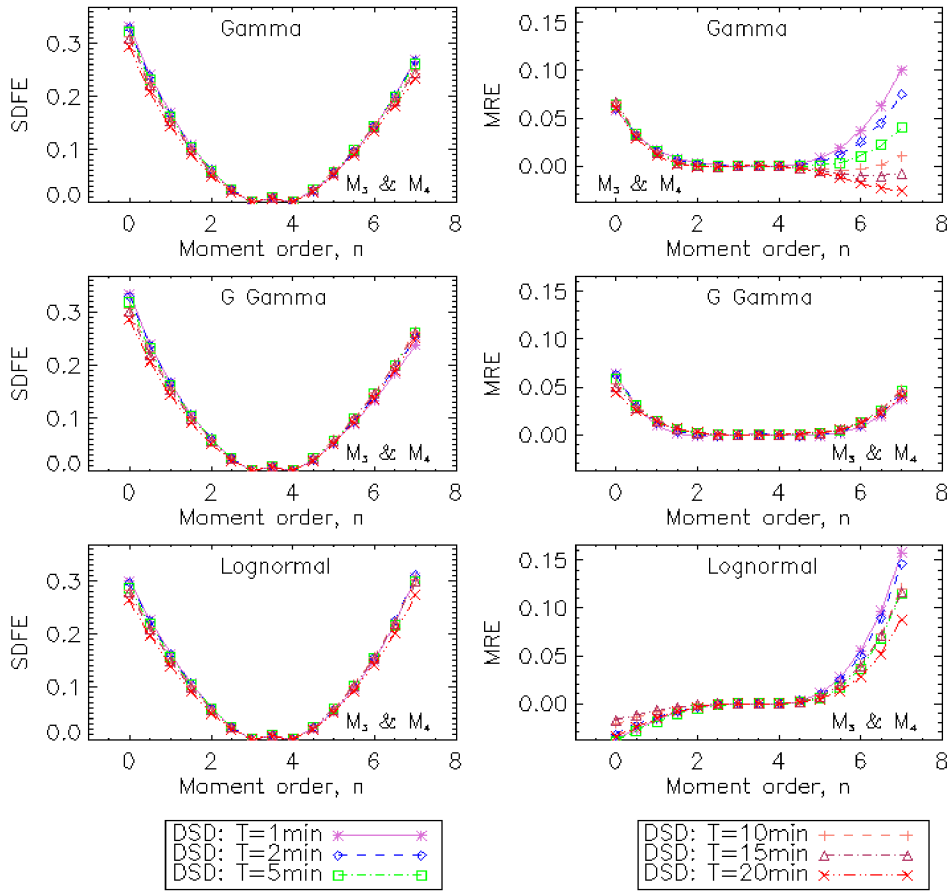


**Figure 2.** Normalized spectra (slate gray dots) and the shape functions of the models (G is gamma: blue, GG is generalized gamma (noted G Gamma): green and LN is lognormal: red). The asterisk represents the sample mean and the error bar represents the absolute value of standard deviation.  $\rho^2$  is the square of the linear Pearson correlation coefficient, between the sample mean and the model. Case of the couple of moments ( $M_3$ ,  $M_4$ ).

algorithms or of DSD model are often derived from DSDs sampled at ground level, at 1 min time step. The results of this study can be used to transform these parameters to the time step most relevant for the application.

This article shows trends in the evolution of the shape parameters with respect to the DSD integration time step, in West Africa, as shown by Moumouni

et al. [2021] with the single-moment normalization method. Unfortunately, to our knowledge, no reference relating to the study of the impact of the DSD integration time step on their double-moment normalization is available to make a comparison with. Meanwhile, it opens up other scientific questions: (i) do the trends in parameters values change from one climatic region to another? (ii) what is the



**Figure 3.** Comparison of moments estimated to moments measured with SDFE and MRE. Case of the couple of moments ( $M_3$ ,  $M_4$ ), relating to the DSD models gamma, generalized gamma (noted G Gamma) and lognormal.

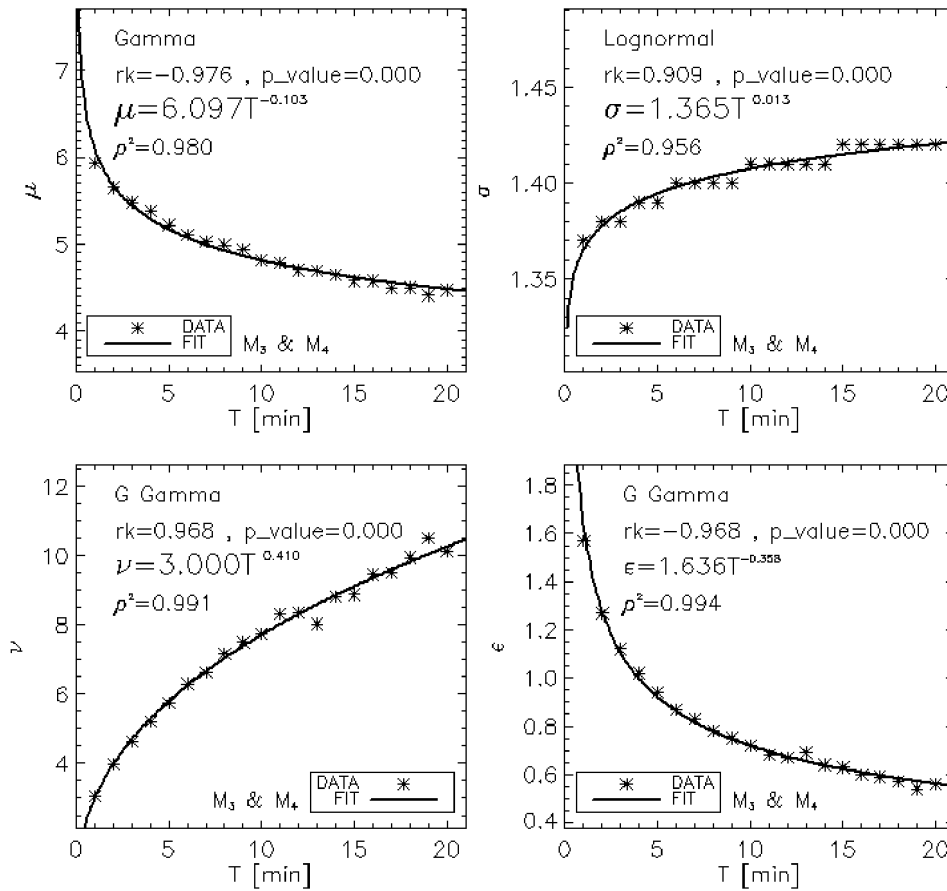
physical significance of these changes? or more broadly, what is the cause of the variability of the shape parameters during the rainfall event? etc. All these important scientific questions for the modeling of the DSD, can be the subject of other works. The gamma, generalized gamma and lognormal models doesn't have the same heavy tails [Halliwell, 2013] as shown in Figure 2. For future work, we will consider this aspect, as well as the fitting of other models (Pareto and inverse gamma) by other methods [Halliwell, 2013].

## 5. Concluding remark

Following the work of Moumouni et al. [2021] on the analysis of the impact of the integration time step

on the single-moment normalization of the DSD, this study analyzes the impact of the integration time step on the double-moment normalization of the DSD. The method proposed by Lee et al. [2004] was used. It was applied to rain DSD data measured in northern Benin (West Africa) during AMMA intensive period experiment. DSD models gamma, generalized gamma, and lognormal are used for spectra fitting. The modeling was evaluated by comparing the estimated moments to the measured moments.

The study revealed, as in the case of single-moment normalization [Moumouni et al., 2021], a strong dependence of the shape function parameters on the integration time step. It also showed that the gamma and lognormal DSD models are well adapted to the data used, but the generalized gamma DSD



**Figure 4.** Link between shape function parameters and integration time step.  $rk$  is the Kendall rank correlation between sample and integration time step.  $p\_value$  is the significance level of the correlation. It's in the interval  $[0.0, 1.0]$ ; a small value indicates a significant correlation. Case of the couple of moments  $(M_3, M_4)$ .  $\rho^2$  is the square of the linear Pearson correlation coefficient, between the sample and the model. Generalized gamma is noted G Gamma.

model is recommended because the DSD moment estimation errors are reduced and independent of the integration time step.

### Author contributions

Data curation, SM; Formal analysis, SM, FPA, E-PZ; Methodology, SM, FPA, E-PZ; Supervision, SM, AEL, MG; Writing—original draft, SM; Writing—review & editing, SM, AEL, MG.

### Conflicts of interest

The authors declare no conflict of interest.

