



HAL
open science

A two-dimensional stochastic rainfall simulator

Nawal Akrou, Cécile Mallet, Laurent Barthès, Aymeric Chazottes

► **To cite this version:**

Nawal Akrou, Cécile Mallet, Laurent Barthès, Aymeric Chazottes. A two-dimensional stochastic rainfall simulator. EGU General Assembly 2015, Apr 2015, Vienna, Austria. pp.id.9488, 2015. insu-01215759

HAL Id: insu-01215759

<https://insu.hal.science/insu-01215759>

Submitted on 19 Oct 2015

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

A two-dimensional stochastic rainfall simulator

Context

- The Precipitations are due to complex meteorological phenomena
- Precipitations can be described as intermittent processes.
- The spatial and temporal variability of this phenomenon is significant and covers large scales

goal

- adapt a one-dimensional time series model previously developed by the authors [Akrou et al., 2015] to a two-dimensional rainfall generator.
- Simulate realistic radar rain maps.

Results

Comparison : one year observed rain maps / four years simulated rain maps

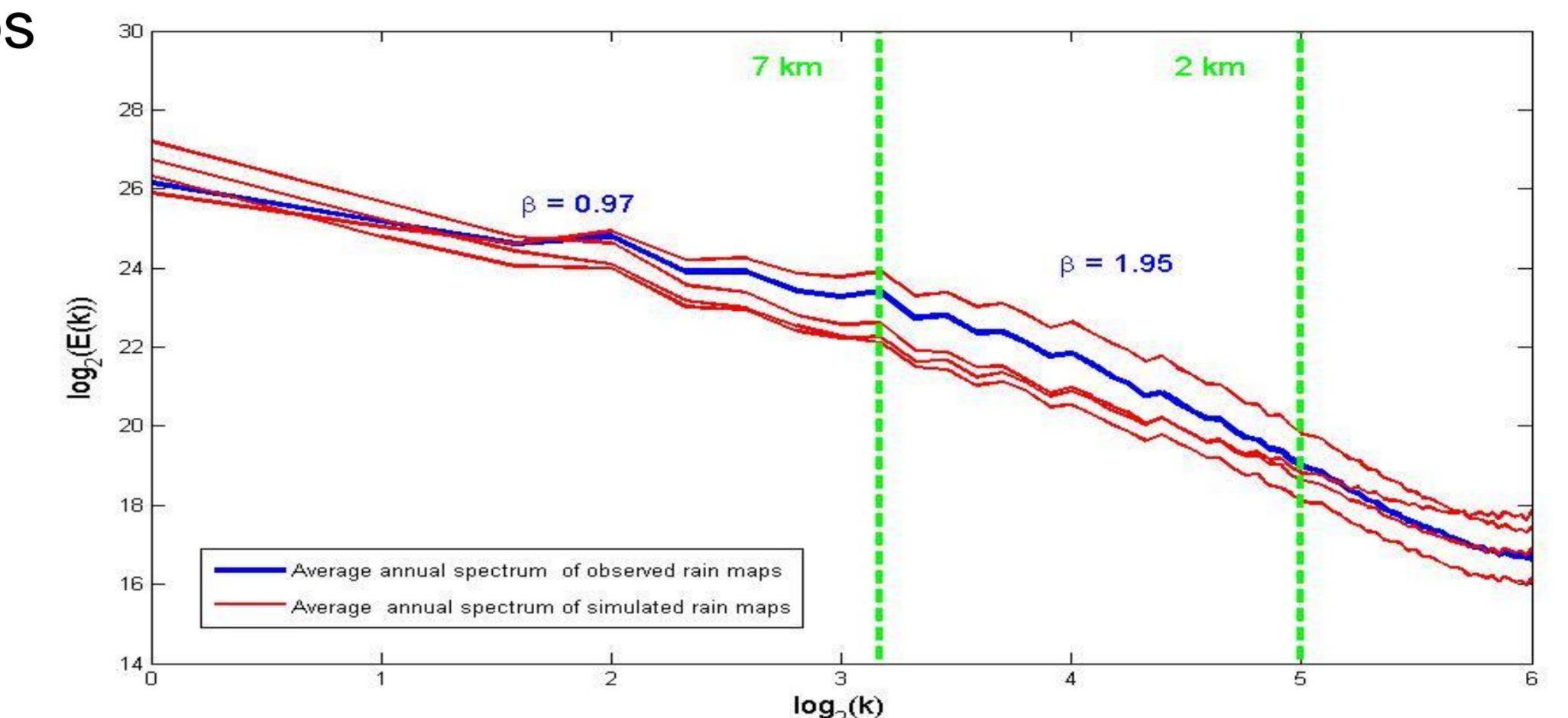


Figure6: Average power spectrum of observed radar rain maps and 4 years simulated rain maps.

