

# Pattern-aware flash flood modeling with a 2D hydraulic-hydrological model and multi-source variational data assimilation

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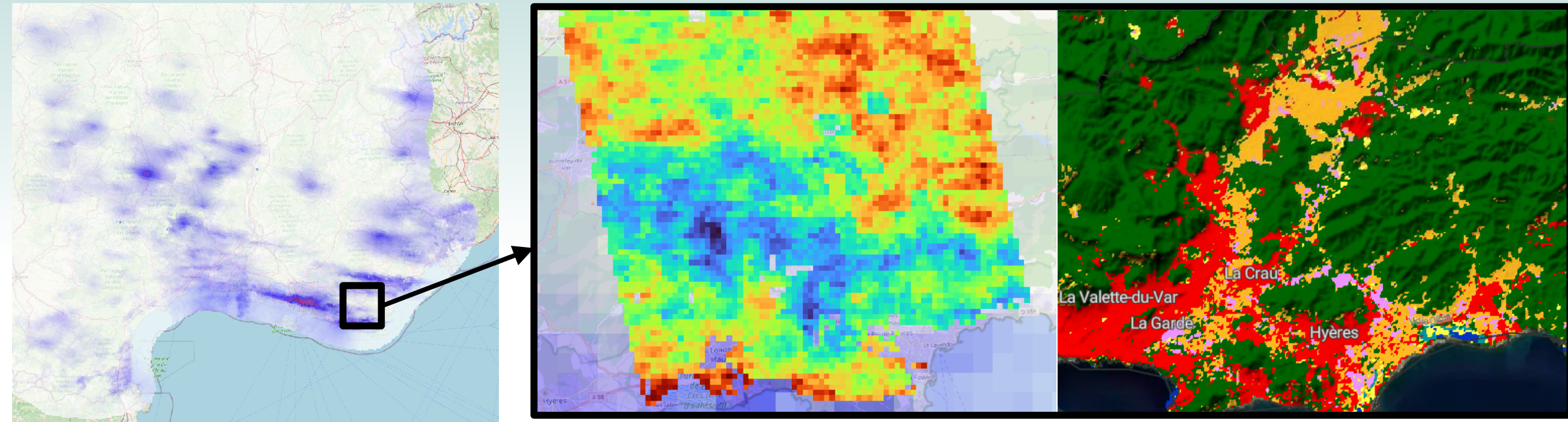
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## Context: accrued observability of fine-scale data informing on flash flood genesis

### High resolution data products from global databases



ANTILOPE J+1  
reanalysis rain product  
Spatial resolution : 1km<sup>2</sup> tiles  
Time step : 15min

Excerpt from ISRIC database  
soil composition data :  
clay content over 250x250m tiles

Land cover from Copernicus  
Global Land Cover (2015)

### Wealth of information from global databases

- High resolution rain data products (in space and time)
- Global estimates of soil occupation, soil composition, soil moisture
- Global DEM and local high resolution LiDAR data

=> an opportunity to model fine-scale, accurate flood genesis phenomenons with distributed parameters and states over large scale catchments at high-resolution

### Scientific challenges

- Accurately modeling multi-scale non-linear phenomenons in hydraulic-hydrological rainfall-runoff models
- Inferring/Calibrating distributed hydraulic parameters from the partial observation of their mixed signatures

### Goal of this work

- Upgrading the DassHydro modeling platform with infiltration terms needed to model fine-scale rainfall-runoff
- Integrating relevant observations of model states and parameters in modeling approach
- Implementing a regionalization scheme for distributed parameters, integrated to the variational data assimilation methods

