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Motivations

- ✓ Raindrop size distributions (DSD) can be written as $N(D)=n_c f(D)$; n_c is the number concentration, $f(D)$ a pdf
- ✓ Several pdfs have been proposed: the gamma is the most frequently assumed [Adirosi et al. 2014].
- ✓ The **third to sixth moments** of the DSD are proportional to relevant **hydrological and meteorological parameters** (i.e. rain rate, liquid water content, kinetic energy, reflectivity): performance of retrievals depends on the form of DSD assumed.
- ✓ Higher the order of moments, the **higher is the weight of the drops in the DSD right tail**.
- ✓ The form of the right tail of a DSD governs both the magnitude and the frequency of the bigger drops.
- ✓ Retrieval of parametric DSDs aims to model the whole drop spectra: the retrieved DSD could not be adequate for specific portions, such as the upper tail, critical for some meteorological parameters.
- ✓ The **method of moments**, commonly used for retrieving DSD parameters, is **known to be biased** [Smith and Kliche 2005] determining unrealistic variability of retrieved parameters [Ulbrich and Atlas, 1998].

Objectives

To investigate the performance of four one-sided continuous distributions with different tails (the **Pareto**, the **Lognormal**, the **Gamma** and the **Weibull** distributions) through fitting both

- ☐ to the large drops only (**upper tail fitting**)
- ☐ to the entire measured spectra (**whole fitting**)

Heavier tail \rightarrow Lighter tail
 Pareto Lognormal Weibull $\gamma < 1$ Gamma Weibull $\gamma > 1$