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**Table 3.** Relations between the shape function parameters and the integration time step ( $T$  (min))

Gamma					Lognormal				
Couples	$rk$	$p\_value$	Relationship	$\rho^2$	$rk$	$p\_value$	Relationship	$\rho^2$	
$(M_0, M_1)$	-0.989	0.000	$\mu = 6.180T^{-0.098}$	0.983	0.912	0.000	$\sigma = 1.396T^{0.012}$	0.936	
$(M_1, M_2)$	-0.979	0.000	$\mu = 6.349T^{-0.129}$	0.987	0.909	0.000	$\sigma = 1.375T^{0.013}$	0.956	
$(M_2, M_3)$	-0.979	0.000	$\mu = 6.714T^{-0.173}$	0.992	0.923	0.000	$\sigma = 1.356T^{0.015}$	0.966	
$(M_3, M_4)$	-0.976	0.000	$\mu = 6.097T^{-0.103}$	0.980	0.909	0.000	$\sigma = 1.365T^{0.013}$	0.956	
$(M_4, M_5)$	-0.984	0.000	$\mu = 6.313T^{-0.129}$	0.989	0.903	0.000	$\sigma = 1.376T^{0.012}$	0.953	
$(M_5, M_6)$	-0.979	0.000	$\mu = 6.778T^{-0.175}$	0.992	0.923	0.000	$\sigma = 1.352T^{0.016}$	0.975	
Generalized gamma									
Couples	$rk$	$p\_value$	Relationship	$\rho^2$	$rk$	$p\_value$	Relationship	$\rho^2$	
$(M_0, M_1)$	0.968	0.000	$\nu = 3.449T^{0.343}$	0.989	-0.968	0.000	$\varepsilon = 1.527T^{-0.321}$	0.994	
$(M_1, M_2)$	0.968	0.000	$\nu = 2.640T^{0.450}$	0.990	-0.968	0.000	$\varepsilon = 1.753T^{-0.377}$	0.994	
$(M_2, M_3)$	0.947	0.000	$\nu = 3.708T^{0.314}$	0.979	-0.966	0.000	$\varepsilon = 1.427T^{-0.294}$	0.994	
$(M_3, M_4)$	0.968	0.000	$\nu = 3.000T^{0.410}$	0.991	-0.968	0.000	$\varepsilon = 1.636T^{-0.358}$	0.994	
$(M_4, M_5)$	0.937	0.000	$\nu = 3.959T^{0.294}$	0.984	-0.947	0.000	$\varepsilon = 1.359T^{-0.280}$	0.995	
$(M_5, M_6)$	0.968	0.000	$\nu = 3.447T^{0.353}$	0.991	-0.976	0.000	$\varepsilon = 1.513T^{-0.325}$	0.994	

They are established for the six couples of moments studied.  $rk$  is the Kendall rank correlation between sample and integration time step.  $p\_value$  is the significance level of the correlation. It's in the interval  $[0.0, 1.0]$ ; a small value indicates a significant correlation.  $\rho^2$  is the square of the linear Pearson correlation coefficient, between the sample and the model.

## Appendix A.

See Tables A1 and A2 and Figures A1–A5.

**Table A1.** Statistics of parameters  $D_c$  and  $N_c$ , for each integration time step ( $T$  (min)) and for each couple of moments

$T$ (min)	$D_c$ (mm)								$\text{Log}(N_c \text{ (m}^{-3} \cdot \text{mm}^{-1}))$						
	Mean	Std.	Skew.	Kurt.	Med.	Min.	Max.	Mean	Std.	Skew.	Kurt.	Med.	Min.	Max.	
$(M_0, M_1)$															
1	1.107	0.263	0.134	0.169	1.113	0.419	2.359	1.938	0.484	-0.299	0.166	1.937	-0.300	3.215	
2	1.108	0.258	0.101	0.135	1.114	0.434	2.267	1.946	0.472	-0.221	-0.071	1.943	-0.093	3.175	
3	1.109	0.255	0.066	0.102	1.117	0.429	2.231	1.952	0.464	-0.187	-0.176	1.946	0.226	3.092	
4	1.111	0.252	0.073	0.142	1.118	0.438	2.206	1.956	0.460	-0.179	-0.190	1.949	0.313	3.082	
5	1.112	0.249	0.038	0.119	1.118	0.450	2.130	1.961	0.455	-0.176	-0.194	1.954	0.296	3.077	
6	1.113	0.248	0.036	0.128	1.117	0.466	2.059	1.965	0.449	-0.145	-0.272	1.958	0.433	3.081	
7	1.115	0.244	0.012	0.106	1.129	0.465	1.997	1.968	0.445	-0.142	-0.296	1.961	0.542	3.068	
8	1.118	0.243	0.027	0.201	1.128	0.469	2.071	1.971	0.443	-0.152	-0.287	1.968	0.509	3.077	
9	1.118	0.243	0.006	0.179	1.131	0.470	2.058	1.978	0.436	-0.120	-0.352	1.969	0.559	3.058	
10	1.119	0.238	-0.022	0.080	1.128	0.472	1.971	1.979	0.436	-0.155	-0.312	1.972	0.683	3.049	
11	1.121	0.238	-0.036	0.106	1.130	0.475	1.975	1.986	0.429	-0.130	-0.327	1.968	0.570	3.051	
12	1.123	0.237	-0.030	0.180	1.133	0.474	1.973	1.990	0.425	-0.113	-0.457	1.990	0.741	3.054	

(continued on next page)

**Table A1.** (continued)

$T$ (min)	$D_c$ (mm)							$\text{Log}(N_c \text{ (m}^{-3}\cdot\text{mm}^{-1}))$						
	Mean	Std.	Skew.	Kurt.	Med.	Min.	Max.	Mean	Std.	Skew.	Kurt.	Med.	Min.	Max.
13	1.125	0.234	-0.019	0.159	1.135	0.474	1.971	1.991	0.424	-0.140	-0.366	1.988	0.635	3.055
14	1.126	0.235	-0.030	0.158	1.130	0.473	1.965	1.992	0.421	-0.143	-0.344	1.984	0.604	3.055
15	1.127	0.235	-0.046	0.202	1.131	0.474	1.959	2.002	0.414	-0.112	-0.450	1.992	0.715	3.058
16	1.129	0.233	-0.059	0.214	1.145	0.474	1.953	1.996	0.420	-0.174	-0.362	2.005	0.610	3.058
17	1.130	0.230	-0.041	0.234	1.140	0.474	1.951	2.004	0.413	-0.133	-0.412	2.009	0.741	3.054
18	1.132	0.228	-0.026	0.270	1.144	0.474	1.956	2.003	0.413	-0.141	-0.406	2.009	0.774	3.050
19	1.132	0.228	-0.044	0.208	1.134	0.475	1.951	2.004	0.408	-0.126	-0.469	2.010	0.860	3.027
20	1.136	0.227	-0.030	0.239	1.150	0.475	1.945	2.010	0.403	-0.093	-0.548	1.999	0.864	2.996
$(M_1, M_2)$														
1	1.274	0.322	0.288	0.642	1.277	0.426	3.082	1.819	0.480	-0.312	0.122	1.820	-0.303	3.149
2	1.277	0.317	0.255	0.576	1.285	0.449	2.821	1.825	0.468	-0.242	-0.091	1.823	-0.172	3.072
3	1.281	0.312	0.199	0.403	1.288	0.445	2.676	1.830	0.461	-0.211	-0.200	1.834	0.150	2.996
4	1.285	0.308	0.192	0.458	1.293	0.452	2.587	1.832	0.456	-0.208	-0.205	1.831	0.190	2.978
5	1.287	0.304	0.130	0.307	1.298	0.477	2.519	1.837	0.450	-0.201	-0.220	1.838	0.154	2.988
6	1.289	0.301	0.124	0.337	1.300	0.483	2.488	1.840	0.445	-0.172	-0.297	1.843	0.284	2.987
7	1.293	0.298	0.099	0.329	1.301	0.483	2.447	1.842	0.440	-0.174	-0.313	1.838	0.376	2.961
8	1.297	0.296	0.097	0.355	1.304	0.491	2.423	1.845	0.438	-0.185	-0.306	1.848	0.391	2.972
9	1.297	0.294	0.062	0.350	1.310	0.491	2.386	1.851	0.431	-0.153	-0.360	1.846	0.461	2.935
10	1.299	0.289	0.071	0.354	1.306	0.494	2.368	1.851	0.429	-0.184	-0.340	1.850	0.571	2.941
11	1.304	0.289	0.031	0.331	1.316	0.493	2.359	1.857	0.423	-0.181	-0.319	1.851	0.423	2.945
12	1.307	0.287	0.036	0.431	1.316	0.491	2.348	1.860	0.418	-0.162	-0.450	1.869	0.605	2.946
13	1.310	0.284	0.053	0.421	1.308	0.491	2.339	1.861	0.416	-0.184	-0.342	1.853	0.552	2.946
14	1.311	0.284	0.021	0.381	1.319	0.492	2.341	1.861	0.413	-0.174	-0.345	1.861	0.502	2.946
15	1.314	0.284	-0.006	0.430	1.323	0.493	2.346	1.870	0.406	-0.152	-0.415	1.873	0.578	2.949
16	1.317	0.281	-0.033	0.451	1.323	0.497	2.339	1.864	0.413	-0.213	-0.338	1.870	0.516	2.949
17	1.319	0.278	-0.008	0.460	1.325	0.498	2.331	1.871	0.404	-0.161	-0.392	1.873	0.627	2.944
18	1.321	0.274	-0.025	0.555	1.330	0.498	2.325	1.870	0.404	-0.170	-0.376	1.864	0.632	2.924
19	1.322	0.273	-0.006	0.467	1.329	0.499	2.317	1.871	0.399	-0.166	-0.439	1.884	0.763	2.849
20	1.326	0.273	0.003	0.490	1.339	0.499	2.302	1.876	0.393	-0.141	-0.516	1.877	0.774	2.779
$(M_2, M_3)$														
1	1.446	0.398	0.541	1.253	1.438	0.432	4.183	1.662	0.484	-0.301	0.066	1.659	-0.389	3.003
2	1.452	0.391	0.506	1.163	1.446	0.461	3.656	1.664	0.472	-0.236	-0.150	1.655	-0.271	2.925
3	1.459	0.386	0.486	1.069	1.453	0.472	3.403	1.666	0.465	-0.213	-0.239	1.661	0.002	2.919
4	1.466	0.382	0.468	1.110	1.456	0.474	3.370	1.667	0.460	-0.211	-0.260	1.659	0.046	2.889
5	1.469	0.376	0.389	0.754	1.467	0.501	3.089	1.670	0.452	-0.200	-0.272	1.667	0.016	2.884
6	1.473	0.373	0.392	0.831	1.465	0.502	3.045	1.671	0.447	-0.182	-0.332	1.667	0.117	2.826
7	1.480	0.370	0.375	0.807	1.477	0.501	3.182	1.672	0.442	-0.181	-0.343	1.664	0.190	2.841
8	1.485	0.369	0.412	0.994	1.478	0.508	3.134	1.673	0.439	-0.196	-0.331	1.674	0.276	2.776