## Hydraulic-hydrological model : DassFlow2D

2D shallow-water equations with infiltration source terms

$$\frac{\partial}{\partial t} \mathbf{U} + \frac{\partial}{\partial x} \mathbf{F}(\mathbf{U}) + \frac{\partial}{\partial y} \mathbf{G}(\mathbf{U}) = \mathbf{S}_g(\mathbf{U}) + \mathbf{S}_f(\mathbf{U}) + \mathbf{S}_{hy}(\mathbf{U})$$

$$\mathbf{U} = \begin{bmatrix} h \\ hu \\ hv \end{bmatrix}, \mathbf{F}(\mathbf{U}) = \begin{bmatrix} hu \\ hu^2 + \frac{gh^2}{2} \\ huv \end{bmatrix}, \mathbf{G}(\mathbf{U}) = \begin{bmatrix} hv \\ huv \\ hv^2 + \frac{gh^2}{2} \end{bmatrix},$$

$$\mathbf{S}_g(\mathbf{U}) = \begin{bmatrix} 0 \\ -gh\nabla b \end{bmatrix}, \mathbf{S}_f(\mathbf{U}) = \begin{bmatrix} 0 \\ -g \frac{n^2 \|\mathbf{u}\|}{h^{1/3}} \mathbf{u} \end{bmatrix}, \mathbf{S}_{rr}(\mathbf{U}) = \begin{bmatrix} r - i \\ 0 \end{bmatrix}$$

SCS-CN approach : empirical net rainfall

Green-Ampt approach : conceptual soil column

$$r - i = r_n = \frac{(R - \lambda S(CN))^2}{(R + (1 - \lambda)S(CN))}, \quad i = K_s \left[ 1 + \frac{(\theta_s - \theta_i) \Psi_f + h}{I} \right]$$

## Variational Data Assimilation method

- Multi-objective cost function :  $J(\theta) = J_{obs}(\theta) + \alpha J_{reg}(\theta)$

$$J_{obs}(\theta) = \|h_{obs} - h\|_{\mathcal{O}_h}^2 + \|Q_{obs} - Q\|_{\mathcal{O}_Q}^2$$

- Control vector (differentiable regionalization operator) :

$$\theta = (K_s(x, y), \Delta\theta(x, y)) \xrightarrow{\text{MRPF}} \theta = (\kappa_{1..6}(R))$$

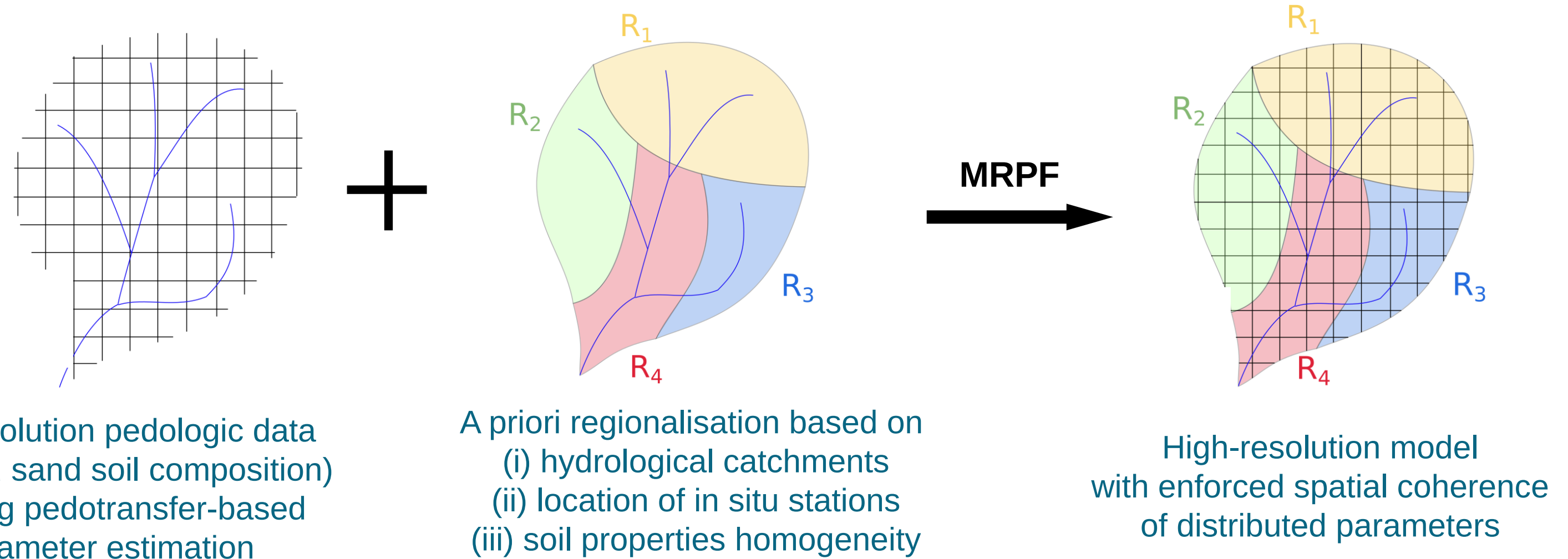
- Adjoint model obtained through automatic differentiation of the forward model