| | β in the regime of scaling 2km - 7 km | | | | | β in the regime of scaling 7km - 130 km | | | | |
|-----------------------|---|------|------|------|------|---|------|------|------|------|
| | Mean | std | Q1 | Q2 | Q3 | mean | Std | Q1 | Q2 | Q3 |
| Observation (2012) | 1.95 | 0.7 | 1.50 | 1.88 | 2.30 | 0.97 | 0.5 | 0.33 | 0.80 | 1.15 |
| simulation 1 (1 year) | 1.80 | 0.8 | 1.30 | 1.96 | 2.28 | 0.80 | 0.6 | 0.32 | 0.75 | 1.10 |
| Simulation 2 (1 year) | 1.85 | 0.78 | 1.28 | 1.97 | 2.34 | 0.86 | 0.52 | 0.32 | 0.77 | 1.12 |
| Simulation 3 (1 year) | 1.93 | 0.8 | 1.41 | 1.90 | 2.31 | 0.94 | 0.6 | 0.33 | 0.77 | 1.13 |
| Simulation 4 (1 year) | 1.87 | 0.72 | 1.35 | 1.91 | 2.32 | 0.95 | 0.55 | 0.32 | 0.80 | 1.14 |

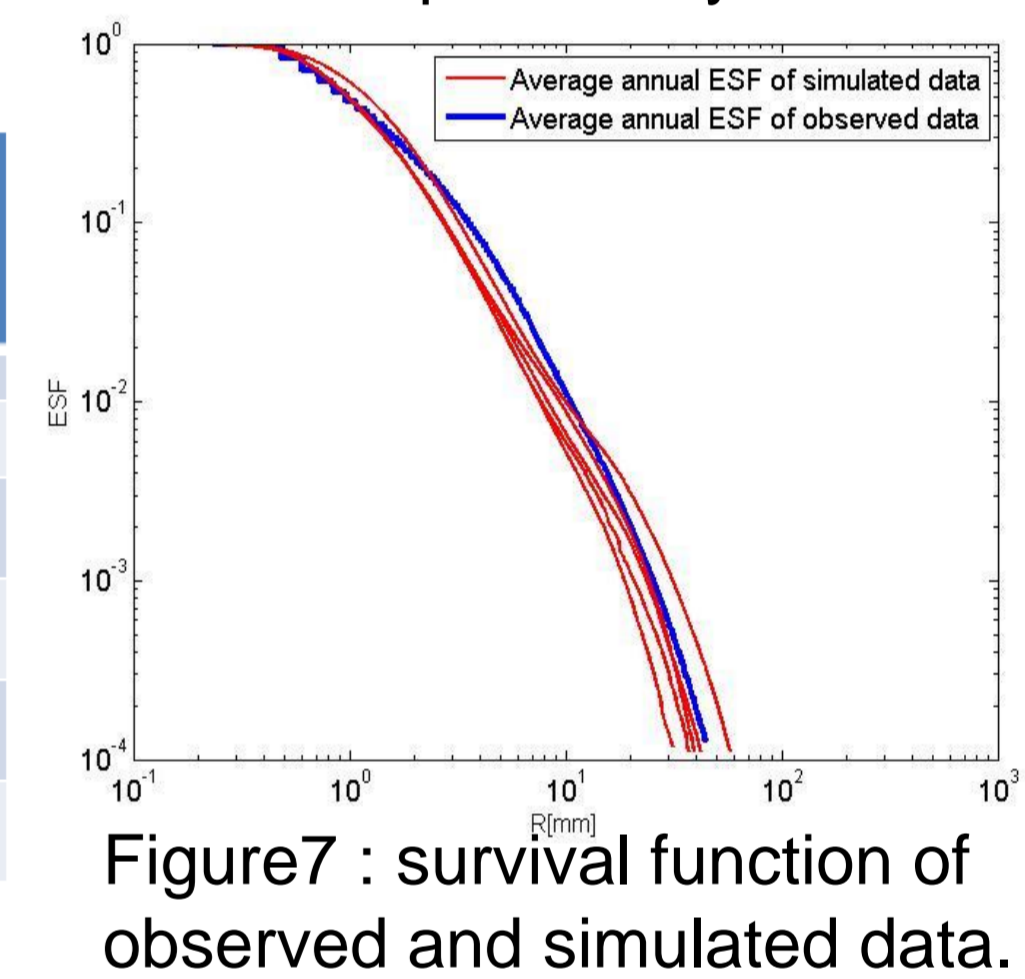


Figure7 : survival function of observed and simulated data.

