



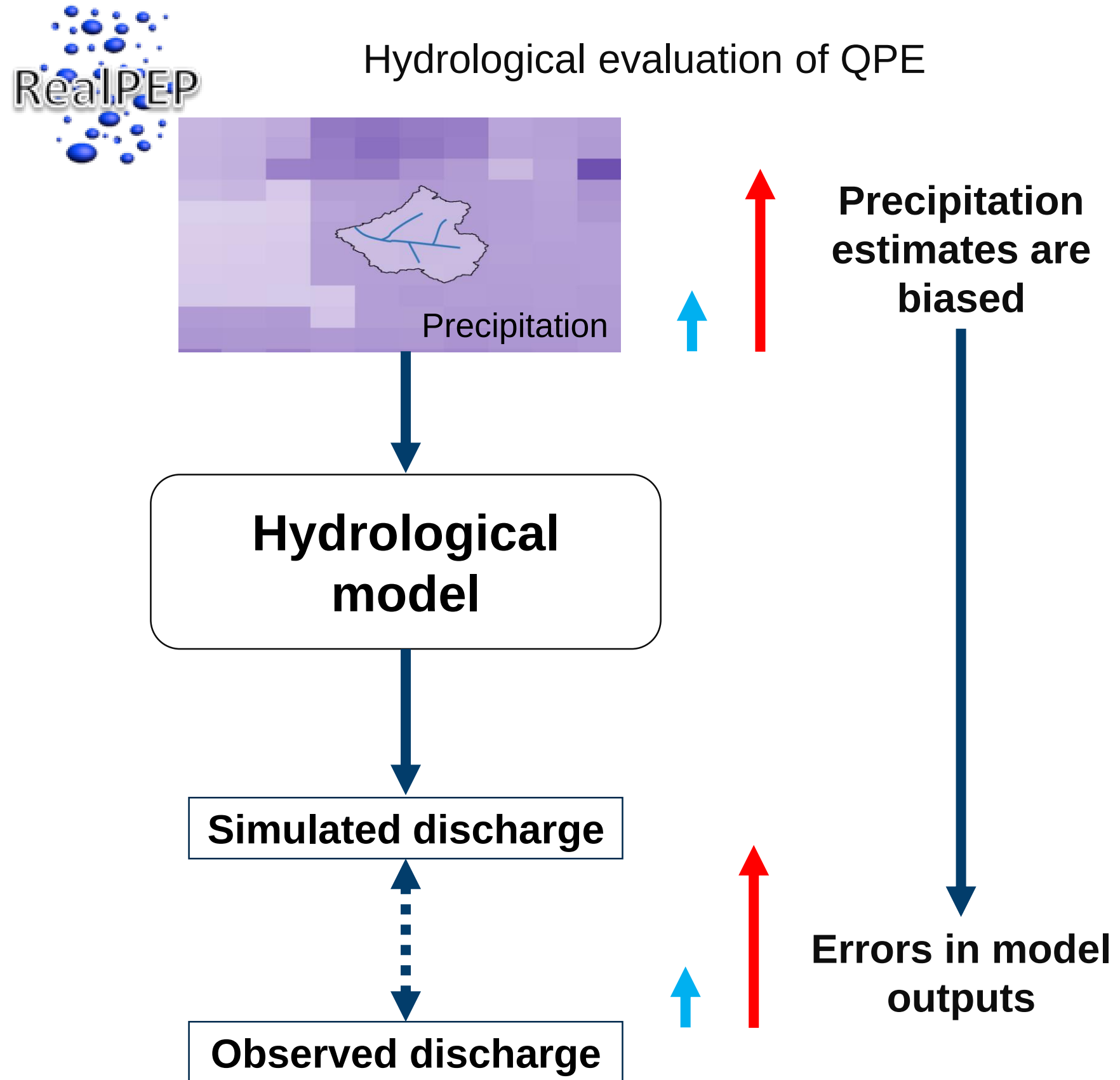
Evaluating quantitative precipitation estimates and hydrological models for simulating the 2021 extreme events in Germany

M. Saadi^{1,2}, C. Furusho-Percot^{1,2}, A. Belleflamme^{1,2}, J.-Y. Chen³, R. Reinoso-Rondinel³, S. Trömel^{3,4}, S. Kollet^{1,2}

¹FZJ/IBG-3, ²Geoverbund ABC/J/HPSC-TerrSys, ³UniBonn/Dept. of Meteorology, ⁴Geoverbund ABC/J/CPEX-Lab

2022-05-25 | EGU General Assembly 2022 | Vienna, Austria

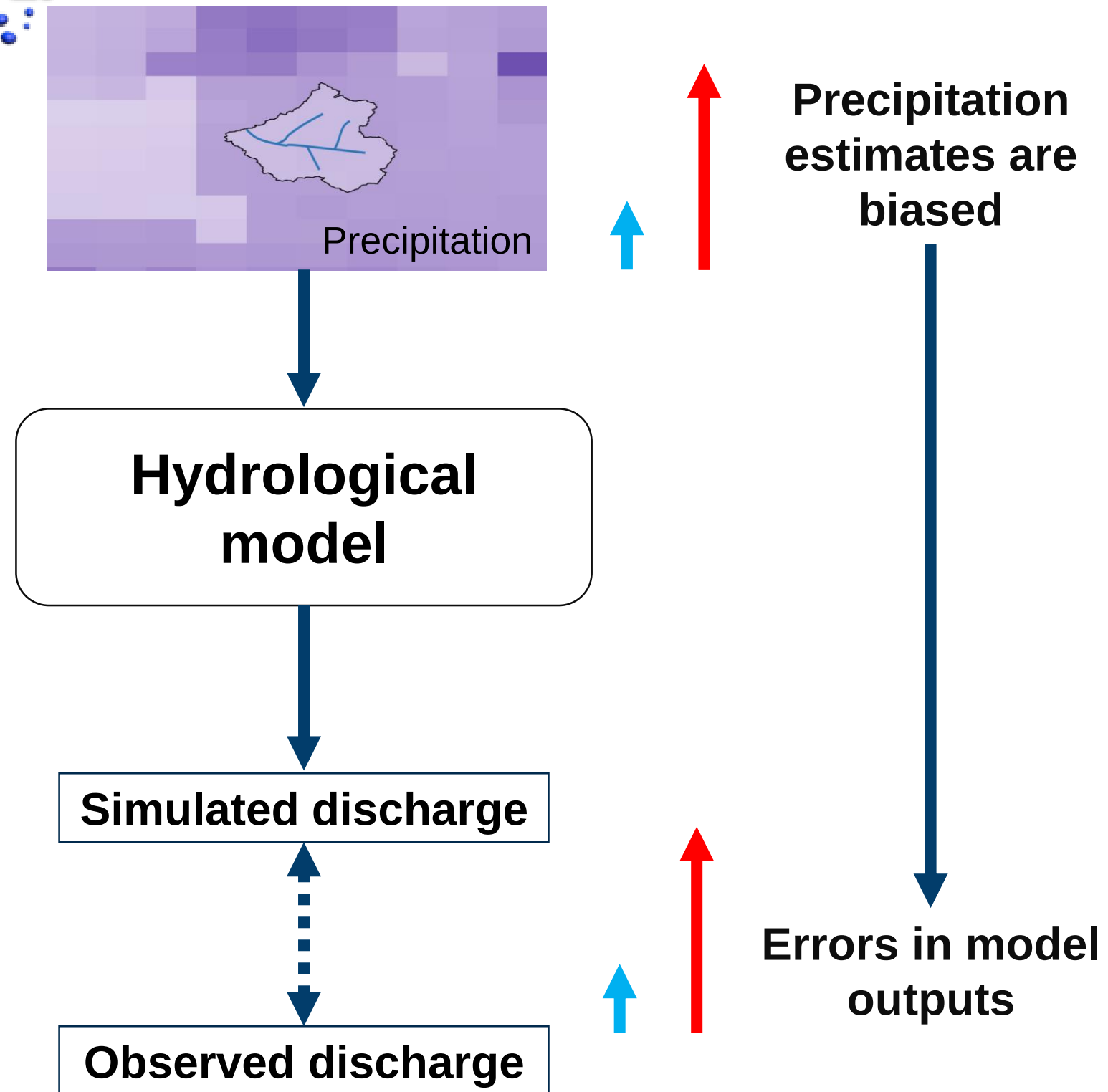
1| Context and objectives



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Hydrological evaluation of QPE



For extreme floods, accurate precipitation estimates are crucial

But event hydrographs are generally unavailable!