- Optimization problem :  $\theta^* = \underset{\theta}{\operatorname{argmin}} J(\theta)$

## Solving the ill-posed high-dimensional inverse problem from spatially sparse observations : Multi-scale Regional Mapping Function (MRPF)

MRPF : implicit infiltration parameter regionalization based on a priori knowledge of inverse problem solutions :

- Green-Ampt infiltration parameters are calculated from pedologic data using pedotransfer functions
- Pedotransfer coefficients are regionalized, hence reducing the number of sought parameters and enforcing spatial coherence of infiltration parameters within a region, integrating a priori knowledge of spatial variability provided by pedologic data



## Integrating data from global databases : Fortran-Python wrapping upgrade of DassFlow2D

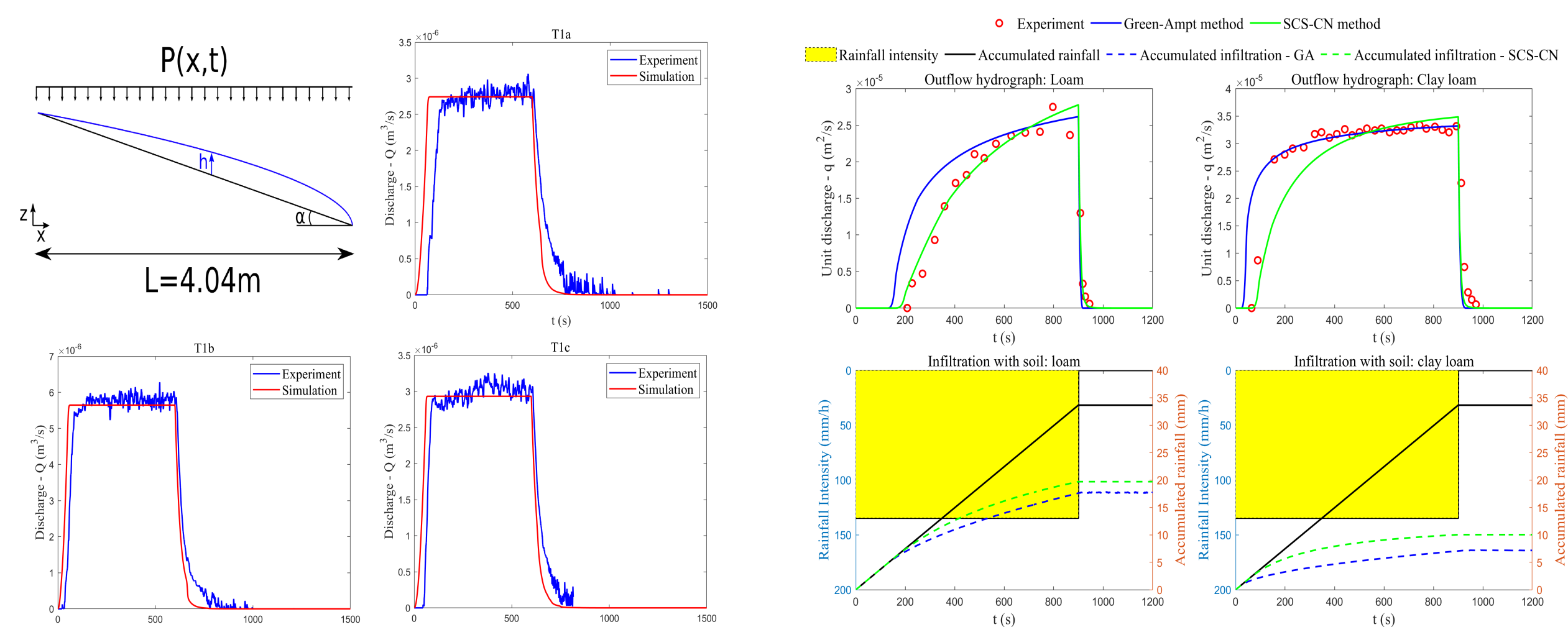
- Python module upgraded to import and preprocess raster data to DassFlow mesh (e.g. ANTILOPE J+1)
- Exports in HDF5 format for easy sharing
- MPI reading and interfacing for large data sets