Table 1 : Statistical characteristic of spectrum slopes

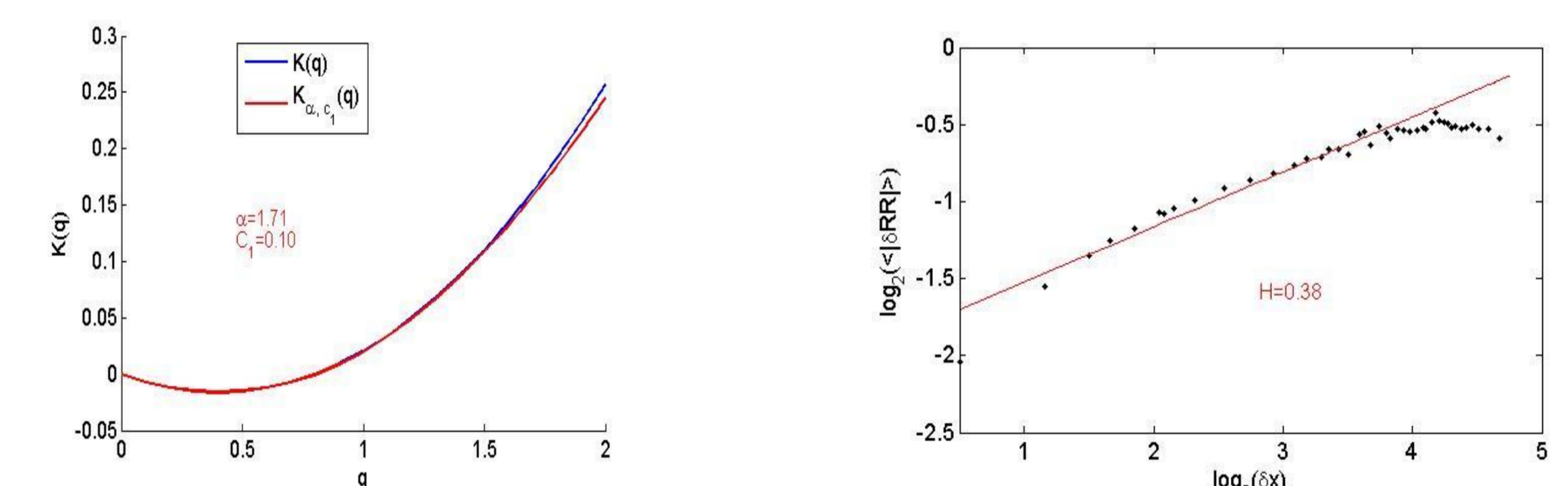


Figure 8 : multifractal analysis of simulated events.

Data

MeteoFrance meteorological RADAR:

location : Trappes near Paris
Spatial resolution : 1 km
Temporal resolution : 5 min
Area : 130 km x 130 km
Date : year 2012 (103 200 rain maps)