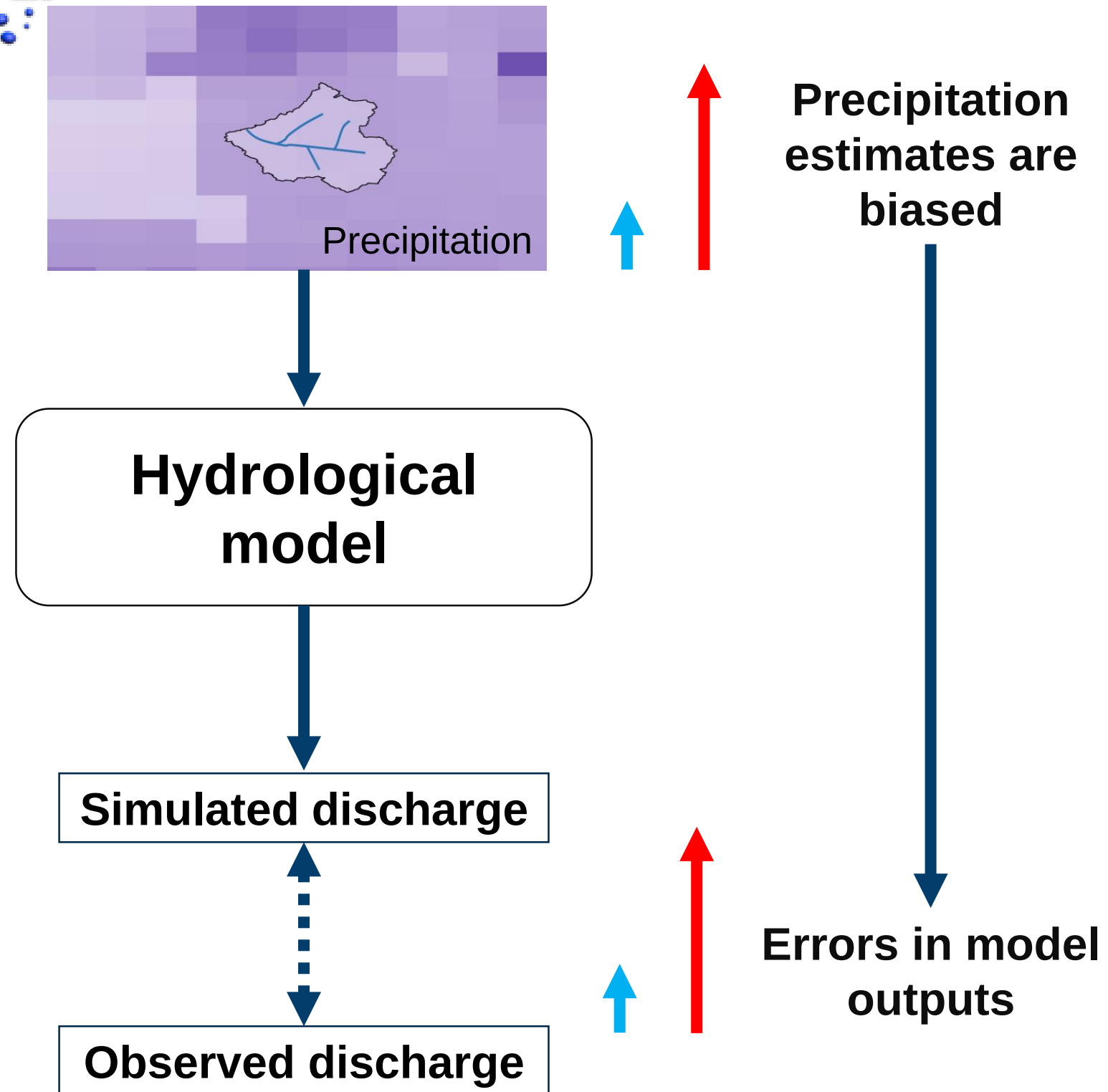


July 2021 events at Altenahr and Erftstadt-Blessem
(source: DW.com)
>200 fatalities, up to € 5.5 Billion in insured losses

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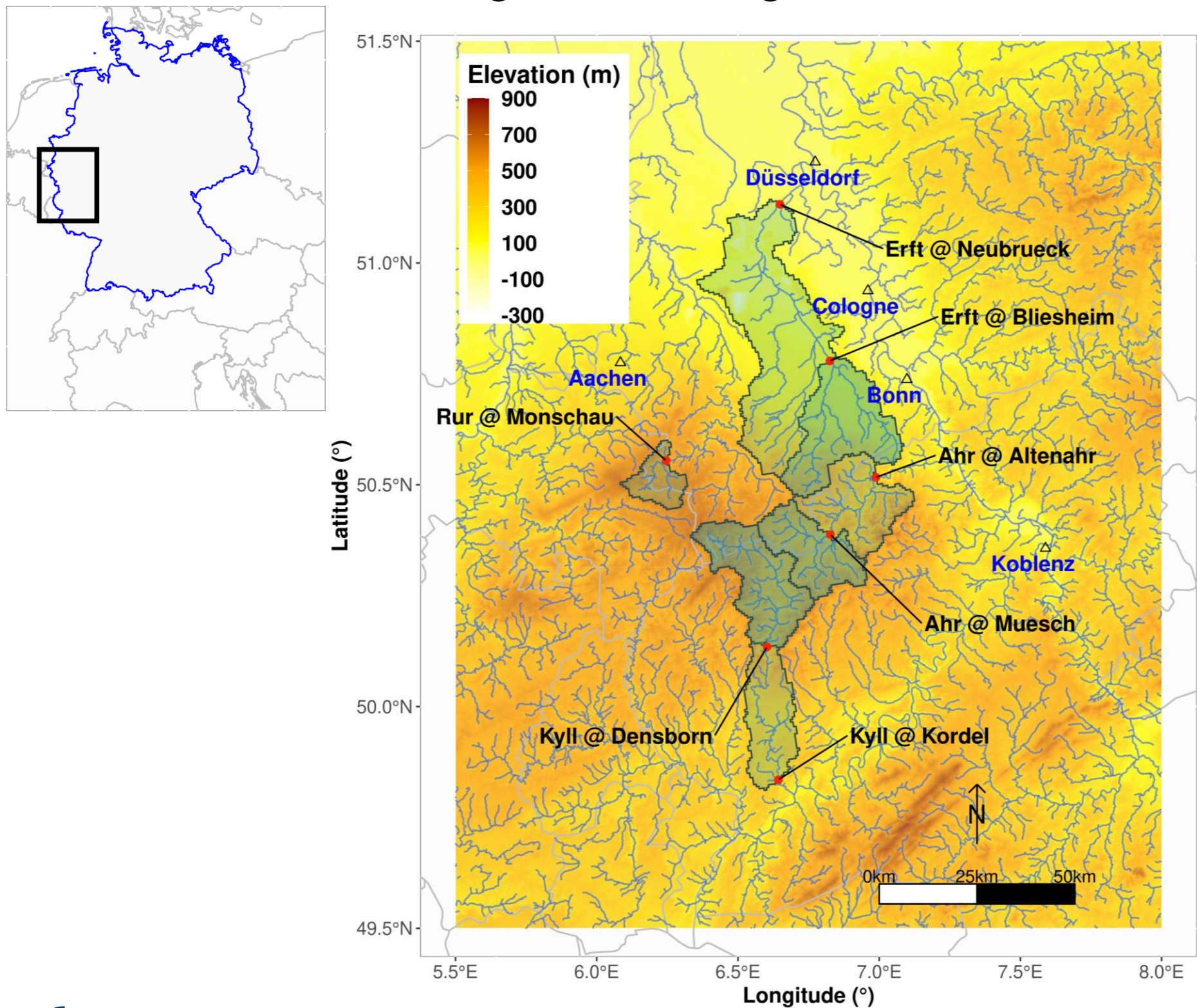
July 2021 events at Altenahr and Erftstadt-Blessem
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Given different precipitation estimates (QPE), what were the chances of exceeding the highest measured peakflow?