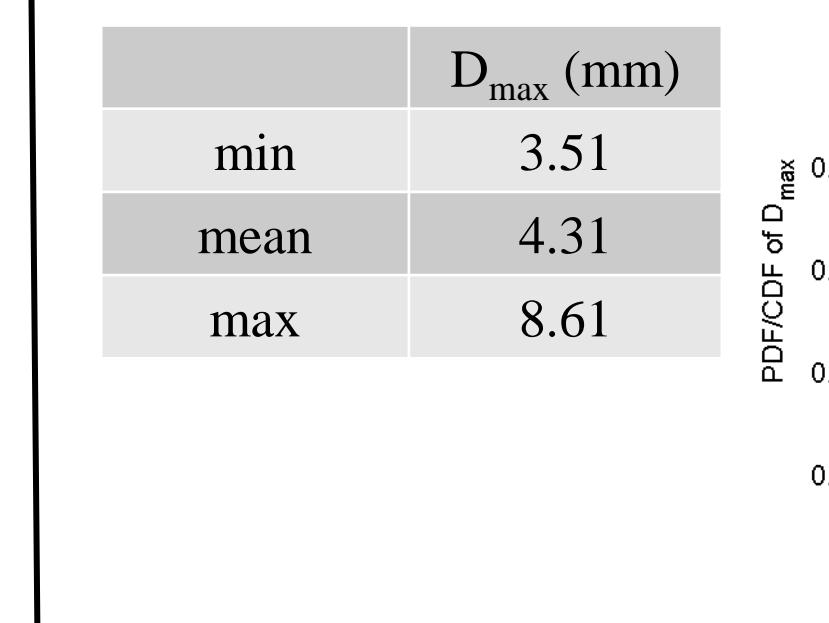
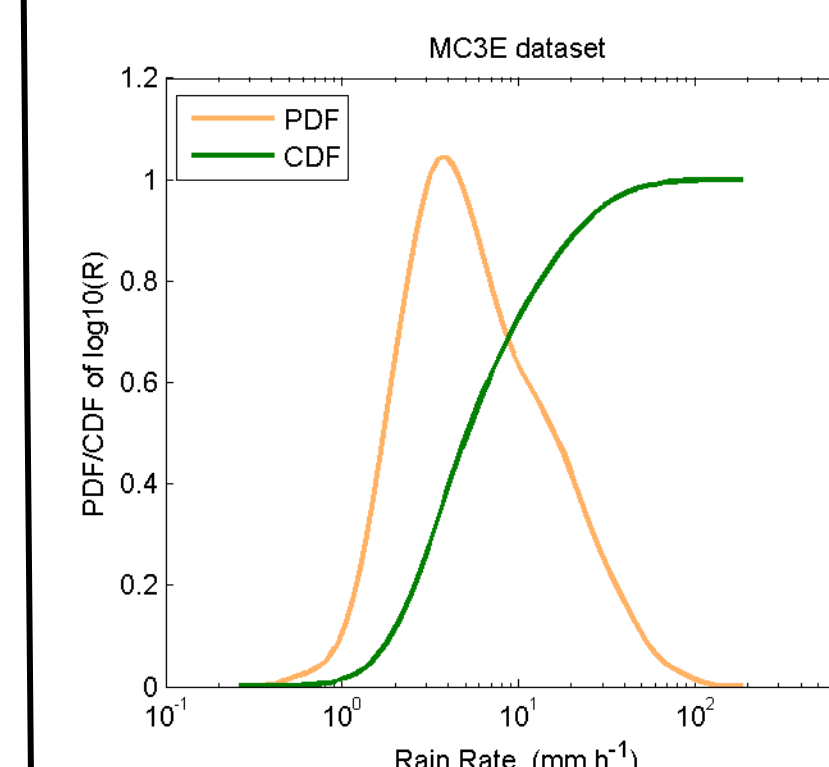