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**Table A1.** (continued)

$T$ (min)	$D_c$ (mm)							$\text{Log}(N_c \text{ (m}^{-3} \cdot \text{mm}^{-1}))$						
	Mean	Std.	Skew.	Kurt.	Med.	Min.	Max.	Mean	Std.	Skew.	Kurt.	Med.	Min.	Max.
9	1.485	0.363	0.320	0.764	1.486	0.507	2.957	1.680	0.429	-0.156	-0.384	1.662	0.356	2.782
10	1.491	0.358	0.341	0.799	1.480	0.511	2.878	1.676	0.427	-0.177	-0.398	1.673	0.434	2.766
11	1.497	0.361	0.335	0.806	1.489	0.511	2.842	1.682	0.422	-0.207	-0.299	1.681	0.277	2.752
12	1.502	0.357	0.317	0.834	1.496	0.508	2.808	1.683	0.418	-0.189	-0.410	1.688	0.440	2.726
13	1.507	0.354	0.322	0.778	1.494	0.508	2.835	1.683	0.413	-0.190	-0.327	1.686	0.416	2.662
14	1.508	0.353	0.288	0.685	1.496	0.509	2.841	1.684	0.409	-0.166	-0.343	1.669	0.397	2.608
15	1.513	0.351	0.252	0.702	1.505	0.510	2.823	1.691	0.405	-0.163	-0.368	1.697	0.431	2.593
16	1.517	0.352	0.236	0.605	1.520	0.532	2.807	1.685	0.412	-0.238	-0.304	1.684	0.418	2.579
17	1.521	0.346	0.258	0.655	1.509	0.533	2.790	1.689	0.401	-0.168	-0.367	1.710	0.514	2.597
18	1.524	0.342	0.208	0.660	1.519	0.532	2.779	1.689	0.400	-0.166	-0.396	1.696	0.480	2.602
19	1.526	0.337	0.221	0.640	1.526	0.526	2.766	1.688	0.395	-0.194	-0.383	1.708	0.560	2.602
20	1.530	0.339	0.247	0.726	1.529	0.525	2.772	1.694	0.387	-0.168	-0.420	1.703	0.607	2.599
$(M_3, M_4)$														
1	1.617	0.490	0.805	1.928	1.586	0.438	4.775	1.479	0.502	-0.295	0.134	1.466	-0.885	2.917
2	1.630	0.481	0.759	1.767	1.603	0.479	4.460	1.474	0.492	-0.241	-0.055	1.460	-0.631	2.870
3	1.642	0.475	0.754	1.789	1.613	0.514	4.337	1.471	0.485	-0.224	-0.134	1.458	-0.251	2.854
4	1.652	0.470	0.737	1.820	1.624	0.512	4.314	1.469	0.478	-0.219	-0.181	1.455	-0.157	2.824
5	1.658	0.463	0.667	1.493	1.636	0.520	4.248	1.469	0.469	-0.203	-0.195	1.458	-0.313	2.809
6	1.666	0.463	0.708	1.608	1.640	0.523	3.920	1.467	0.466	-0.206	-0.168	1.454	-0.311	2.704
7	1.675	0.459	0.700	1.637	1.653	0.521	4.215	1.466	0.459	-0.197	-0.227	1.454	-0.265	2.731
8	1.682	0.456	0.727	1.906	1.650	0.526	4.273	1.466	0.455	-0.191	-0.280	1.460	-0.100	2.682
9	1.682	0.448	0.617	1.441	1.662	0.523	3.870	1.471	0.443	-0.137	-0.319	1.449	-0.034	2.655
10	1.692	0.445	0.676	1.688	1.659	0.528	3.983	1.464	0.442	-0.188	-0.257	1.450	-0.099	2.616
11	1.700	0.450	0.693	1.646	1.665	0.529	3.766	1.469	0.435	-0.202	-0.176	1.450	-0.037	2.589
12	1.709	0.447	0.695	1.683	1.681	0.525	3.742	1.468	0.433	-0.212	-0.208	1.451	-0.059	2.573
13	1.713	0.443	0.703	1.748	1.685	0.524	3.902	1.468	0.425	-0.165	-0.214	1.462	-0.089	2.519
14	1.716	0.438	0.632	1.493	1.687	0.525	3.750	1.467	0.420	-0.119	-0.271	1.455	-0.113	2.476
15	1.723	0.435	0.542	1.195	1.690	0.528	3.656	1.472	0.420	-0.178	-0.153	1.463	-0.097	2.452
16	1.728	0.442	0.712	1.964	1.695	0.617	4.106	1.466	0.423	-0.238	-0.082	1.454	-0.063	2.467
17	1.735	0.431	0.568	1.097	1.704	0.585	3.475	1.468	0.414	-0.161	-0.211	1.458	-0.028	2.479
18	1.738	0.430	0.593	1.384	1.707	0.582	3.723	1.467	0.408	-0.125	-0.315	1.463	0.043	2.479
19	1.743	0.423	0.561	1.115	1.711	0.565	3.359	1.464	0.408	-0.250	0.018	1.468	-0.043	2.474
20	1.745	0.423	0.540	1.097	1.718	0.563	3.351	1.472	0.395	-0.154	-0.192	1.474	0.056	2.465
$(M_4, M_5)$														
1	1.783	0.596	1.039	2.571	1.723	0.443	5.840	1.284	0.535	-0.333	0.359	1.274	-1.422	2.854
2	1.808	0.587	0.987	2.306	1.746	0.506	5.284	1.266	0.530	-0.338	0.421	1.256	-1.409	2.806
3	1.828	0.584	0.971	2.155	1.772	0.529	5.116	1.254	0.528	-0.365	0.514	1.245	-1.324	2.777
4	1.843	0.577	0.921	2.000	1.782	0.528	5.055	1.247	0.520	-0.361	0.489	1.232	-1.200	2.745

(continued on next page)