2| Catchments, models and data

2.1 | Seven catchments

Draining the Eifel range



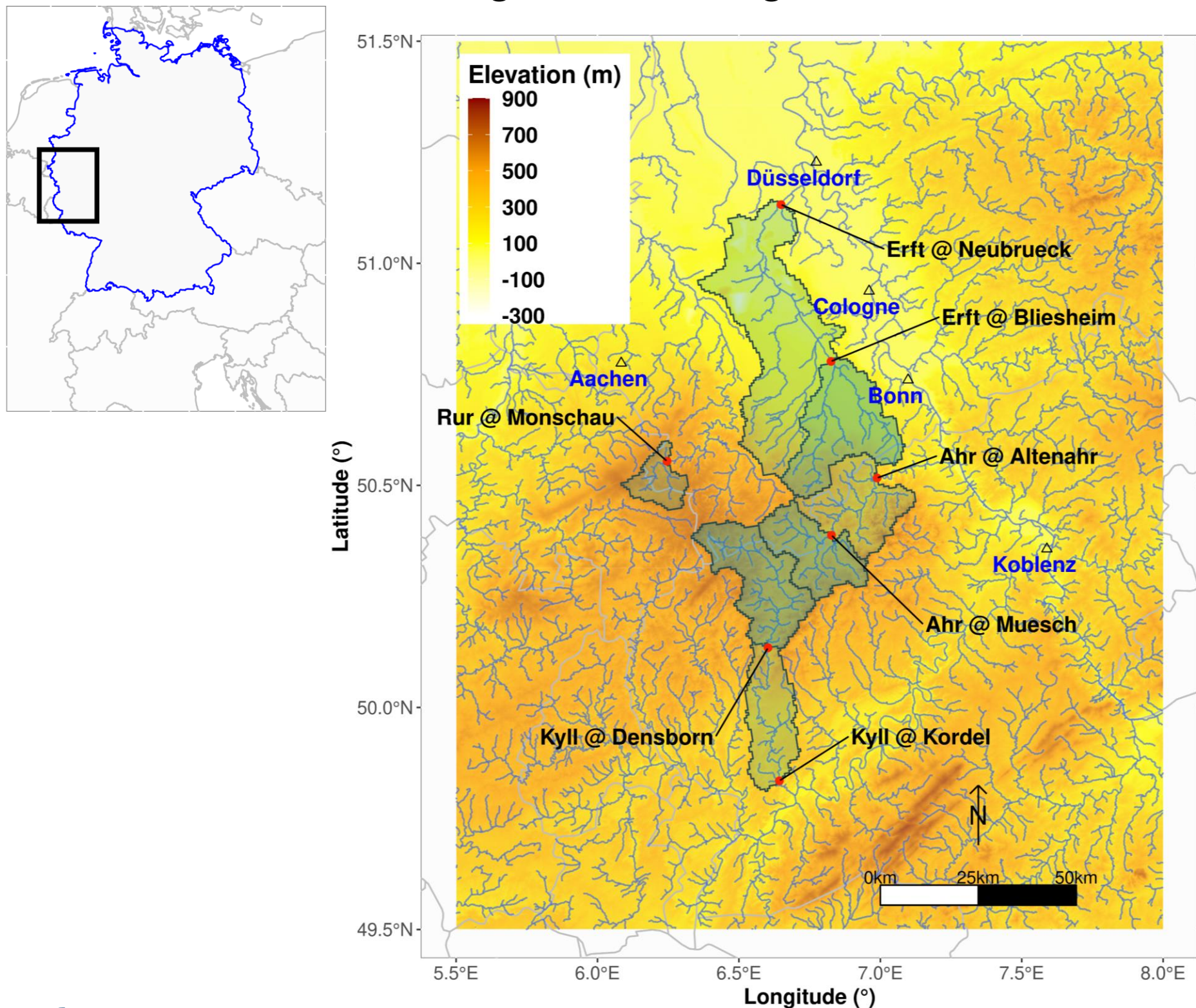
Humid climate, with oceanic influence

Median elevation (m)	150 – 580
Area (km ²)	140 – 1670
Mean precipitation (mm/yr)	700 – 1070
Aridity index (-)	0.52 – 0.89
Mean discharge (mm/yr)	130 – 760