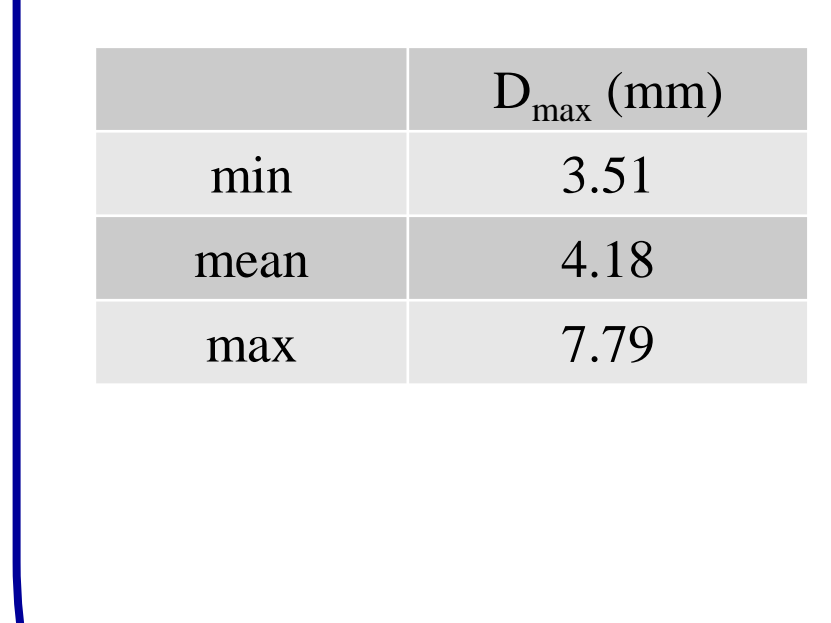
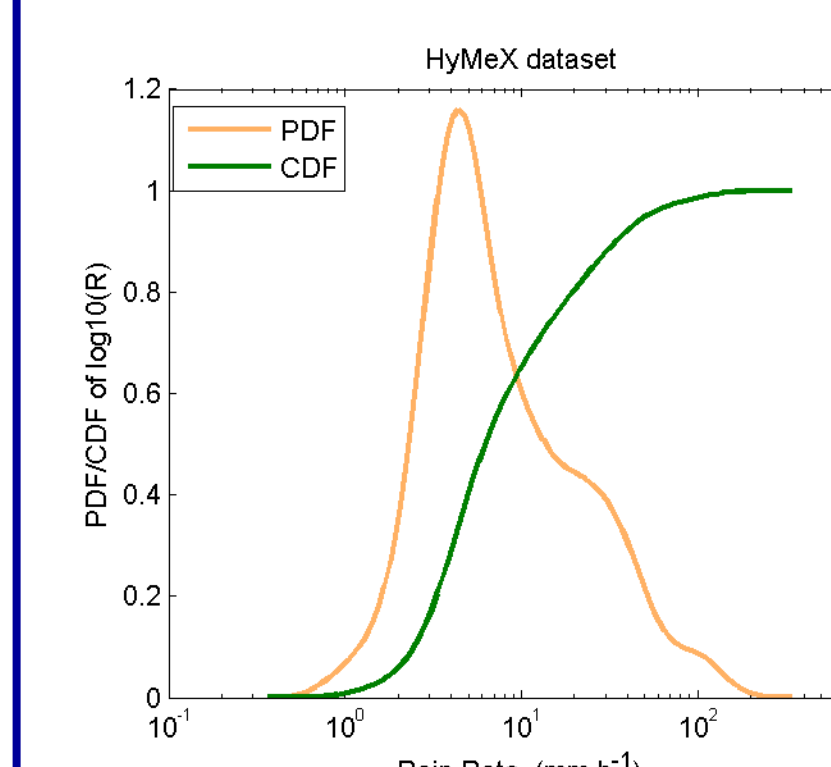
Measured Datasets