**Table A1.** (continued)

$T$ (min)	$D_c$ (mm)							$\text{Log}(N_c \text{ (m}^{-3} \cdot \text{mm}^{-1}))$						
	Mean	Std.	Skew.	Kurt.	Med.	Min.	Max.	Mean	Std.	Skew.	Kurt.	Med.	Min.	Max.
5	1.854	0.572	0.874	1.833	1.800	0.537	5.094	1.242	0.512	-0.380	0.637	1.231	-1.272	2.718
6	1.866	0.573	0.939	2.002	1.820	0.539	4.978	1.236	0.511	-0.402	0.667	1.230	-1.225	2.559
7	1.879	0.569	0.922	1.958	1.820	0.536	4.878	1.231	0.504	-0.425	0.746	1.226	-1.198	2.597
8	1.888	0.560	0.887	1.862	1.830	0.545	4.971	1.228	0.498	-0.377	0.464	1.235	-1.011	2.564
9	1.891	0.553	0.838	1.687	1.845	0.538	4.574	1.231	0.487	-0.348	0.566	1.222	-0.904	2.511
10	1.906	0.552	0.875	1.946	1.857	0.544	4.956	1.219	0.485	-0.403	0.682	1.210	-0.896	2.443
11	1.915	0.552	0.885	1.949	1.864	0.548	4.777	1.222	0.475	-0.328	0.380	1.212	-0.691	2.396
12	1.928	0.553	0.906	2.026	1.882	0.542	4.760	1.218	0.476	-0.391	0.524	1.211	-0.684	2.385
13	1.932	0.547	0.889	2.125	1.886	0.540	4.892	1.219	0.466	-0.301	0.337	1.210	-0.508	2.349
14	1.937	0.542	0.837	1.702	1.885	0.542	4.539	1.215	0.462	-0.311	0.612	1.203	-0.777	2.325
15	1.948	0.540	0.763	1.445	1.895	0.546	4.450	1.218	0.464	-0.388	0.777	1.203	-0.752	2.328
16	1.952	0.541	0.898	2.202	1.890	0.676	4.892	1.211	0.457	-0.293	0.260	1.194	-0.521	2.337
17	1.964	0.534	0.803	1.375	1.906	0.674	4.356	1.209	0.457	-0.384	0.669	1.201	-0.700	2.342
18	1.968	0.531	0.842	1.760	1.917	0.666	4.526	1.208	0.446	-0.275	0.282	1.191	-0.492	2.337
19	1.976	0.526	0.872	1.989	1.939	0.625	4.580	1.201	0.446	-0.431	0.855	1.177	-0.769	2.327
20	1.976	0.521	0.741	1.314	1.928	0.620	4.208	1.212	0.427	-0.256	0.387	1.197	-0.480	2.313
$(M_5, M_6)$														
1	1.938	0.702	1.157	2.601	1.847	0.448	6.354	1.089	0.571	-0.360	0.475	1.081	-1.642	2.782
2	1.980	0.700	1.107	2.375	1.892	0.537	6.180	1.051	0.576	-0.425	0.743	1.043	-1.855	2.732
3	2.012	0.701	1.118	2.403	1.925	0.544	5.997	1.027	0.579	-0.495	1.015	1.022	-1.935	2.686
4	2.035	0.697	1.061	2.191	1.938	0.543	5.817	1.010	0.575	-0.536	1.227	1.006	-1.961	2.654
5	2.054	0.697	1.040	2.132	1.958	0.551	5.832	0.998	0.569	-0.588	1.550	0.999	-2.068	2.611
6	2.071	0.697	1.105	2.363	1.988	0.553	5.677	0.985	0.569	-0.634	1.707	0.983	-2.054	2.442
7	2.089	0.694	1.090	2.312	2.003	0.549	5.589	0.975	0.564	-0.690	1.980	0.967	-2.069	2.440
8	2.103	0.686	1.063	2.167	2.018	0.562	5.392	0.967	0.561	-0.696	1.976	0.977	-2.012	2.425
9	2.109	0.680	1.034	2.080	2.029	0.553	5.297	0.966	0.553	-0.719	2.161	0.964	-1.983	2.357
10	2.130	0.679	1.004	1.932	2.047	0.561	5.369	0.950	0.550	-0.739	2.217	0.947	-1.926	2.257
11	2.144	0.675	0.959	1.731	2.055	0.568	5.223	0.947	0.544	-0.669	1.682	0.945	-1.786	2.187
12	2.160	0.679	1.006	1.833	2.077	0.561	5.205	0.940	0.546	-0.807	2.355	0.932	-1.923	2.194
13	2.164	0.671	0.956	1.846	2.075	0.556	5.313	0.943	0.535	-0.731	2.190	0.938	-1.902	2.220
14	2.172	0.667	0.990	1.900	2.086	0.558	5.180	0.934	0.530	-0.727	2.537	0.936	-2.137	2.200
15	2.188	0.663	0.872	1.391	2.085	0.566	5.145	0.933	0.535	-0.728	2.003	0.925	-1.536	2.185
16	2.194	0.659	0.910	1.594	2.099	0.706	5.251	0.923	0.526	-0.644	1.617	0.924	-1.539	2.187
17	2.210	0.658	0.934	1.469	2.118	0.698	5.162	0.918	0.533	-0.861	2.570	0.923	-1.762	2.186
18	2.217	0.655	0.942	1.561	2.126	0.692	4.978	0.914	0.523	-0.857	2.771	0.913	-1.849	2.176
19	2.232	0.648	0.925	1.626	2.157	0.708	5.123	0.900	0.521	-0.878	2.776	0.903	-1.777	2.162
20	2.227	0.639	0.816	1.083	2.131	0.705	4.863	0.916	0.495	-0.703	2.421	0.910	-1.657	2.145

Std.: Standard deviation, Skew.: Skewness, Kurt.: Kurtosis, Med.: Median, Min.: Minimum, Max.: Maximum.



**Table A2.** Average values of the constant  $C_n$ , for each integration time step and for each couple of moments.  $n$  is the moment order and  $T$  is in minute

$T \backslash n$	0.0	0.5	1.0	1.5	2.0	2.5	3.0	3.5	4.0	4.5	5.0	5.5	6.0	6.5	7.0
$(M_0, M_1)$															
1	1.000	0.982	1.000	1.054	1.146	1.286	1.483	1.757	2.131	2.645	3.350	4.323	5.676	7.568	10.234
2	1.000	0.982	1.000	1.054	1.149	1.291	1.494	1.776	2.166	2.706	3.457	4.508	5.993	8.109	11.152
3	1.000	0.982	1.000	1.055	1.151	1.296	1.502	1.792	2.194	2.755	3.540	4.650	6.234	8.518	11.844
4	1.000	0.981	1.000	1.056	1.153	1.299	1.509	1.803	2.213	2.786	3.595	4.744	6.396	8.797	12.327
5	1.000	0.981	1.000	1.056	1.154	1.301	1.513	1.810	2.227	2.811	3.639	4.821	6.530	9.031	12.734
6	1.000	0.981	1.000	1.056	1.155	1.303	1.517	1.819	2.242	2.839	3.686	4.904	6.674	9.281	13.171
7	1.000	0.981	1.000	1.057	1.156	1.307	1.524	1.831	2.262	2.871	3.740	4.993	6.823	9.532	13.597
8	1.000	0.981	1.000	1.057	1.157	1.309	1.527	1.835	2.270	2.885	3.765	5.038	6.904	9.678	13.859
9	1.000	0.981	1.000	1.057	1.157	1.309	1.527	1.837	2.273	2.893	3.780	5.066	6.956	9.773	14.031
10	1.000	0.981	1.000	1.058	1.159	1.313	1.535	1.851	2.298	2.934	3.849	5.183	7.152	10.105	14.597
11	1.000	0.981	1.000	1.058	1.160	1.315	1.539	1.857	2.308	2.951	3.878	5.232	7.240	10.263	14.881
12	1.000	0.981	1.000	1.059	1.161	1.318	1.544	1.867	2.325	2.979	3.926	5.313	7.376	10.493	15.276
13	1.000	0.981	1.000	1.059	1.162	1.319	1.547	1.871	2.331	2.989	3.939	5.331	7.400	10.524	15.319
14	1.000	0.981	1.000	1.059	1.162	1.320	1.549	1.875	2.338	3.002	3.963	5.375	7.482	10.679	15.610
15	1.000	0.980	1.000	1.059	1.164	1.323	1.553	1.883	2.352	3.026	4.004	5.446	7.605	10.892	15.982
16	1.000	0.980	1.000	1.059	1.164	1.324	1.555	1.886	2.356	3.032	4.015	5.465	7.642	10.964	16.119
17	1.000	0.980	1.000	1.060	1.165	1.327	1.561	1.897	2.376	3.066	4.073	5.565	7.809	11.244	16.592
18	1.000	0.980	1.000	1.060	1.166	1.327	1.562	1.897	2.376	3.067	4.075	5.570	7.823	11.279	16.671
19	1.000	0.980	1.000	1.060	1.166	1.328	1.565	1.903	2.388	3.089	4.114	5.642	7.956	11.526	17.136
20	1.000	0.980	1.000	1.060	1.166	1.328	1.564	1.901	2.383	3.079	4.094	5.602	7.880	11.383	16.868
$(M_1, M_2)$															
1	1.146	1.052	1.000	0.984	1.000	1.047	1.128	1.248	1.414	1.639	1.939	2.337	2.865	3.568	4.506
2	1.149	1.052	1.000	0.984	1.000	1.048	1.131	1.255	1.428	1.665	1.984	2.414	2.994	3.779	4.849
3	1.151	1.053	1.000	0.984	1.000	1.049	1.134	1.261	1.439	1.684	2.017	2.470	3.086	3.930	5.094
4	1.153	1.054	1.000	0.983	1.000	1.050	1.136	1.264	1.445	1.695	2.037	2.504	3.144	4.028	5.257
5	1.154	1.054	1.000	0.983	1.000	1.050	1.137	1.267	1.451	1.705	2.055	2.535	3.197	4.116	5.404
6	1.155	1.054	1.000	0.983	1.000	1.051	1.138	1.270	1.457	1.717	2.075	2.569	3.254	4.211	5.562
7	1.156	1.055	1.000	0.983	1.000	1.051	1.140	1.273	1.463	1.727	2.092	2.596	3.300	4.287	5.686
8	1.157	1.055	1.000	0.983	1.000	1.051	1.140	1.274	1.465	1.731	2.101	2.613	3.329	4.338	5.775
9	1.157	1.055	1.000	0.983	1.000	1.051	1.141	1.275	1.468	1.736	2.109	2.628	3.354	4.380	5.846
10	1.159	1.056	1.000	0.983	1.000	1.052	1.143	1.280	1.477	1.751	2.135	2.670	3.422	4.492	6.028
11	1.160	1.056	1.000	0.982	1.000	1.053	1.144	1.282	1.479	1.755	2.142	2.683	3.448	4.538	6.110
12	1.161	1.057	1.000	0.982	1.000	1.053	1.145	1.285	1.485	1.766	2.160	2.713	3.495	4.615	6.235
13	1.162	1.057	1.000	0.982	1.000	1.053	1.146	1.286	1.486	1.768	2.161	2.713	3.494	4.611	6.226
14	1.162	1.057	1.000	0.982	1.000	1.053	1.146	1.287	1.488	1.772	2.170	2.730	3.525	4.666	6.326
15	1.164	1.058	1.000	0.982	1.000	1.054	1.147	1.289	1.493	1.781	2.185	2.755	3.566	4.735	6.441
16	1.164	1.058	1.000	0.982	1.000	1.054	1.148	1.290	1.494	1.782	2.187	2.760	3.577	4.756	6.482
17	1.165	1.058	1.000	0.982	1.000	1.054	1.149	1.293	1.501	1.794	2.208	2.794	3.631	4.843	6.620
18	1.166	1.058	1.000	0.982	1.000	1.054	1.149	1.293	1.500	1.793	2.207	2.794	3.635	4.854	6.645
19	1.166	1.058	1.000	0.982	1.000	1.055	1.151	1.296	1.506	1.804	2.225	2.825	3.689	4.949	6.814
20	1.166	1.058	1.000	0.982	1.000	1.055	1.150	1.294	1.503	1.797	2.213	2.805	3.653	4.887	6.705