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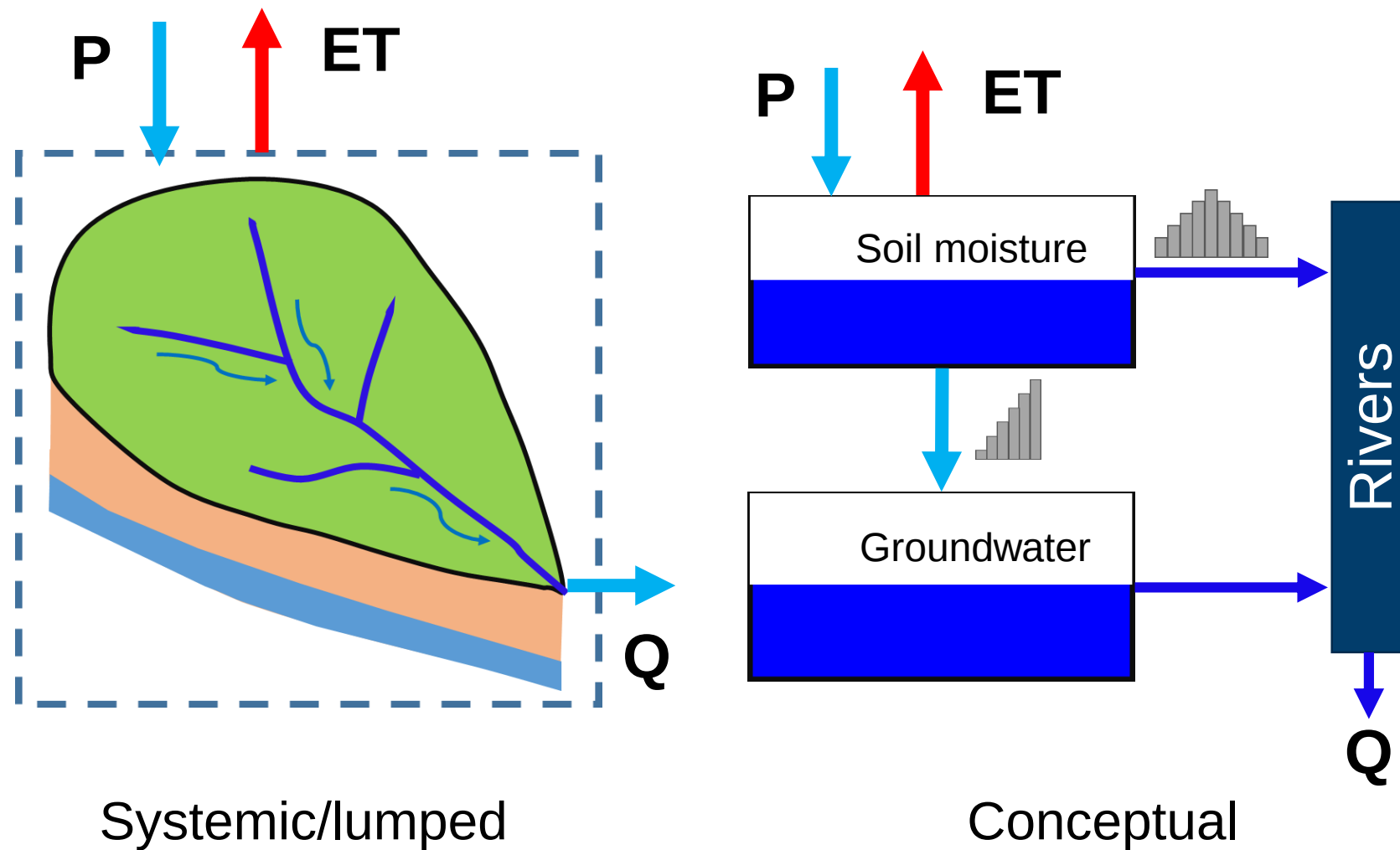
Highest measured peakflows

Catchment	Historical peakflow (m ³ /s)
Ahr @ Altenahr	236
Ahr @ Muesch	132
Erft @ Bliesheim	55.8
Erft @ Neubrueck	46.64
Kyll @ Densborn	180
Kyll @ Kordel	218
Rur @ Monschau	109.63

2| Catchments, models and data

2.2 | Two hourly models

GR4H



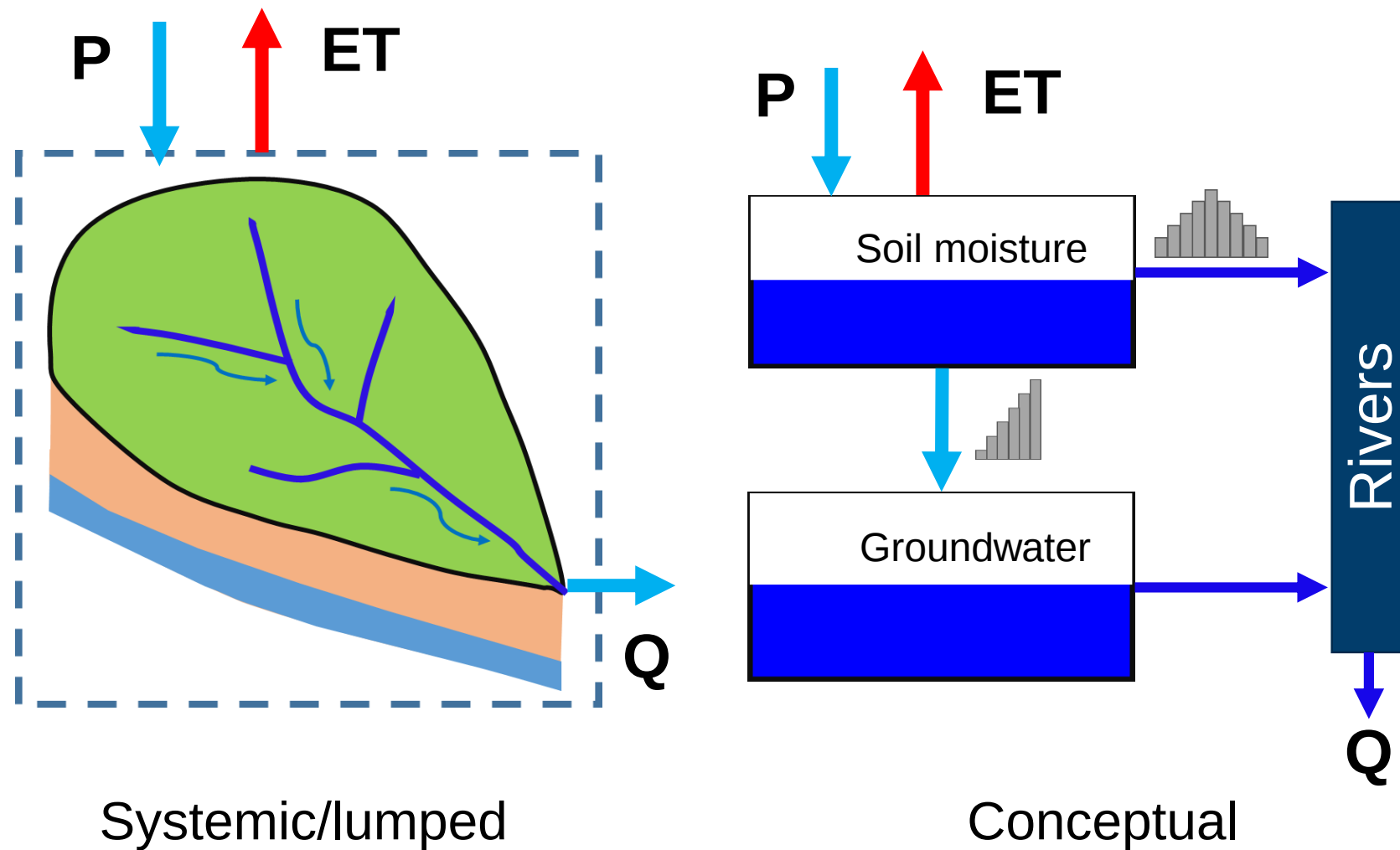
12 optimal parameters, from historical observations

Runs on local computer

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2.2 | Two hourly models

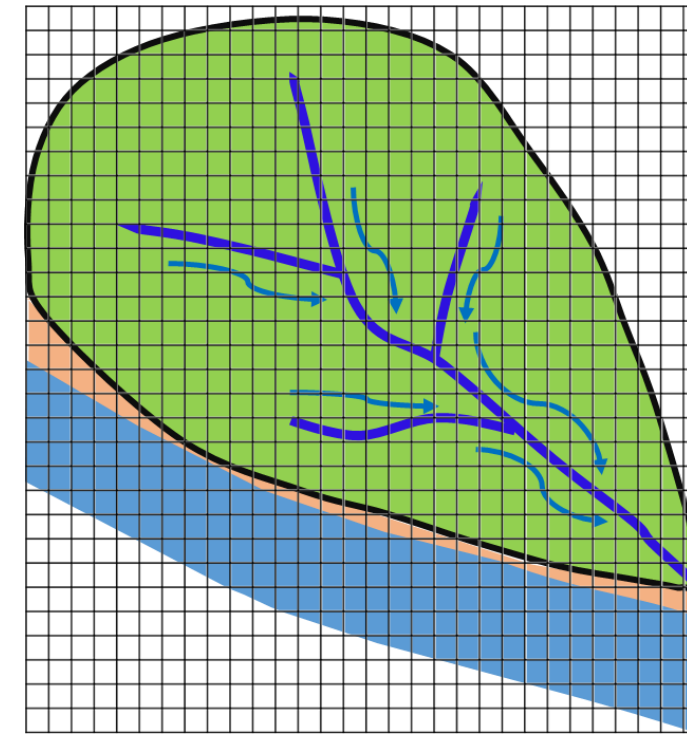
GR4H



12 optimal parameters, from historical observations

Runs on local computer

ParFlowCLM



Reductionist/distributed
(611 m)