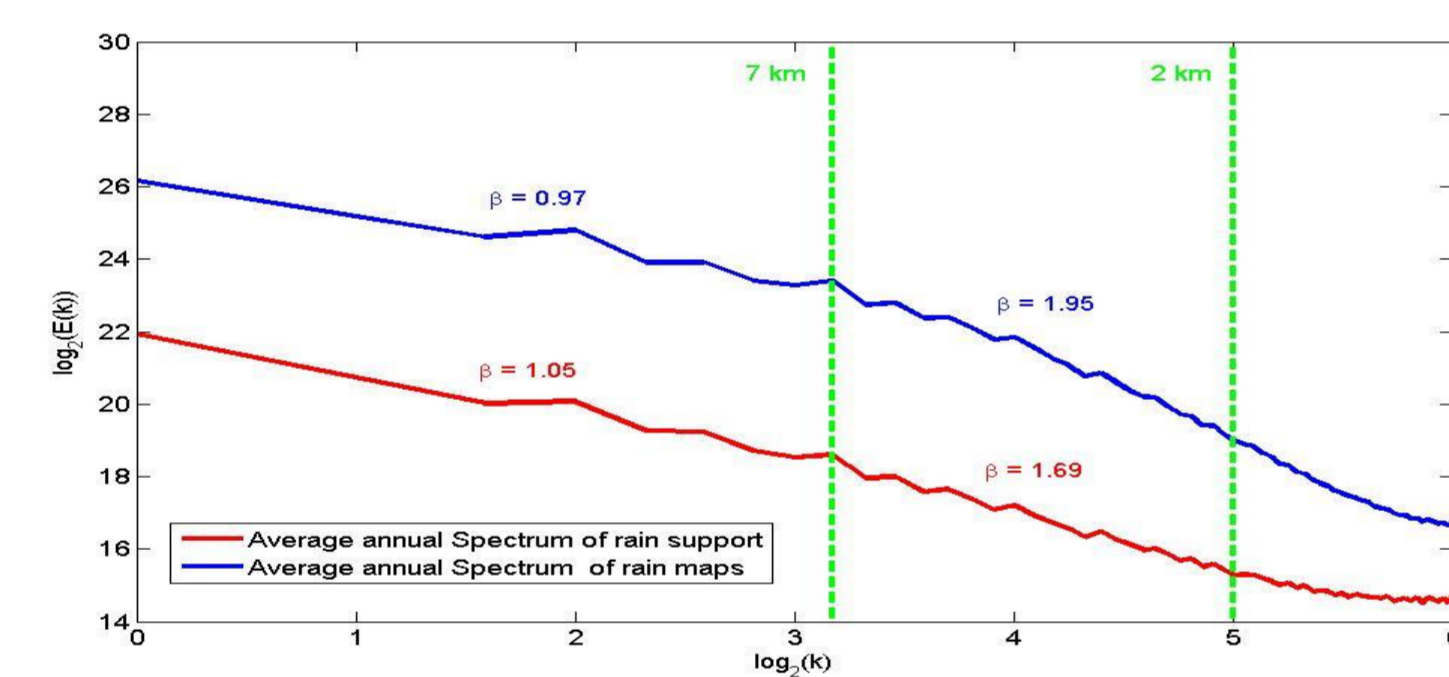


Figure1: Average power spectrum of observed radar rain maps and the corresponding rain supports.

Model

1- Simulation of rain support using SIS method

Sequential Indicator simulation

Hypothesis :

- isotropic and stationary
- spherical model

Variogram parameters are estimated on observed rain maps (Figure 2) Obtained parameters.

Simulation parameters:

Range: a uniform [10, 100]
Still: $c=0.17$
Nugget: $c_0=0$
occurrence probability : μ Uniform [0.15, 1]