HyMex
(485 1-min DSDs)

Data consist of 1-min drop spectra collected during **HyMex SOP1.1 (Rome)** and **MC3E (Central Oklahoma)** by a NASA two-dimensional video disdrometer (2DVD) provided to support GPM pre-launch ground validation activities. Only the **spectra with a minimum of 100 drops and $D_{max} > 3.5$ mm** are retained: large drops are adequately sampled, but DSDs of light rain can be discarded.

MC3E
(975 1-min DSDs)



Tail definition: upper part of the drop spectrum with $D > D_{th}$ \rightarrow 90th percentile of drop diameters measured in a given minute

Fitting Method

The fitting method permits to directly fit any theoretical distribution to the empirical one minimizing numerically the modified mean square error (MSE) norm $N1$ [Papalexiou et al., 2013]:

$$N1 = \frac{1}{N} \sum_{i=N-n+1}^n \left(\frac{\overline{F}_T(D_i)}{\overline{F}_E(D_i)} - 1 \right)^2$$

$n =$ is the total number of drops measured in 1 minute

- **Tail Fitting** $N =$ number of drops with $D \geq D_{th}$
- **Whole Fitting** $N = n$

$\overline{F}_E(D_i)$ is the **empirical** exceedance probability function (EPF). First the 1-min $N(D)$ is computed from the 2DVD raw data

$$N(D_i) = \frac{1}{\Delta t} \sum_{k=1}^{M_i} \frac{1}{A_k v_k}$$

where Δt (s) is the time interval, M_i is the total number of drops in a given minute, v_k (m s⁻¹) is the fall velocity and A_k (mm²) is the cross-sectional area.

Then the probability density function of drop diameter (f_D) is computed and the empirical EPF can be obtained. Note that when the three parameter gamma distribution is used to model the measured DSD

$$f_D(D) = \frac{\Lambda^{\mu+1}}{\Gamma(\mu+1)} e^{-\Lambda D} D^\mu; \quad \mu = \text{shape parameter and } \Lambda \text{ (mm}^{-1}\text{)} = \text{slope parameter}$$

$\overline{F}_T(D_i)$ is the **theoretical** exceedance probability function (EPF) computed considering each single drop measured in a given minute:

Pareto (**PA**) $\rightarrow \overline{F}_{PA}(x) = \left(1 + \gamma \left(\frac{x}{\beta}\right)\right)^{-\frac{1}{\gamma}}$ with $\beta > 0$ and $\gamma > 0$

Lognormal (**LN**) $\rightarrow \overline{F}_{LN}(x) = \frac{1}{2} \operatorname{erfc}\left(\ln\left(\frac{x}{\beta}\right)^{\frac{1}{\gamma}}\right)$; with $\beta > 0$ and $\gamma > 0$

Weibull (**WE**) $\rightarrow \overline{F}_{WE}(x) = \exp\left(-\left(\frac{x}{\beta}\right)^\gamma\right)$ with $\beta > 0$ and $\gamma > 0$

Gamma (**GA**) $\rightarrow \overline{F}_{GA}(x) = \Gamma\left(\gamma, \frac{x}{\beta}\right) / \Gamma(\gamma)$;

$$\Gamma(s, x) = \int_x^\infty t^{s-1} e^{-t} dt; \quad \Gamma(s) = \int_0^\infty t^{s-1} e^{-t} dt;$$

where β and γ are the scale and shape parameter, respectively. The shape parameter controls the asymptotic behavior of the tail.