Parameters from landscape properties, with 3 Manning's values

Runs on GPUs of the JUWELS HPC system
(4 nodes x 512 GiB)

3-D Richards equation

$$\begin{cases} S_s S_w(p) \frac{\partial p}{\partial t} + \phi \frac{\partial S_w(p)}{\partial t} = \nabla q + q_s \\ q = -k_s k_r(p) \nabla(p - z) \end{cases}$$

Kinematic wave approximation

$$\begin{cases} \frac{\partial h}{\partial t} = \nabla(vh) + q_s \\ S_0 = S_f \end{cases}$$

Mechanistic

2| Catchments, models and data

2.3 | Six QPE products for the 14.07.2021

QPE products

Rain gauges

RADOLAN

Horizontal
reflectivity Z_h

RZH

Horizontal reflectivity
 Z_h + Specific differential
phase K_{DP}

RZHKDP

Specific attenuation A_H
+ Specific differential
phase K_{DP}

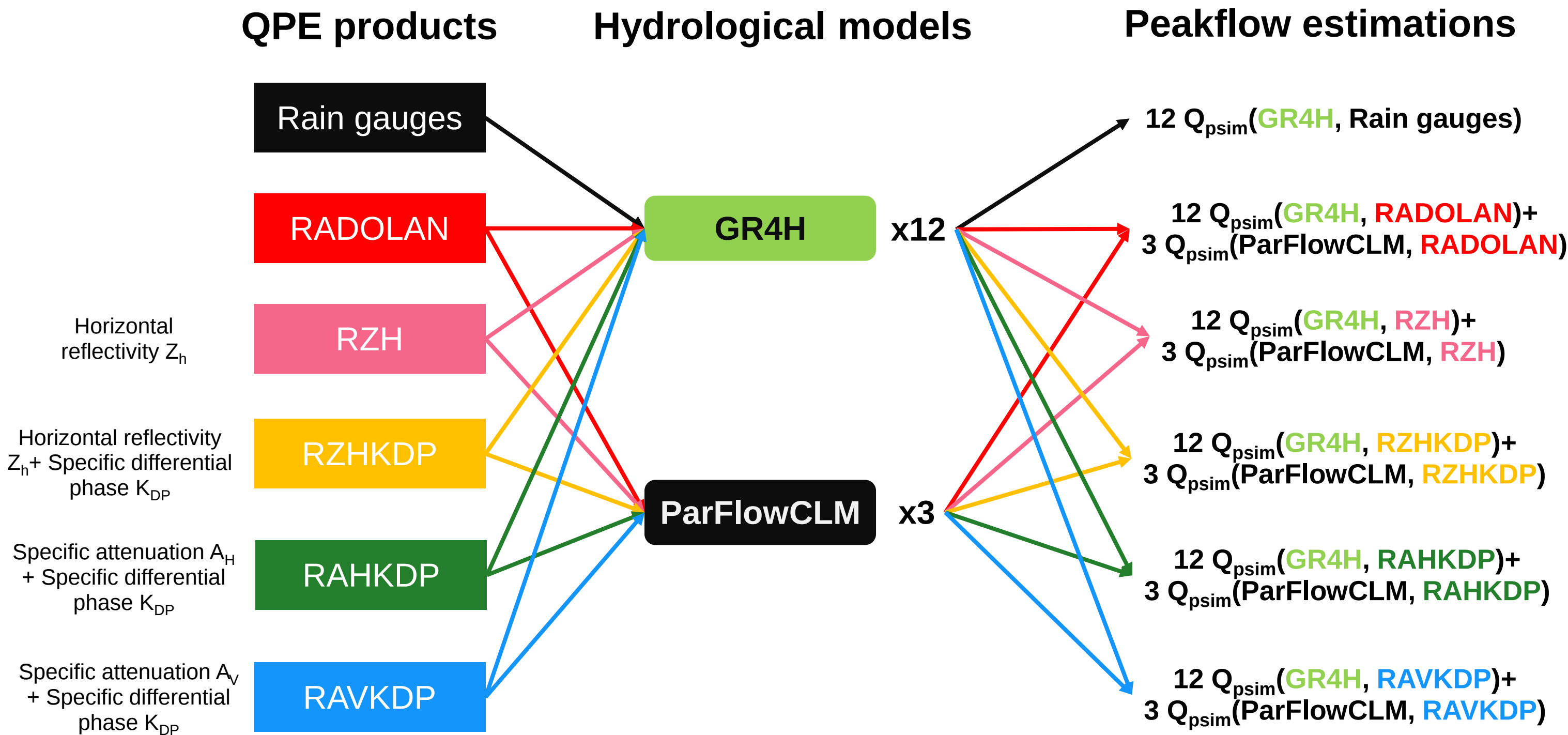
RAHKDP

Specific attenuation A_V
+ Specific differential
phase K_{DP}

RAVKDP

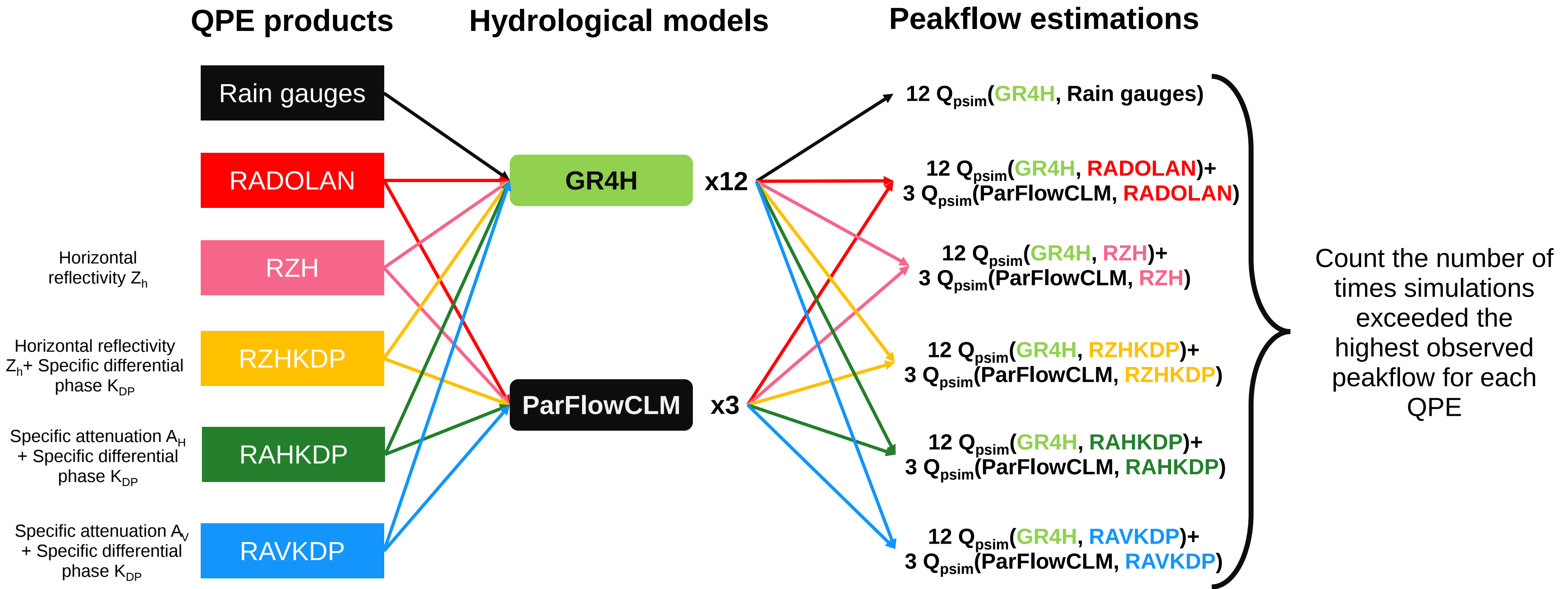
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2.3 | Six QPE products for the 14.07.2021