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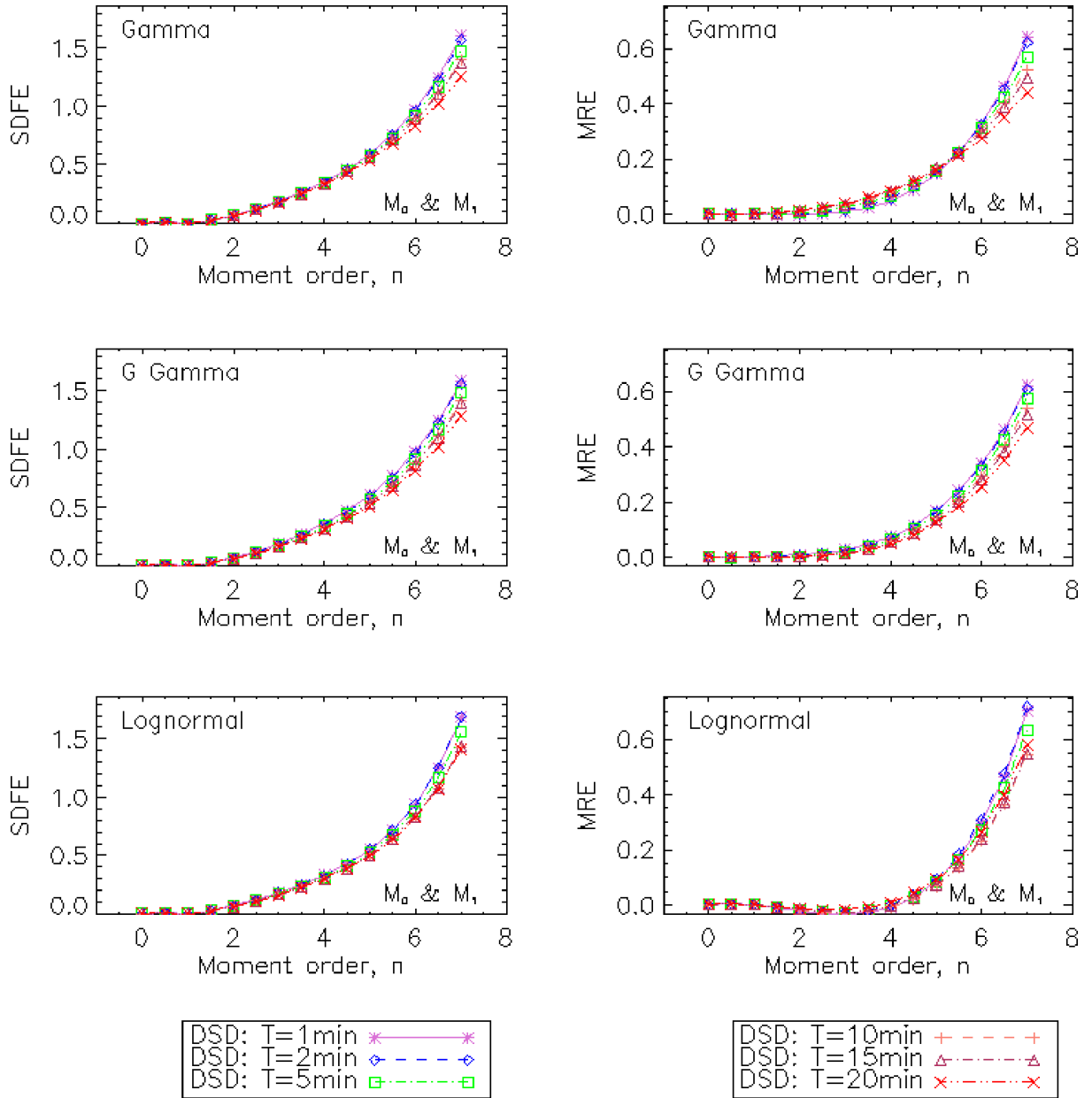
**Table A2.** (continued)