Whole Fitting

Although the two datasets were collected in two different areas of the world, the estimated parameters β and γ are similar. The shape parameter (γ) of Pareto distribution is very low (means equal to 0.01): the Pareto degenerates to the exponential tail ($\gamma = 0$). Furthermore its performance is the worst (i.e. $N1$ values are the highest).

LN, WE, and GA fittings were compared in pair:

HYMEX DATASET			MC3E DATASET				
	LN	WE	GA	LN	WE	GA	
LN	\	38.4%	30.5%	LN	\	39.5%	33.6%
WE	61.6%	\	43.9%	WE	60.5%	\	46.3%
GA	69.5%	56.1%	\	GA	66.4%	53.7%	\

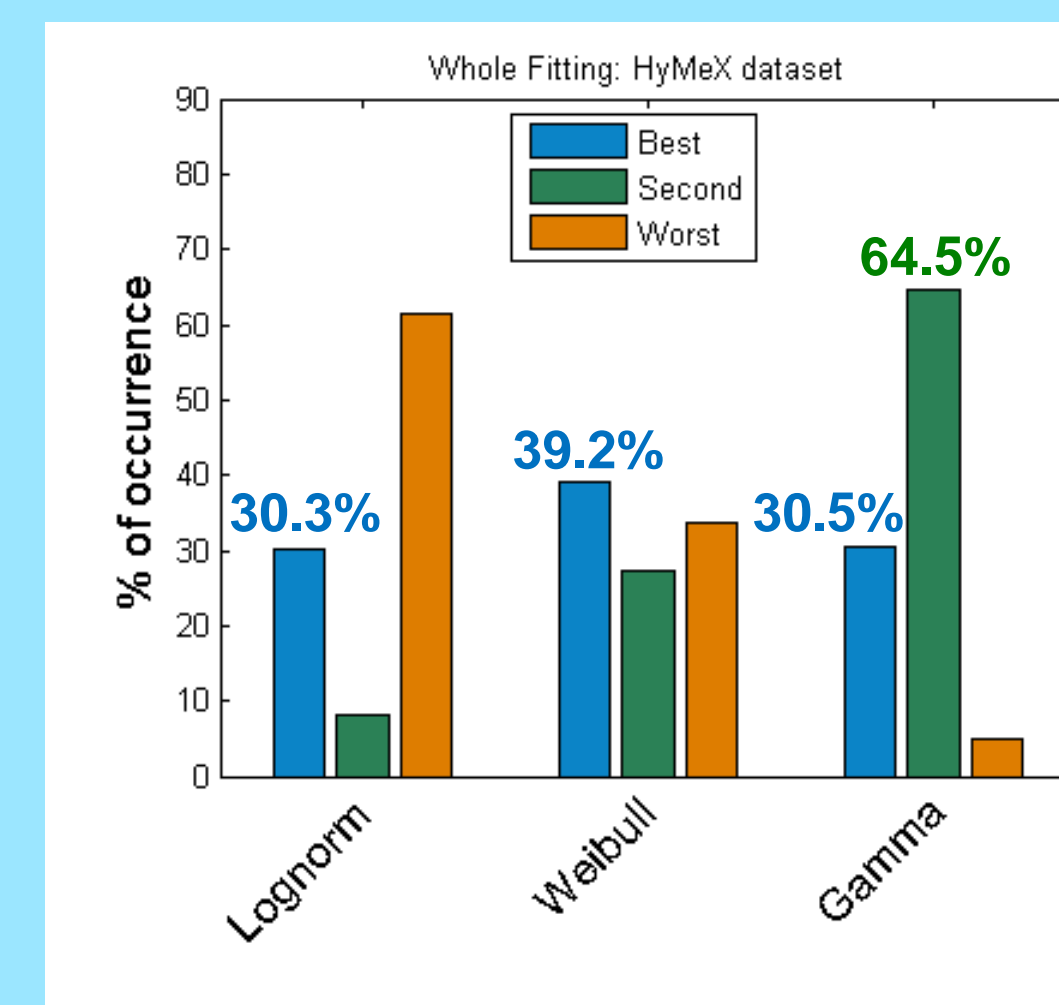
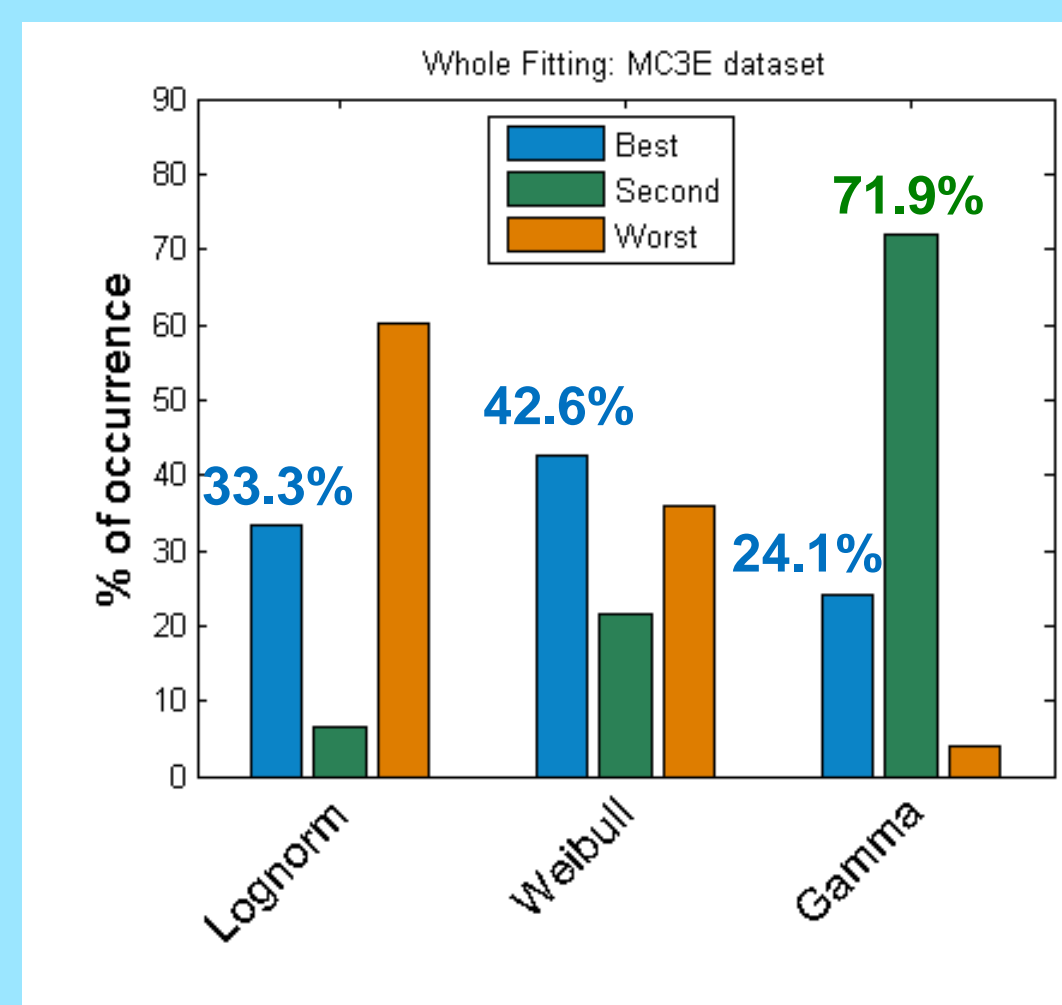
The gamma distribution was better fitted in about 60% of the 1-min measured spectra.