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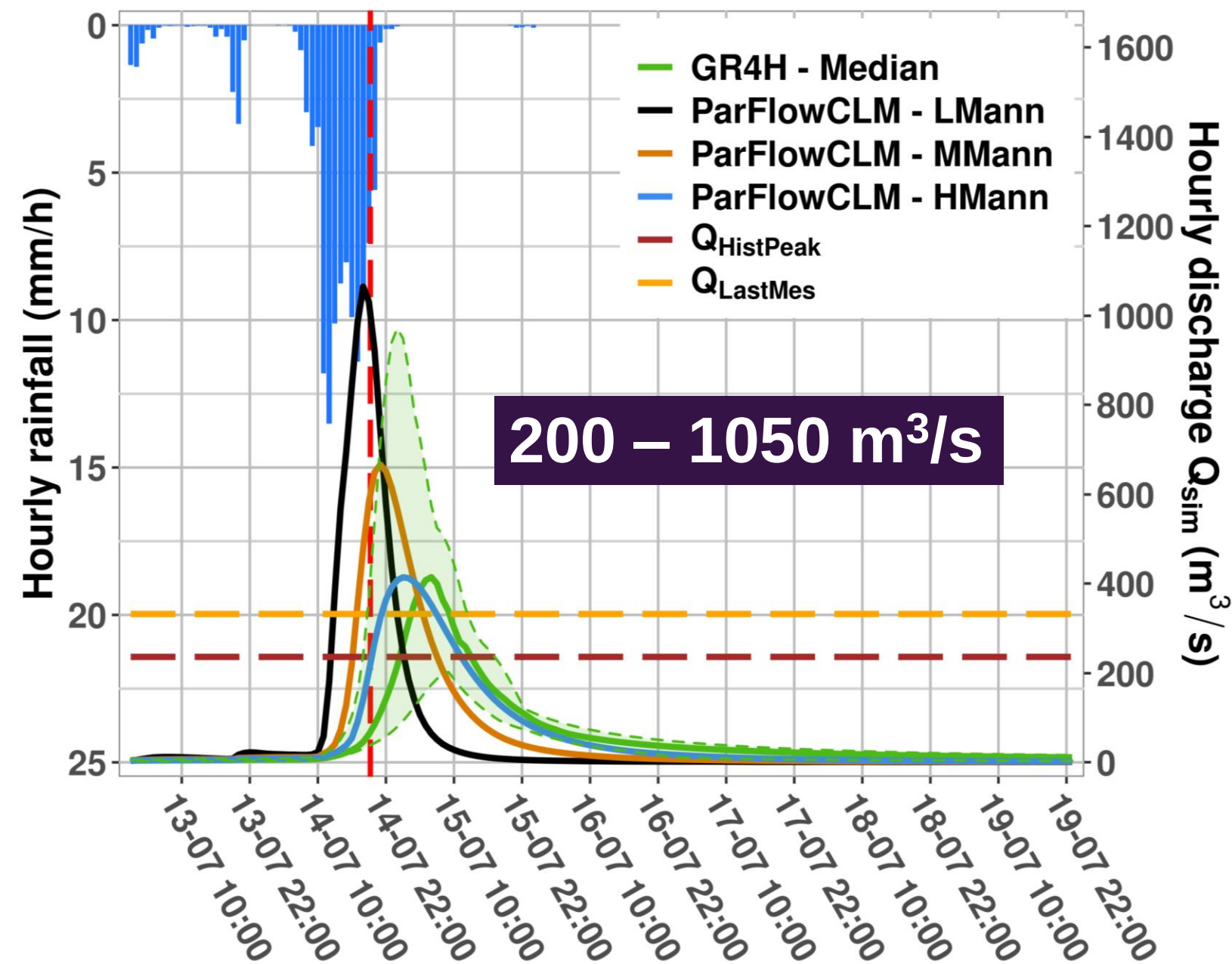


3| Results

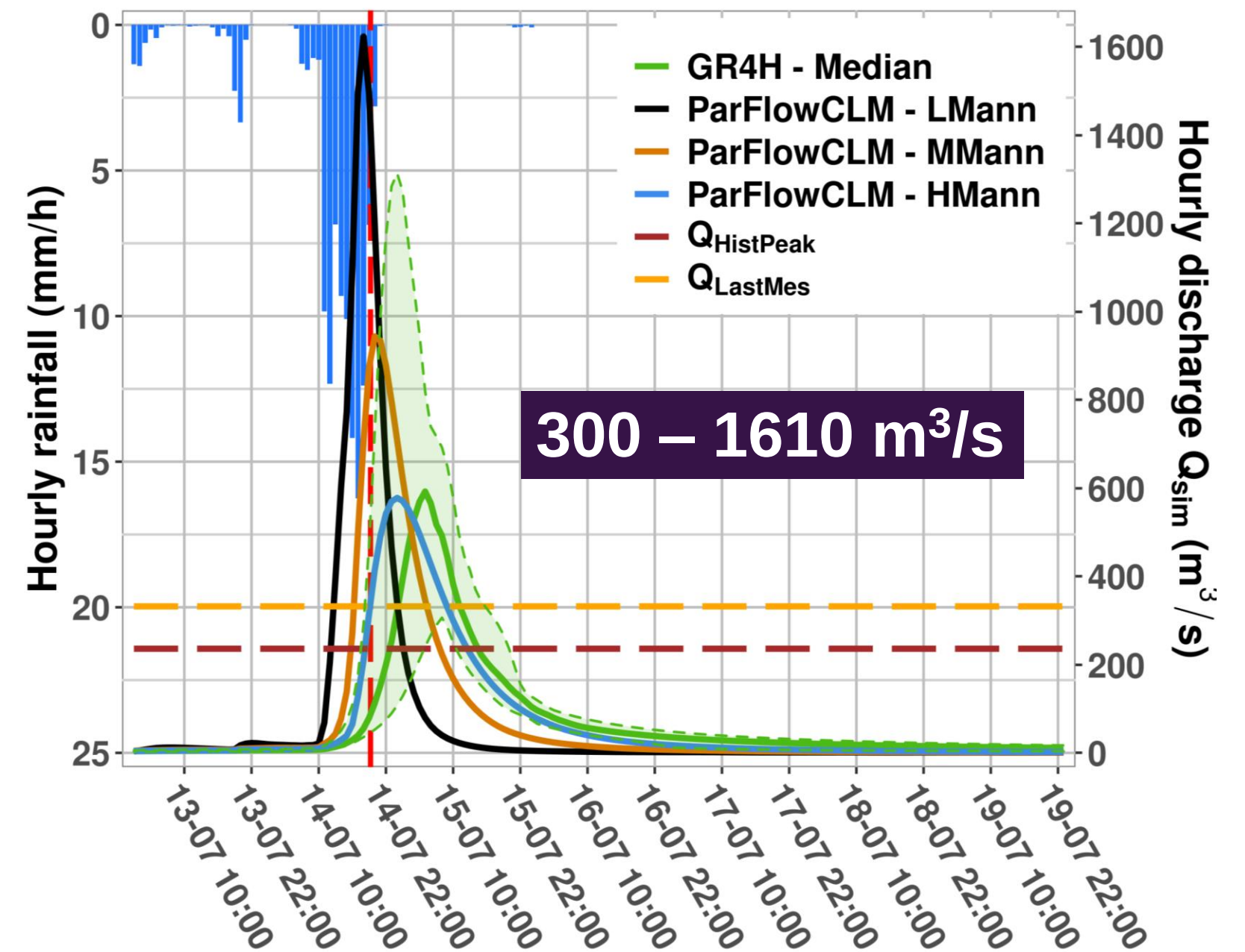
3.1 | Result 1: Uncertainty in peakflow estimation