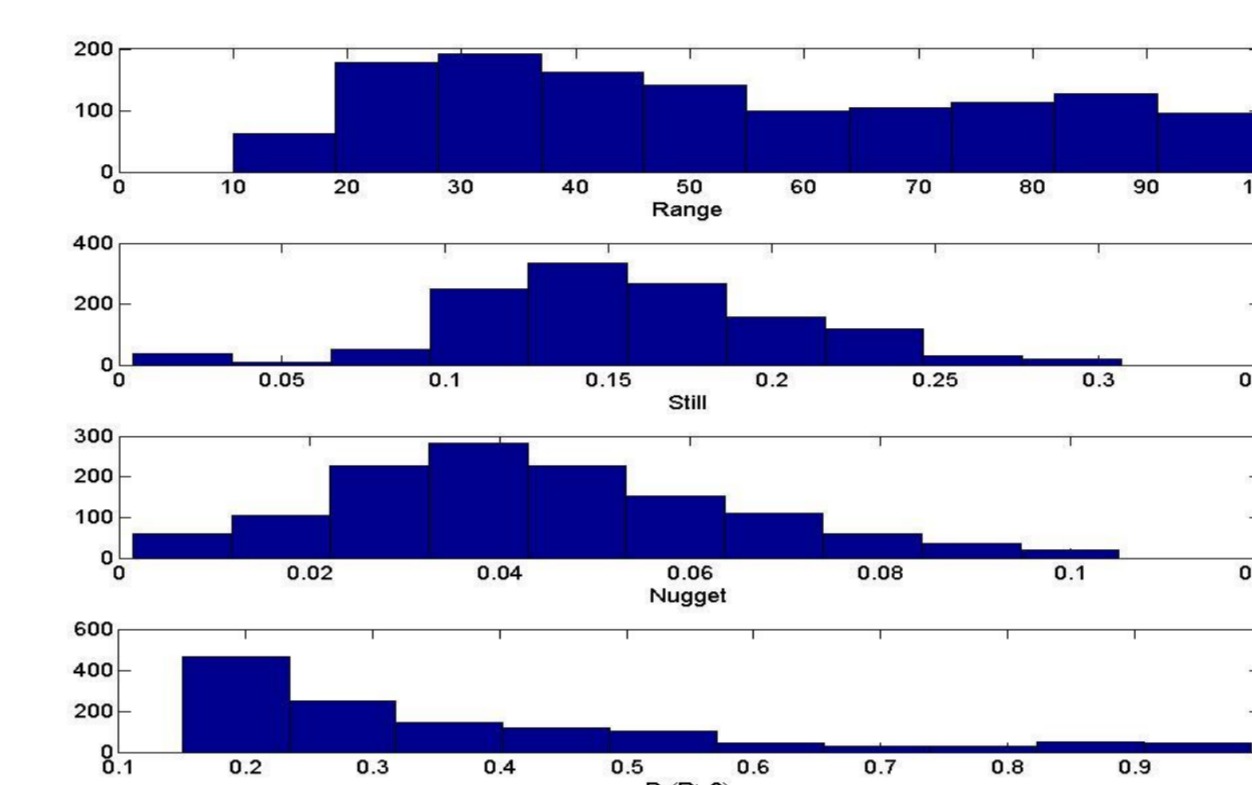


Figure2: Histogram of parameters estimated on observed rain maps

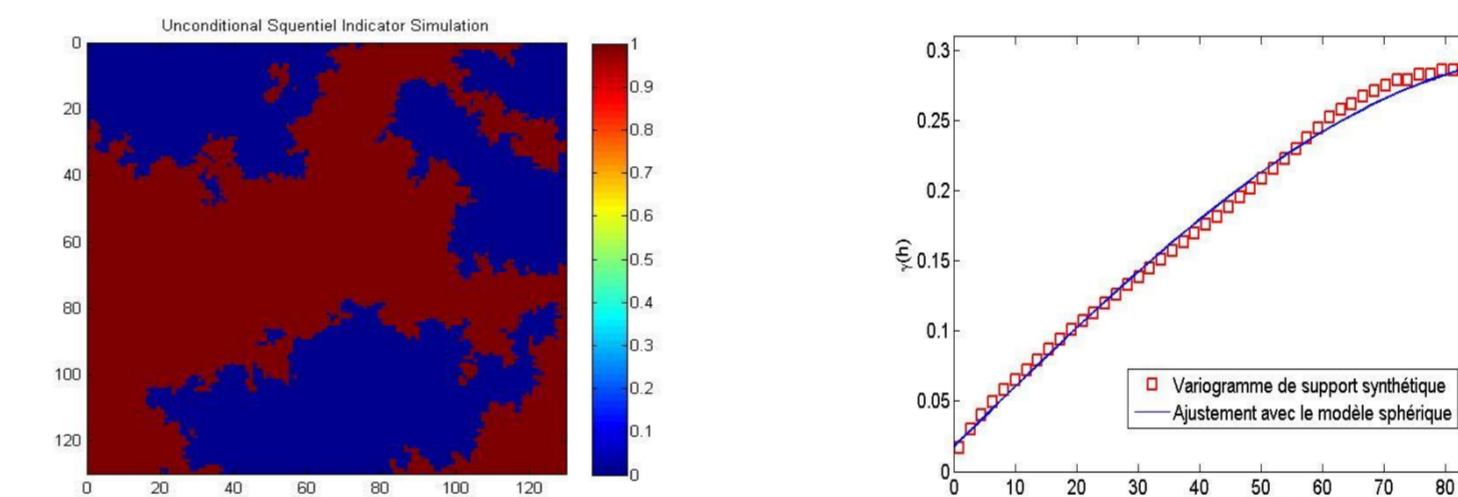
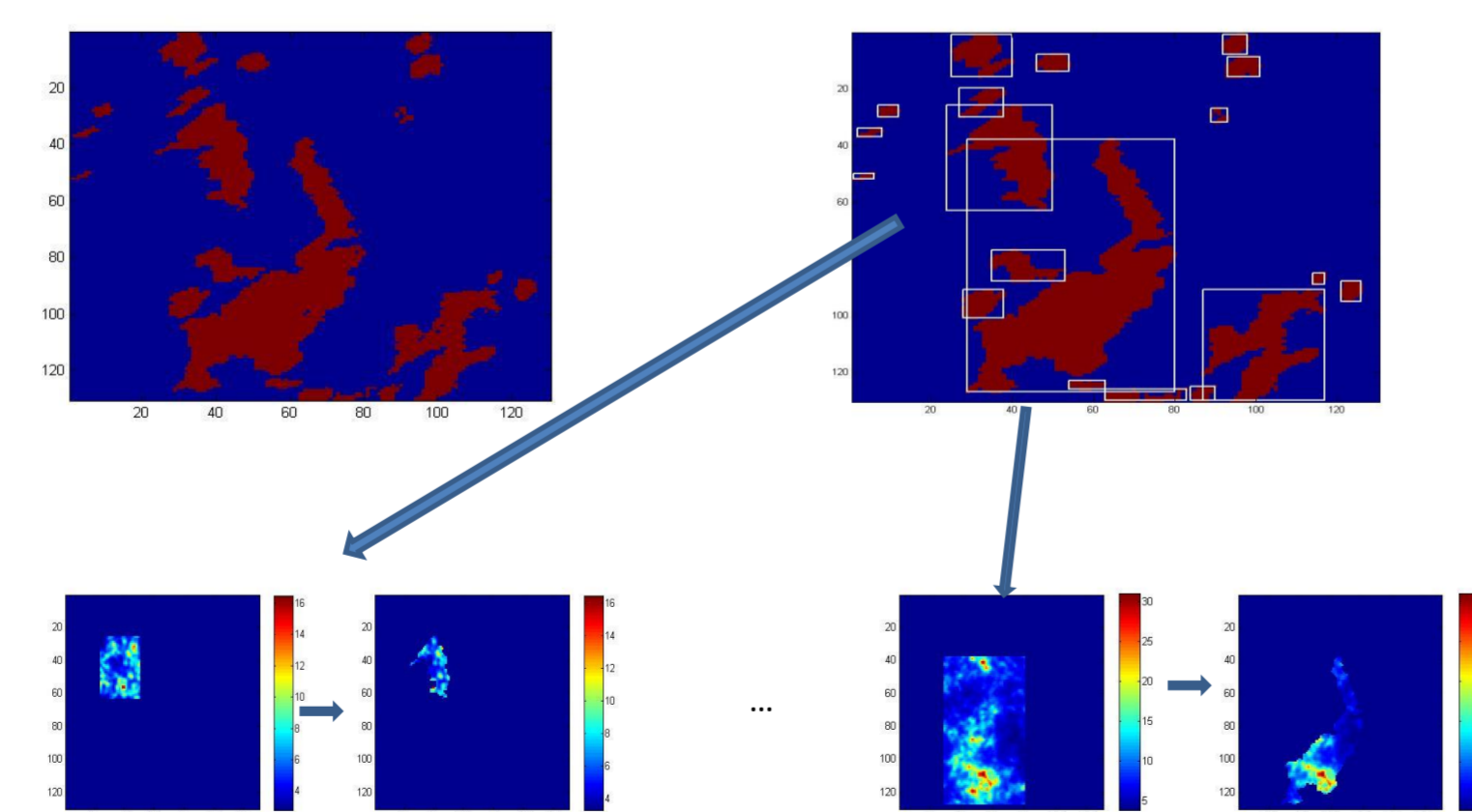