$T \backslash n$	0.0	0.5	1.0	1.5	2.0	2.5	3.0	3.5	4.0	4.5	5.0	5.5	6.0	6.5	7.0
$(M_2, M_3)$															
1	1.460	1.260	1.128	1.045	1.000	0.986	1.000	1.041	1.111	1.212	1.350	1.531	1.767	2.072	2.463
2	1.471	1.267	1.131	1.046	1.000	0.986	1.000	1.043	1.116	1.222	1.370	1.567	1.827	2.168	2.614
3	1.480	1.272	1.134	1.047	1.000	0.985	1.000	1.044	1.119	1.230	1.383	1.590	1.866	2.232	2.716
4	1.486	1.275	1.136	1.048	1.000	0.985	1.000	1.044	1.121	1.234	1.391	1.605	1.891	2.274	2.784
5	1.491	1.278	1.137	1.048	1.000	0.985	1.000	1.045	1.122	1.238	1.399	1.618	1.914	2.312	2.846
6	1.496	1.281	1.138	1.049	1.000	0.985	1.000	1.046	1.124	1.242	1.407	1.632	1.938	2.350	2.910
7	1.502	1.284	1.140	1.049	1.000	0.985	1.000	1.046	1.126	1.245	1.412	1.642	1.955	2.379	2.956
8	1.505	1.285	1.140	1.050	1.000	0.985	1.000	1.046	1.127	1.247	1.416	1.650	1.968	2.402	2.994
9	1.506	1.286	1.141	1.050	1.000	0.984	1.000	1.047	1.128	1.249	1.420	1.657	1.980	2.421	3.026
10	1.514	1.291	1.143	1.051	1.000	0.984	1.000	1.048	1.130	1.254	1.429	1.672	2.004	2.460	3.088
11	1.518	1.292	1.144	1.051	1.000	0.984	1.000	1.048	1.130	1.255	1.431	1.677	2.014	2.479	3.121
12	1.523	1.295	1.145	1.051	1.000	0.984	1.000	1.048	1.132	1.258	1.437	1.687	2.031	2.505	3.162
13	1.526	1.296	1.146	1.051	1.000	0.984	1.000	1.048	1.132	1.258	1.437	1.685	2.027	2.499	3.152
14	1.527	1.297	1.146	1.052	1.000	0.984	1.000	1.048	1.133	1.260	1.441	1.693	2.042	2.525	3.197
15	1.532	1.300	1.147	1.052	1.000	0.984	1.000	1.049	1.134	1.263	1.446	1.703	2.058	2.551	3.239
16	1.534	1.301	1.148	1.052	1.000	0.984	1.000	1.049	1.134	1.262	1.446	1.703	2.060	2.556	3.251
17	1.540	1.304	1.149	1.053	1.000	0.984	1.000	1.050	1.136	1.267	1.454	1.716	2.081	2.589	3.301
18	1.540	1.304	1.149	1.053	1.000	0.984	1.000	1.050	1.136	1.266	1.454	1.717	2.083	2.595	3.313
19	1.544	1.306	1.151	1.053	1.000	0.983	1.000	1.050	1.138	1.270	1.461	1.729	2.105	2.633	3.380
20	1.542	1.305	1.150	1.053	1.000	0.983	1.000	1.050	1.136	1.267	1.456	1.720	2.089	2.606	3.334
$(M_3, M_4)$															
1	2.001	1.639	1.392	1.224	1.111	1.039	1.000	0.988	1.000	1.035	1.094	1.177	1.289	1.434	1.618
2	2.042	1.665	1.408	1.233	1.116	1.041	1.000	0.987	1.000	1.037	1.101	1.192	1.316	1.478	1.688
3	2.073	1.684	1.420	1.240	1.119	1.042	1.000	0.987	1.000	1.039	1.105	1.201	1.332	1.506	1.733
4	2.092	1.695	1.426	1.243	1.121	1.043	1.000	0.987	1.000	1.040	1.108	1.207	1.344	1.526	1.766
5	2.108	1.706	1.432	1.247	1.122	1.044	1.000	0.986	1.000	1.041	1.110	1.212	1.353	1.543	1.793
6	2.127	1.717	1.439	1.251	1.124	1.044	1.000	0.986	1.000	1.041	1.113	1.217	1.363	1.559	1.820
7	2.144	1.727	1.445	1.254	1.126	1.045	1.000	0.986	1.000	1.042	1.114	1.221	1.370	1.571	1.840
8	2.152	1.731	1.447	1.255	1.127	1.045	1.000	0.986	1.000	1.043	1.116	1.225	1.376	1.582	1.859
9	2.159	1.736	1.451	1.257	1.128	1.045	1.000	0.986	1.000	1.043	1.117	1.227	1.381	1.590	1.872
10	2.184	1.751	1.459	1.262	1.130	1.046	1.000	0.985	1.000	1.044	1.119	1.232	1.389	1.604	1.895
11	2.191	1.755	1.461	1.263	1.130	1.046	1.000	0.985	1.000	1.044	1.120	1.234	1.395	1.615	1.912
12	2.209	1.766	1.468	1.266	1.132	1.047	1.000	0.985	1.000	1.045	1.122	1.237	1.400	1.623	1.926
13	2.212	1.767	1.468	1.266	1.132	1.047	1.000	0.985	1.000	1.044	1.121	1.236	1.398	1.620	1.921
14	2.220	1.772	1.471	1.268	1.133	1.047	1.000	0.985	1.000	1.045	1.123	1.240	1.405	1.632	1.941
15	2.235	1.781	1.476	1.271	1.134	1.048	1.000	0.985	1.000	1.045	1.124	1.242	1.410	1.641	1.957
16	2.236	1.781	1.476	1.270	1.134	1.048	1.000	0.985	1.000	1.045	1.124	1.244	1.413	1.647	1.967
17	2.257	1.794	1.483	1.275	1.136	1.048	1.000	0.985	1.000	1.046	1.127	1.248	1.419	1.657	1.982
18	2.256	1.793	1.483	1.274	1.136	1.048	1.000	0.985	1.000	1.046	1.127	1.249	1.422	1.662	1.991
19	2.273	1.803	1.489	1.278	1.138	1.049	1.000	0.985	1.000	1.047	1.129	1.253	1.430	1.677	2.018
20	2.263	1.797	1.485	1.275	1.136	1.048	1.000	0.985	1.000	1.046	1.127	1.249	1.424	1.666	2.000

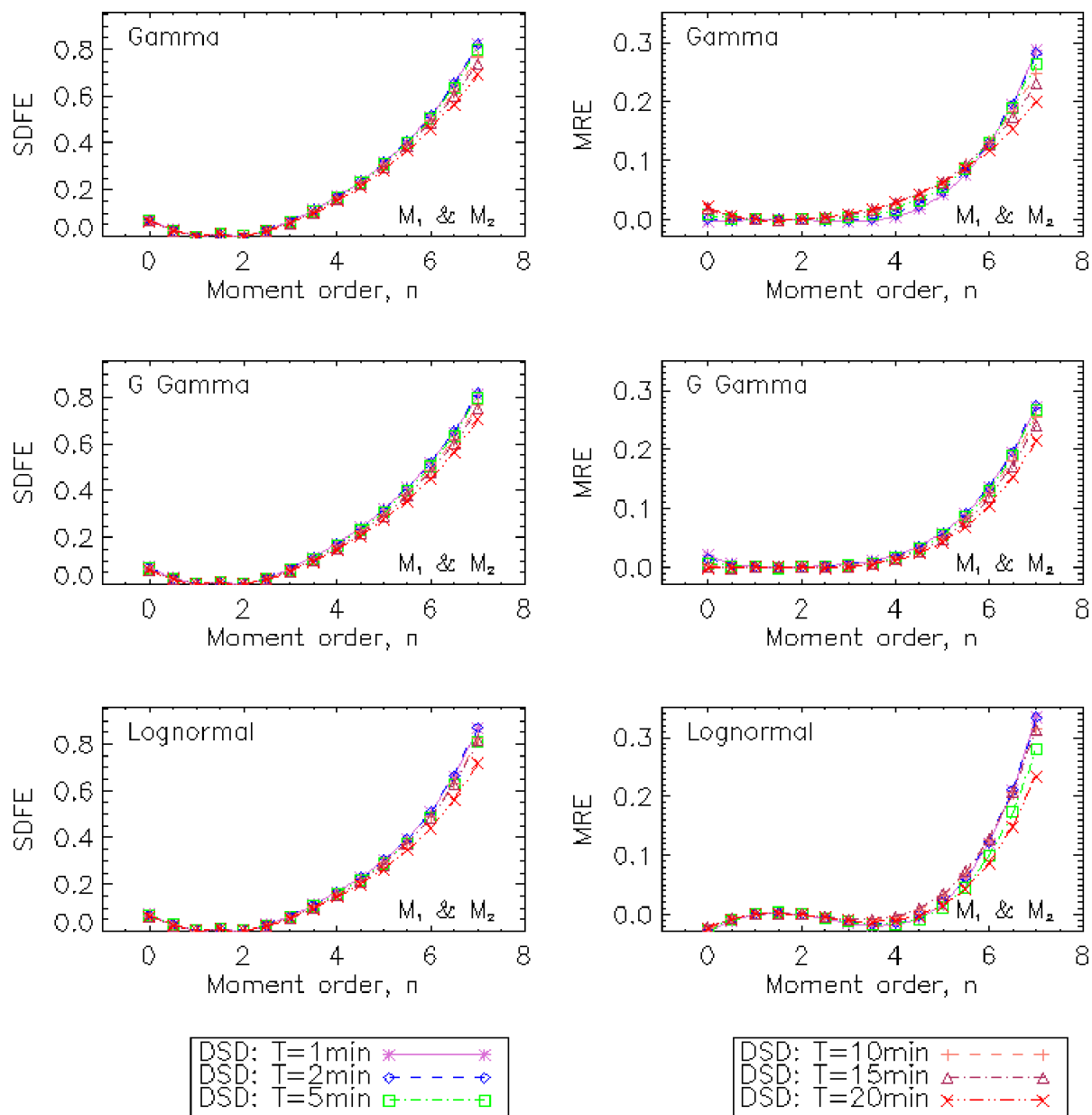
(continued on next page)