Ahr @ Altenahr

(a) GR4H vs. ParFlowCLM | QPE: RADOLAN



(b) GR4H vs. ParFlowCLM | QPE: RAVKDP

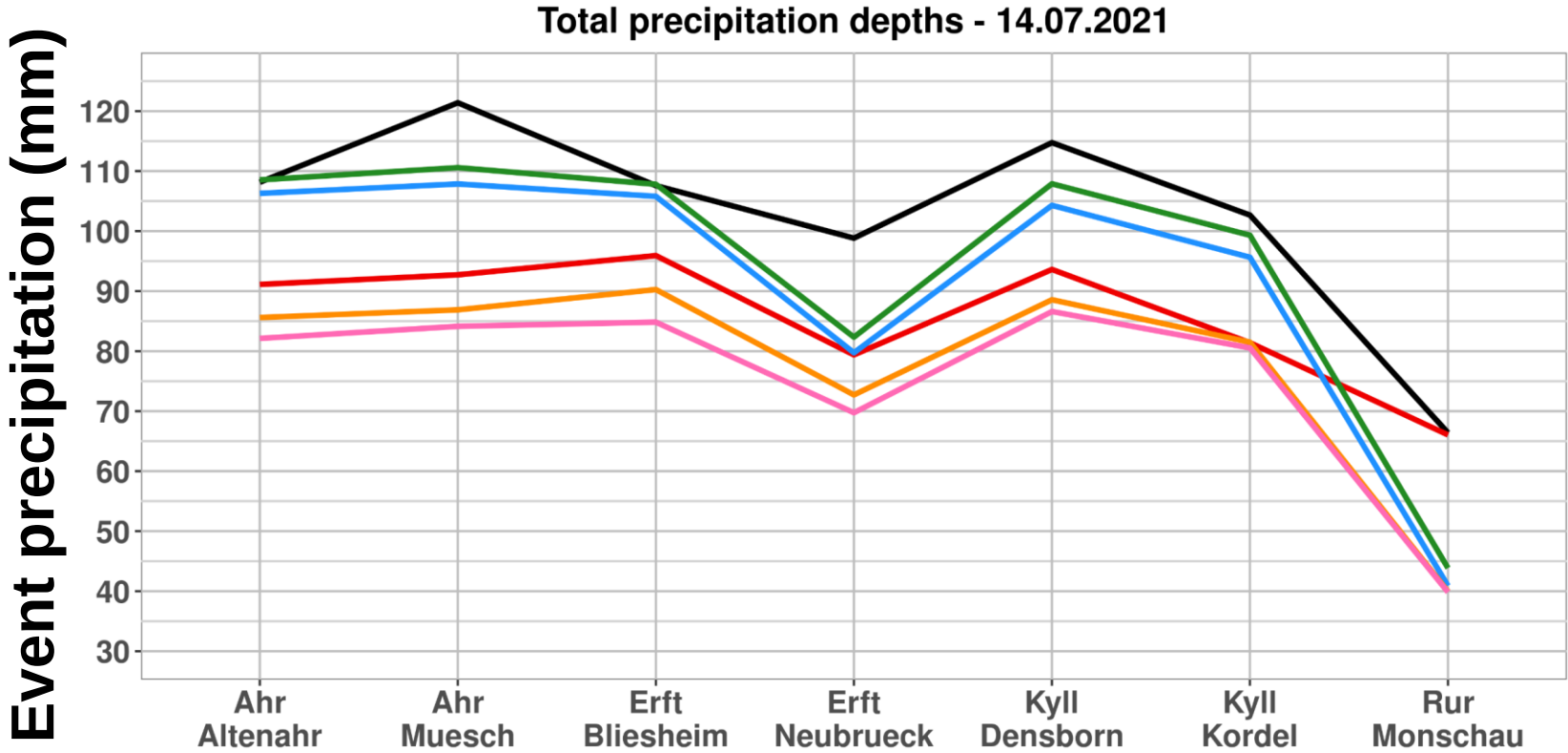


High uncertainty in estimated peakflows

A combination of uncertainties from model parameters and input QPE

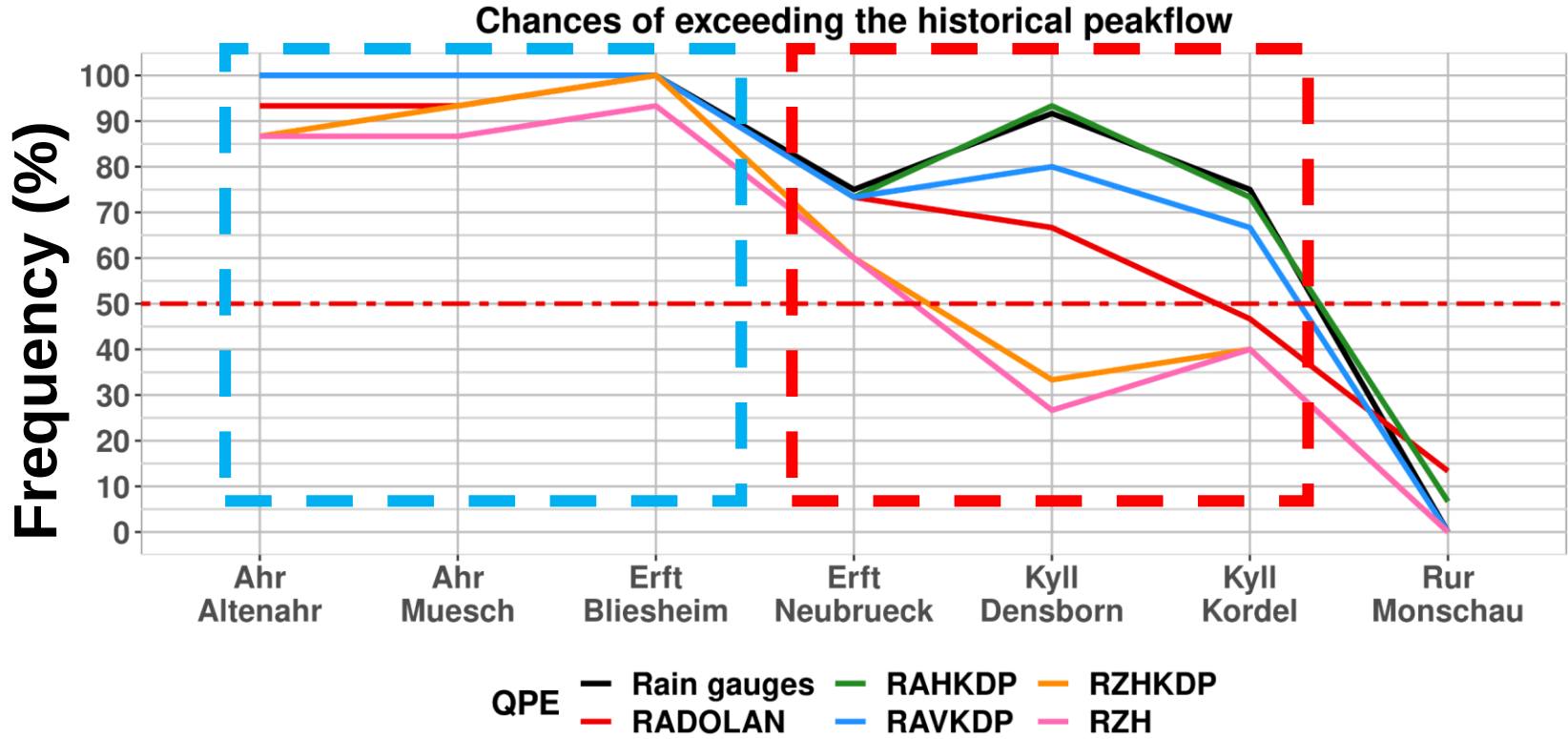
3| Results

3.2 | Result 2: Chances of having a record-breaking peakflow



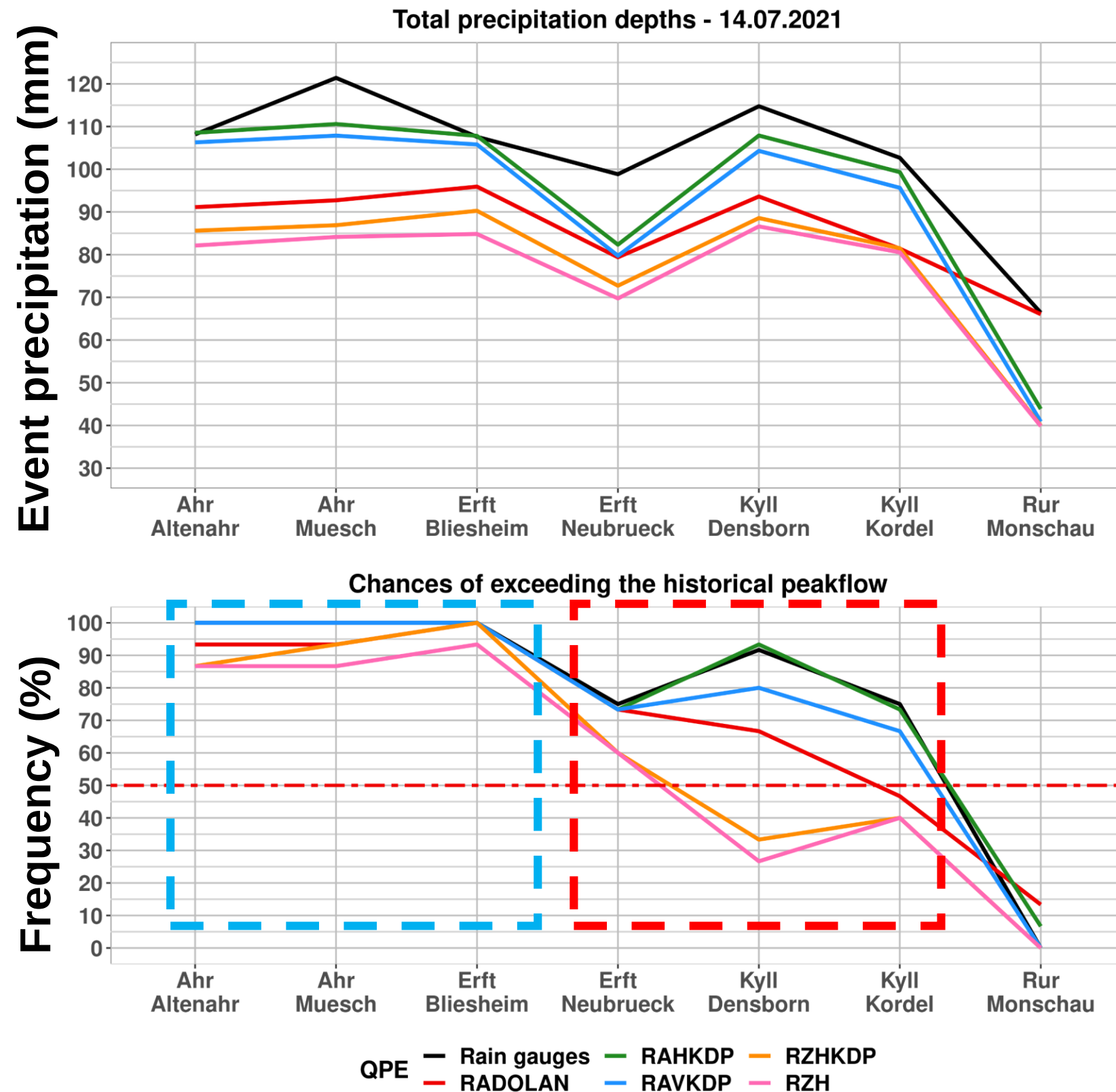
Radar-based products underestimated the total event precipitation

This was limited thanks to using polarimetric variables



3| Results

3.2 | Result 2: Chances of having a record-breaking peakflow



Radar-based products underestimated the total event precipitation

This was limited thanks to using polarimetric variables

The effect of the QPE on the (simulated) severity of the event varied among catchments

1. Very high chances no matter what QPE product is used
2. Very low chances for the Rur @ Monschau
3. High dependency on QPE for the remaining catchments

Thank you for your attention!

Questions?

Contact: m.saadi@fz-juelich.de



References

Chen, J.-Y., Trömel, S., Ryzhkov, A., Simmer, C., 2021. Journal of Hydrometeorology 22, 2617–2631. <https://doi.org/10.1175/JHM-D-20-0299.1>

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Status: this preprint is currently under review for the journal NHESS.

How uncertain are precipitation and peakflow estimates for the July 2021 flooding event?

Mohamed Saadi^{1,2}, Carina Furusho-Percot^{1,2}, Alexandre Belleflamme^{1,2}, Ju-Yu Chen³, Silke Trömel^{3,4}, and Stefan Kollet^{1,2}

¹Institute for Bio- and Geosciences (IBG-3, Agrosphere), Forschungszentrum Jülich, Jülich 52425, Germany

²Centre for High-Performance Scientific Computing in Terrestrial Systems, Geoverbund ABC/J, Jülich 52428, Germany

³Institute for Geosciences, Department of Meteorology, Universität Bonn, Bonn 53121, Germany

⁴Laboratory for Clouds and Precipitation Exploration, Geoverbund ABC/J, Bonn 53121, Germany