Figure3: example of simulated rain support with parameters ($\mu=0.55$, $a=86.44$, $c=0.17$ et $c_0=0$) and its variogramme

2- Simulation of rain events using multifractal model FIF

FIF (Schertzer and lovejoy, 1987)

parameters: $\alpha=1.60$, $C_1=0.10$ et $H=0.40$



4- Transition processing

$F(z)=\min(\|z-y\|)$ normalized and $\epsilon \in [0,1]$
 $z \in \{x, I(x)=1\}$ inspired (Shleiss et al. 2014)
 $y \in \{x, I(x)=0\}$ $I(x)$ Rain support

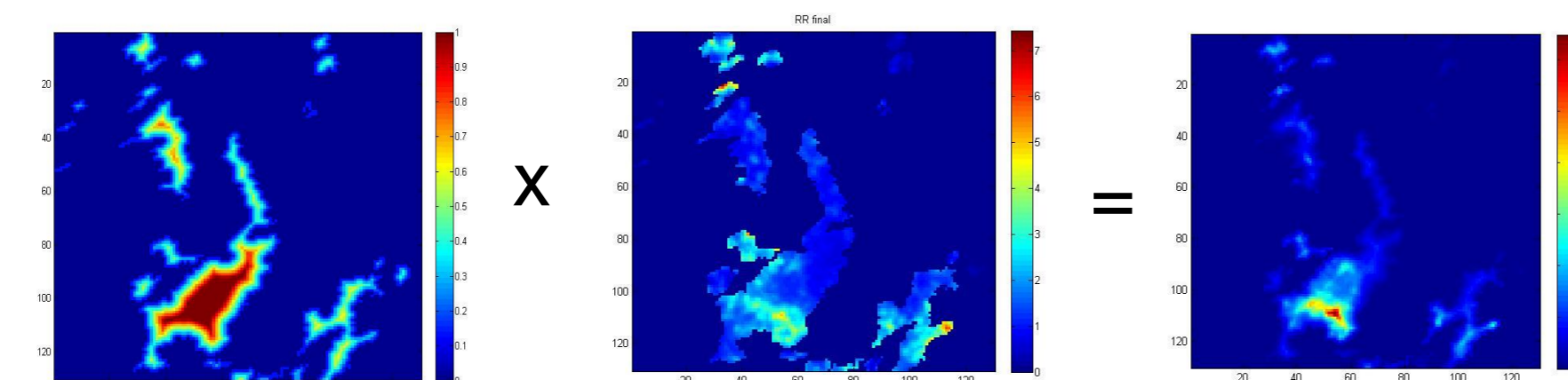
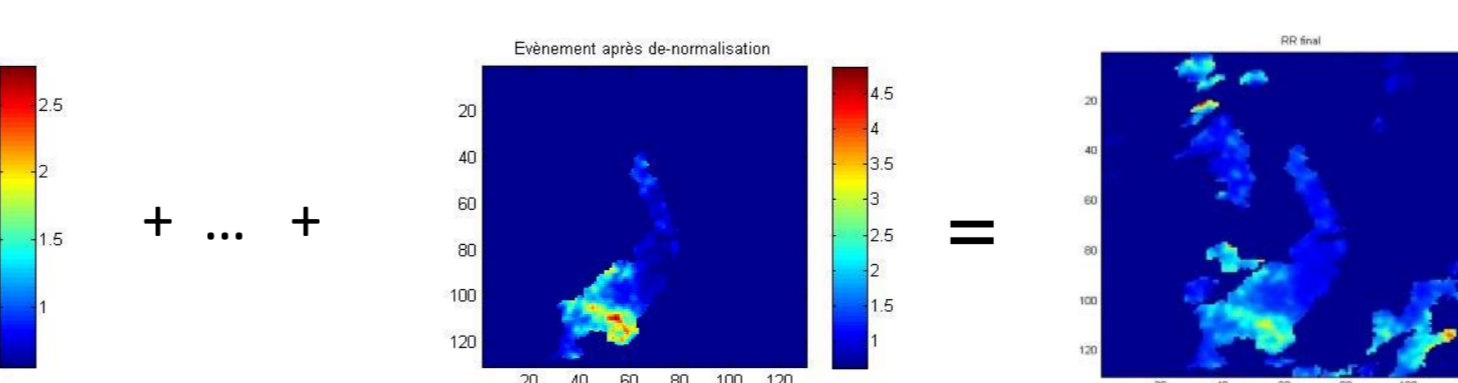