**Table A2.** (continued)

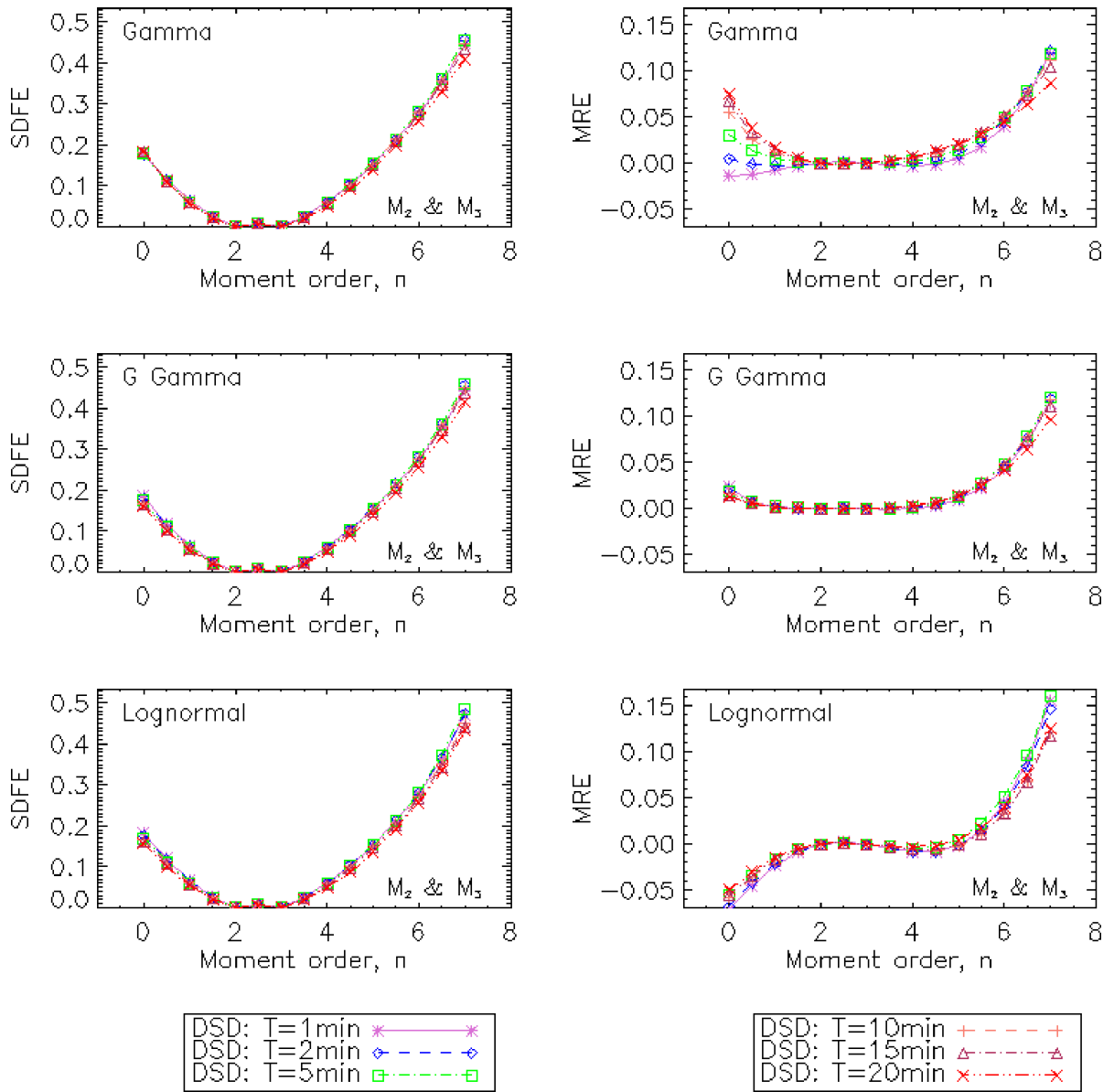
$T \backslash n$	0.0	0.5	1.0	1.5	2.0	2.5	3.0	3.5	4.0	4.5	5.0	5.5	6.0	6.5	7.0
$(M_4, M_5)$															
1	2.863	2.243	1.822	1.531	1.329	1.189	1.094	1.033	1.000	0.990	1.000	1.029	1.078	1.146	1.237
2	2.995	2.328	1.877	1.567	1.351	1.202	1.101	1.036	1.000	0.989	1.000	1.032	1.086	1.164	1.267
3	3.090	2.388	1.915	1.590	1.366	1.210	1.105	1.037	1.000	0.988	1.000	1.034	1.091	1.174	1.285
4	3.149	2.425	1.938	1.605	1.375	1.216	1.108	1.038	1.000	0.988	1.000	1.035	1.095	1.182	1.299
5	3.203	2.459	1.960	1.619	1.383	1.221	1.110	1.039	1.000	0.988	1.000	1.036	1.098	1.188	1.310
6	3.258	2.494	1.982	1.633	1.392	1.225	1.113	1.040	1.000	0.987	1.000	1.037	1.101	1.194	1.322
7	3.305	2.522	1.999	1.643	1.398	1.229	1.114	1.041	1.000	0.987	1.000	1.038	1.103	1.199	1.330
8	3.336	2.541	2.011	1.651	1.403	1.232	1.116	1.041	1.000	0.987	1.000	1.039	1.105	1.203	1.338
9	3.362	2.558	2.022	1.658	1.407	1.234	1.117	1.042	1.000	0.987	1.000	1.039	1.107	1.206	1.343
10	3.428	2.598	2.046	1.672	1.416	1.239	1.119	1.043	1.000	0.987	1.000	1.040	1.109	1.211	1.351
11	3.453	2.613	2.055	1.678	1.419	1.241	1.120	1.043	1.000	0.986	1.000	1.041	1.111	1.215	1.359
12	3.499	2.640	2.072	1.688	1.424	1.244	1.122	1.044	1.000	0.986	1.000	1.041	1.113	1.218	1.364
13	3.497	2.638	2.070	1.686	1.423	1.243	1.121	1.043	1.000	0.986	1.000	1.041	1.112	1.216	1.362
14	3.529	2.658	2.082	1.694	1.428	1.246	1.123	1.044	1.000	0.986	1.000	1.042	1.114	1.221	1.371
15	3.570	2.683	2.097	1.703	1.433	1.249	1.124	1.044	1.000	0.986	1.000	1.042	1.116	1.225	1.377
16	3.575	2.685	2.099	1.703	1.434	1.249	1.124	1.045	1.000	0.986	1.000	1.043	1.117	1.228	1.383
17	3.637	2.722	2.121	1.717	1.442	1.254	1.127	1.045	1.000	0.986	1.000	1.043	1.118	1.230	1.386
18	3.638	2.723	2.122	1.718	1.442	1.254	1.127	1.045	1.000	0.986	1.000	1.044	1.120	1.233	1.391
19	3.689	2.755	2.141	1.730	1.449	1.258	1.129	1.046	1.000	0.985	1.000	1.045	1.122	1.239	1.403
20	3.653	2.732	2.127	1.721	1.444	1.255	1.127	1.045	1.000	0.986	1.000	1.044	1.120	1.235	1.396
$(M_5, M_6)$															
1	4.167	3.144	2.460	1.991	1.664	1.434	1.271	1.156	1.078	1.028	1.000	0.991	1.000	1.024	1.064
2	4.532	3.379	2.614	2.093	1.732	1.478	1.299	1.173	1.086	1.031	1.000	0.990	1.000	1.028	1.073
3	4.784	3.539	2.717	2.160	1.776	1.506	1.316	1.183	1.091	1.033	1.000	0.990	1.000	1.030	1.079
4	4.961	3.650	2.788	2.207	1.806	1.526	1.329	1.190	1.095	1.034	1.000	0.989	1.000	1.031	1.083
5	5.115	3.747	2.850	2.247	1.832	1.543	1.339	1.196	1.098	1.035	1.000	0.989	1.000	1.032	1.087
6	5.276	3.848	2.914	2.288	1.858	1.559	1.349	1.202	1.101	1.036	1.000	0.989	1.000	1.034	1.090
7	5.403	3.924	2.961	2.318	1.877	1.571	1.356	1.206	1.103	1.037	1.000	0.988	1.000	1.034	1.092
8	5.506	3.989	3.003	2.344	1.895	1.583	1.364	1.210	1.105	1.038	1.000	0.988	1.000	1.035	1.095
9	5.584	4.038	3.034	2.365	1.908	1.591	1.368	1.213	1.107	1.038	1.000	0.988	1.000	1.036	1.096
10	5.752	4.139	3.096	2.402	1.931	1.605	1.377	1.218	1.109	1.039	1.000	0.988	1.000	1.036	1.099
11	5.853	4.201	3.135	2.427	1.947	1.616	1.384	1.222	1.111	1.040	1.000	0.987	1.000	1.037	1.101
12	5.966	4.268	3.175	2.452	1.962	1.624	1.389	1.225	1.113	1.040	1.000	0.987	1.000	1.038	1.102
13	5.941	4.251	3.163	2.443	1.956	1.620	1.386	1.223	1.112	1.040	1.000	0.987	1.000	1.038	1.102
14	6.055	4.321	3.207	2.472	1.974	1.632	1.393	1.227	1.114	1.041	1.000	0.987	1.000	1.039	1.105
15	6.171	4.390	3.249	2.498	1.991	1.642	1.399	1.231	1.116	1.041	1.000	0.987	1.000	1.039	1.107
16	6.224	4.423	3.270	2.511	1.999	1.648	1.404	1.233	1.117	1.042	1.000	0.987	1.000	1.040	1.108
17	6.357	4.501	3.316	2.539	2.016	1.657	1.409	1.236	1.118	1.042	1.000	0.987	1.000	1.040	1.108
18	6.401	4.528	3.334	2.551	2.024	1.663	1.413	1.238	1.120	1.043	1.000	0.986	1.000	1.041	1.110
19	6.569	4.630	3.397	2.590	2.049	1.678	1.422	1.244	1.122	1.044	1.000	0.986	1.000	1.042	1.114
20	6.451	4.557	3.352	2.562	2.031	1.667	1.415	1.240	1.120	1.043	1.000	0.986	1.000	1.041	1.112