## Numerical results

### Validation of the source terms

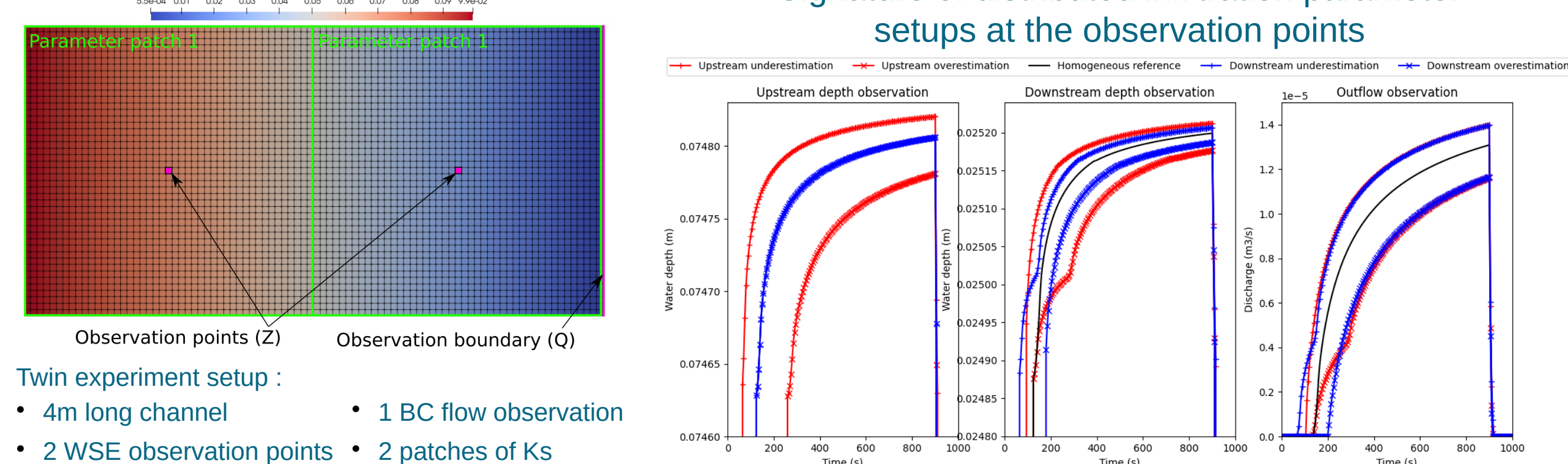
Validation of rainfall source term against experimental bench data from [Kirstetter et al., 2016]

Validation of infiltration source terms based on experimental bench data on homogeneous shallow soils from [DeLima, 1992]