Note that the μ and Λ parameters can be obtained from the estimated β and γ as:

- $\mu = \gamma - 1$
- $\Lambda = 1/\beta$.

PERFORMANCE OF THE FITTINGS

The distribution that better fits a given drop spectrum is the one with the minimum $N1$.



Despite the Weibull distribution is the one with the highest percentages of success, it is closely followed by the Gamma distribution. In fact when the Gamma is not the best, its $N1$ values are slightly bigger than the ones of the Weibull.

Results

As for the whole fitting, the Pareto distribution degenerates to exponential (shape parameter ~ 0); the mean and the median of $N1$ are the highest, with a percentage of success around 1%. These findings confirm that a **heavy tail is not adequate** to model the upper part of the disdrometer measured drop spectra.

LN, WE and GA fittings compared in pair:

HYMEX DATASET			
	LN	WE	GA
LN	\	40.6%	39.2%
WE	59.4%	\	49.7%
GA	60.8%	50.3%	\

MC3E DATASET			
	LN	WE	GA
LN	\	52.9%	51.3%
WE	47.1%	\	44.0%
GA	48.7%	56.8%	\

When the tail fitting is performed:

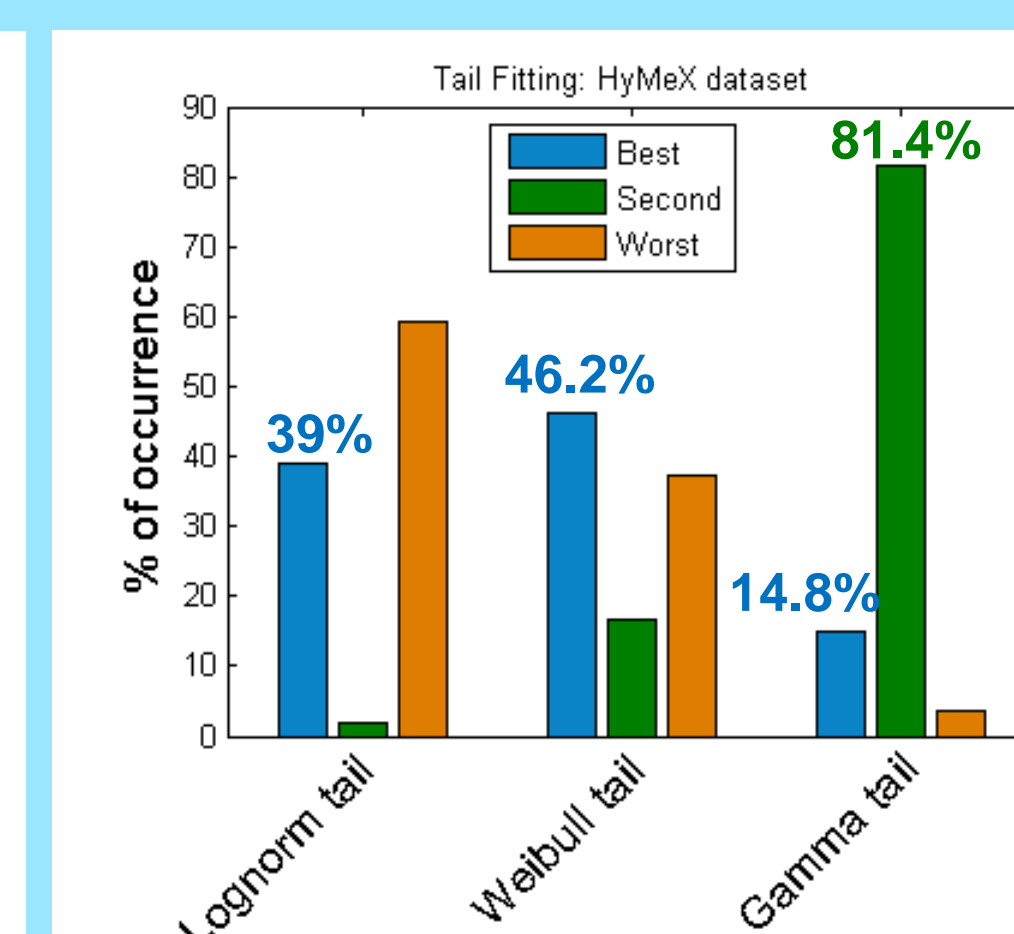
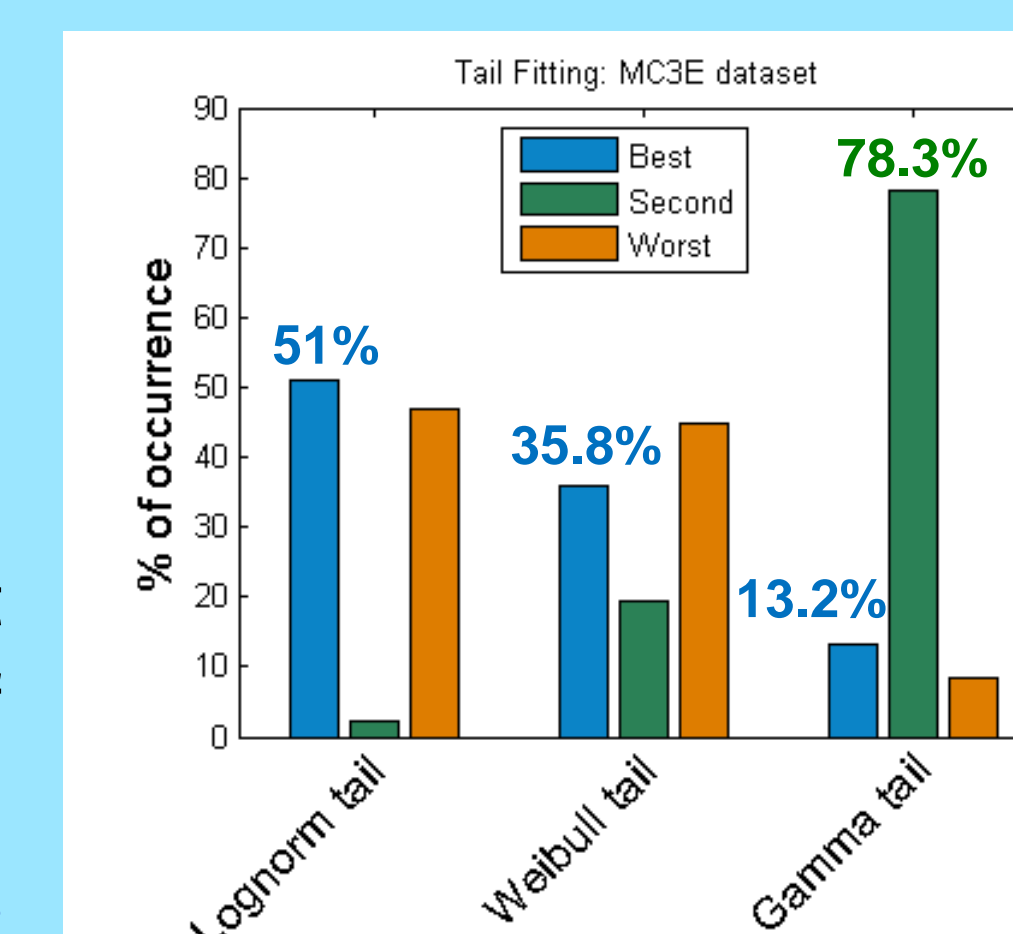
- the Lognormal distribution is the best fit for MC3E
- the Weibull distribution is the best fit for HyMex
- the percentages of spectra better fitted by the Gamma distribution decrease with respect to the whole fitting
- when the Gamma tail distribution is not the best, it is the second one with a very high percentage of success

The results suggest that the **most of the disdrometer measured drop spectra** can be described with a **light-tailed distribution**.

The **practical implication** is that in a light-tailed distribution the larger drops are predicted less frequent than in a heavy-tailed distribution.

Upper tail Fitting

PERFORMANCE OF THE FITTINGS



Heavy tail \rightarrow Very light tail \rightarrow Light tail

Note that when the Weibull distribution performs better the shape parameters are greater than 1.