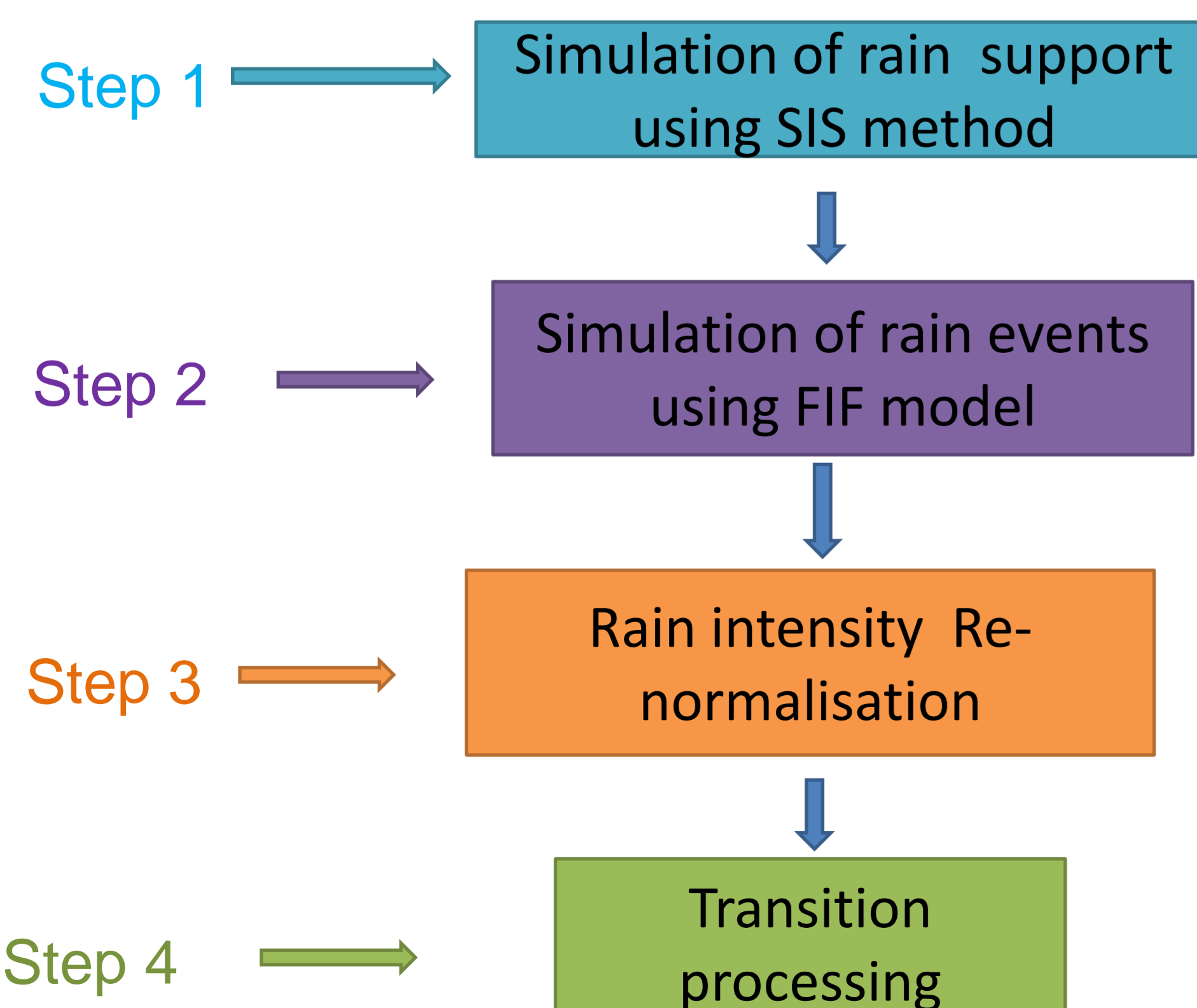
Figure4:parameters estimated on mean rain rate knowing events surface