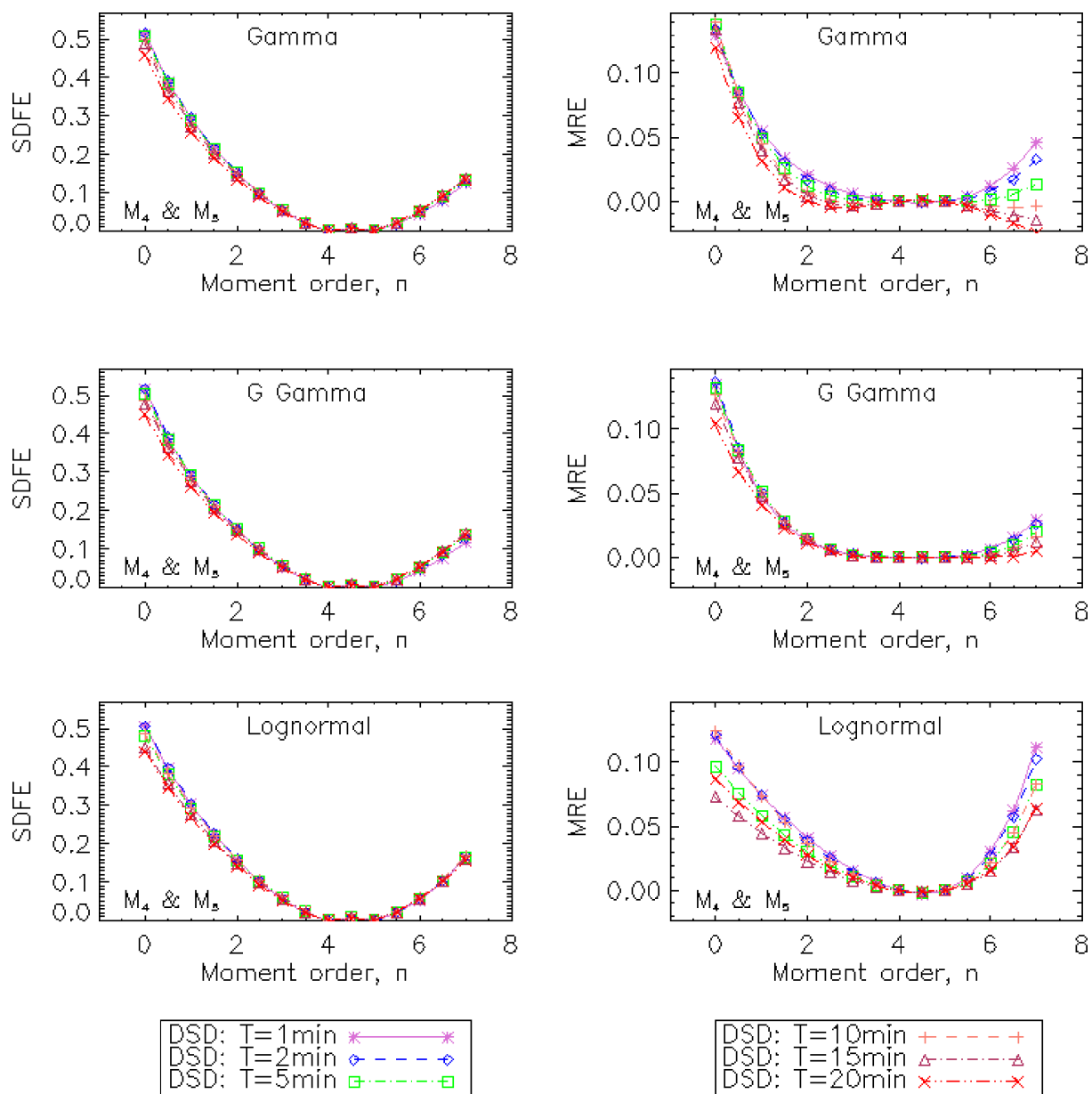
**Figure A1.** Comparison of moments estimated to moments measured with SDFE and MRE. Case of the couple of moments ( $M_0$ ,  $M_1$ ), relating to the DSD models gamma, generalized gamma (noted G Gamma) and lognormal.



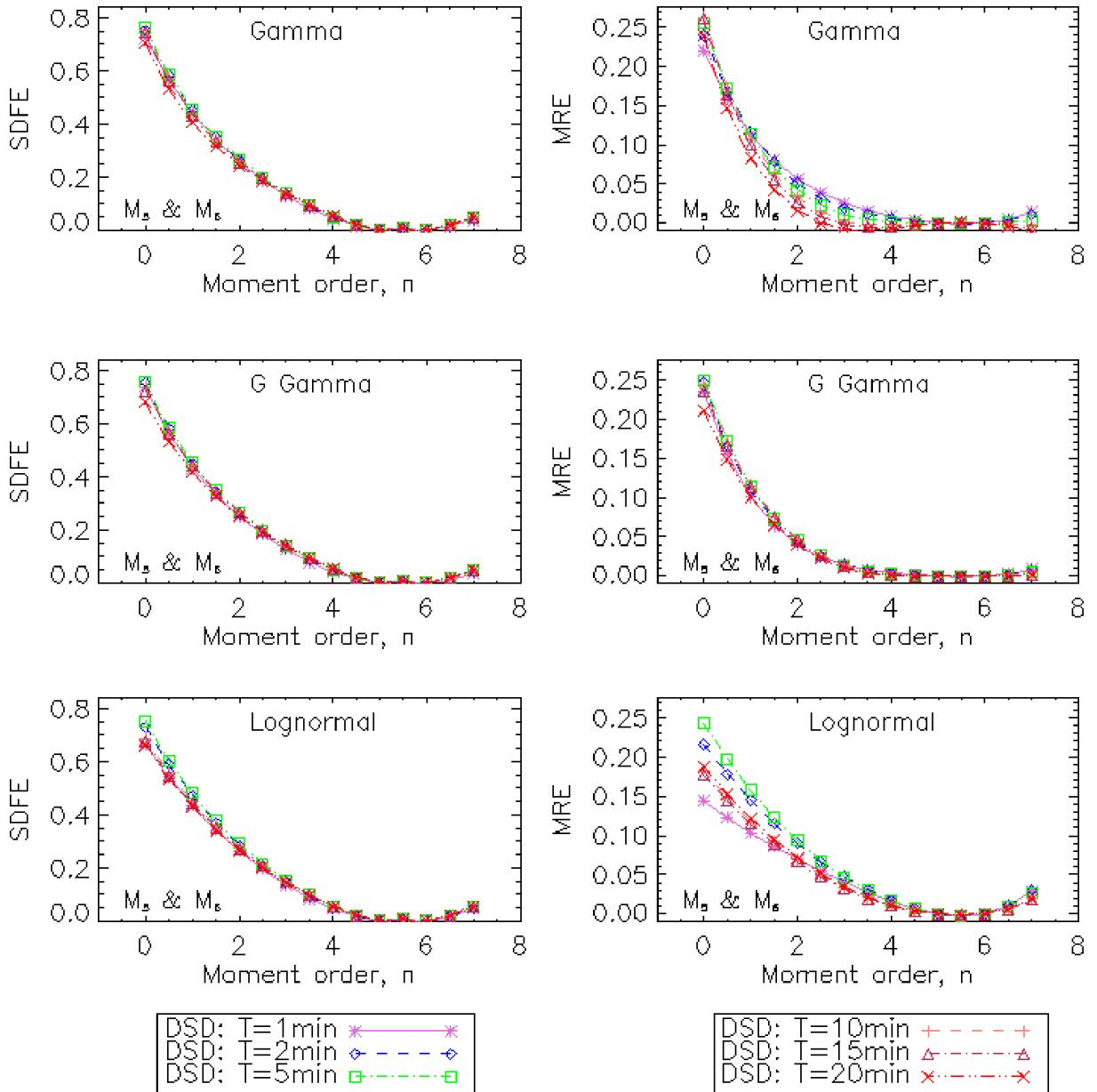
**Figure A2.** Comparison of moments estimated to moments measured with SDFE and MRE. Case of the couple of moments ( $M_1, M_2$ ), relating to the DSD models gamma, generalized gamma (noted G Gamma) and lognormal.



**Figure A3.** Comparison of moments estimated to moments measured with SDFE and MRE. Case of the couple of moments ( $M_2$ ,  $M_3$ ), relating to the DSD models gamma, generalized gamma (noted G Gamma) and lognormal.



**Figure A4.** Comparison of moments estimated to moments measured with SDFE and MRE. Case of the couple of moments ( $M_4$ ,  $M_5$ ), relating to the DSD models gamma, generalized gamma (noted G Gamma) and lognormal.



**Figure A5.** Comparison of moments estimated to moments measured with SDFE and MRE. Case of the couple of moments ( $M_5$ ,  $M_6$ ), relating to the DSD models gamma, generalized gamma (noted G Gamma) and lognormal.

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