Figure5: mean events rain rate as function of events surface



Methodology

- Based on sequential steps:



3- Renormalization

Following relationship is used:

$$RR = \frac{RR}{Mean(RR_i)} RR_m$$

where RR_m are randomly drawn following alpha stable distribution

alpha stable distribution parameters are estimated on 330 346 observed rain events (Figure 3).

Obtained parameters: $\alpha=1.18$; $\beta=1$; $\gamma=0.03$ s

$\mu = 0,09$ s $+0,65$

where s is the event surface

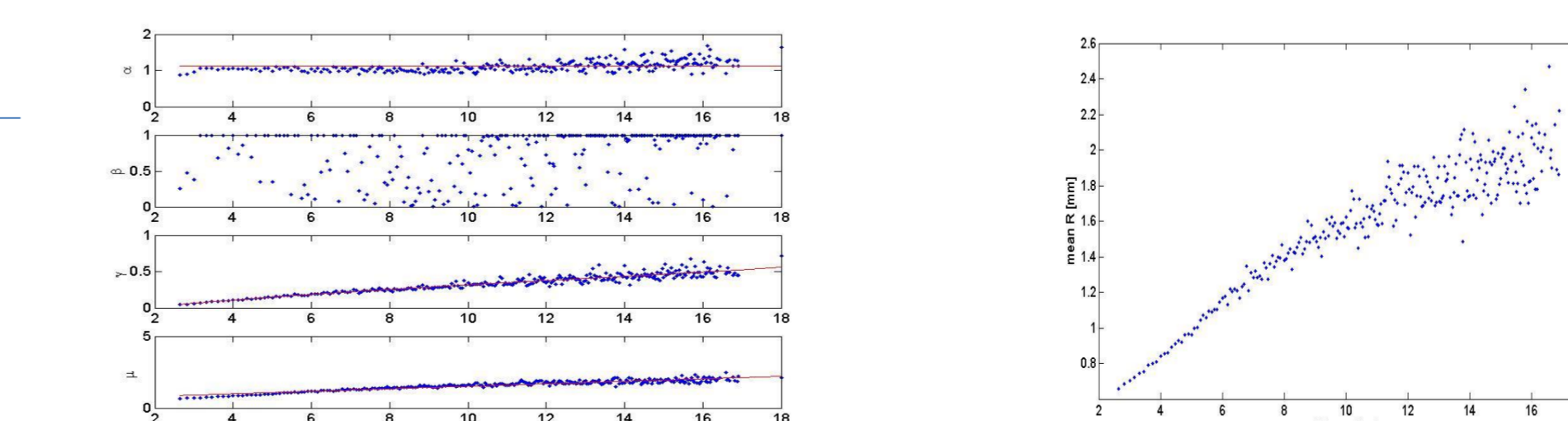


Figure4:parameters estimated on mean rain rate knowing events surface

Figure5: mean events rain rate as function of events surface

conclusion

- generalization of the one-dimensional model (akrou and al. 2015) to a two-dimensional model.
- the simulated two-dimensional fields look realistic, they moreover have coherent statistical properties (cumulative rain rate distribution, power spectrum and structure function) with observed one.
- The proposed simulation processes is very general and can be adapted to any climatic area

references

- Akrou, N., Aymeric, C., Verrier, S., Mallet, C., Barthes, L.: 2015: Simulation of yearly rainfall time series at micro-scale resolution with actual properties: intermittency, scale invariance, rainfall distribution, submitted to Water Resources Research (under revision)
- Schertzer, D., S. Lovejoy, 1987: Physically based rain and cloud modeling by anisotropic, multiplicative turbulent cascades. J. Geophys. Res. 92, 9692-9714
- Schleiss, M., S. Chamoun, and A. Berne (2014), Stochastic simulation of intermittent rainfall using the concept of dry drift, Water Resources Research, 50 (3), 2329-2349