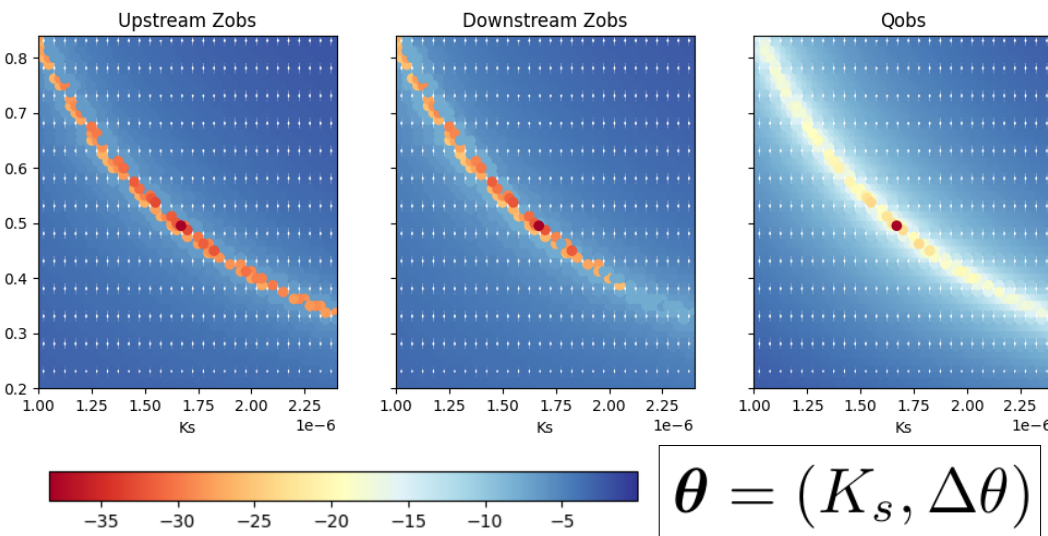


## Twin experiments and hydraulic signature analysis

Signature of distributed infiltration parameter setups at the observation points

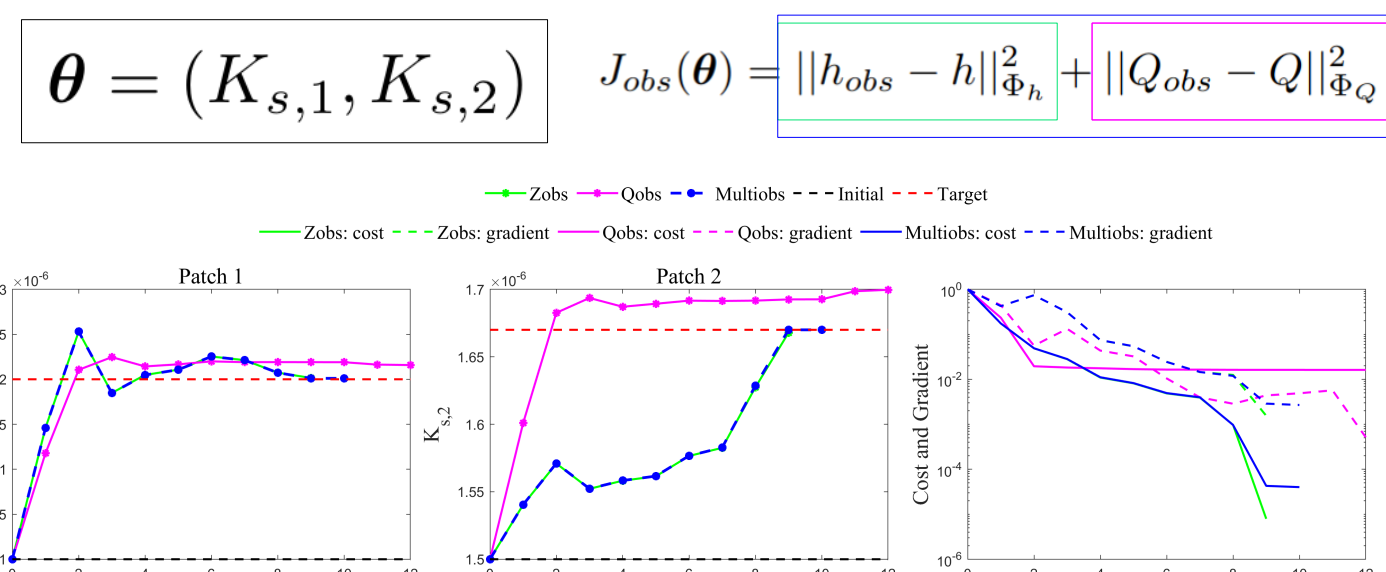


- Twin experiment setup :
  - 4m long channel
  - 1 BC flow observation
  - 2 WSE observation points
  - 2 patches of Ks



Sampling of the cost function parts in the parameter space : equifinality between initial water content and infiltrability

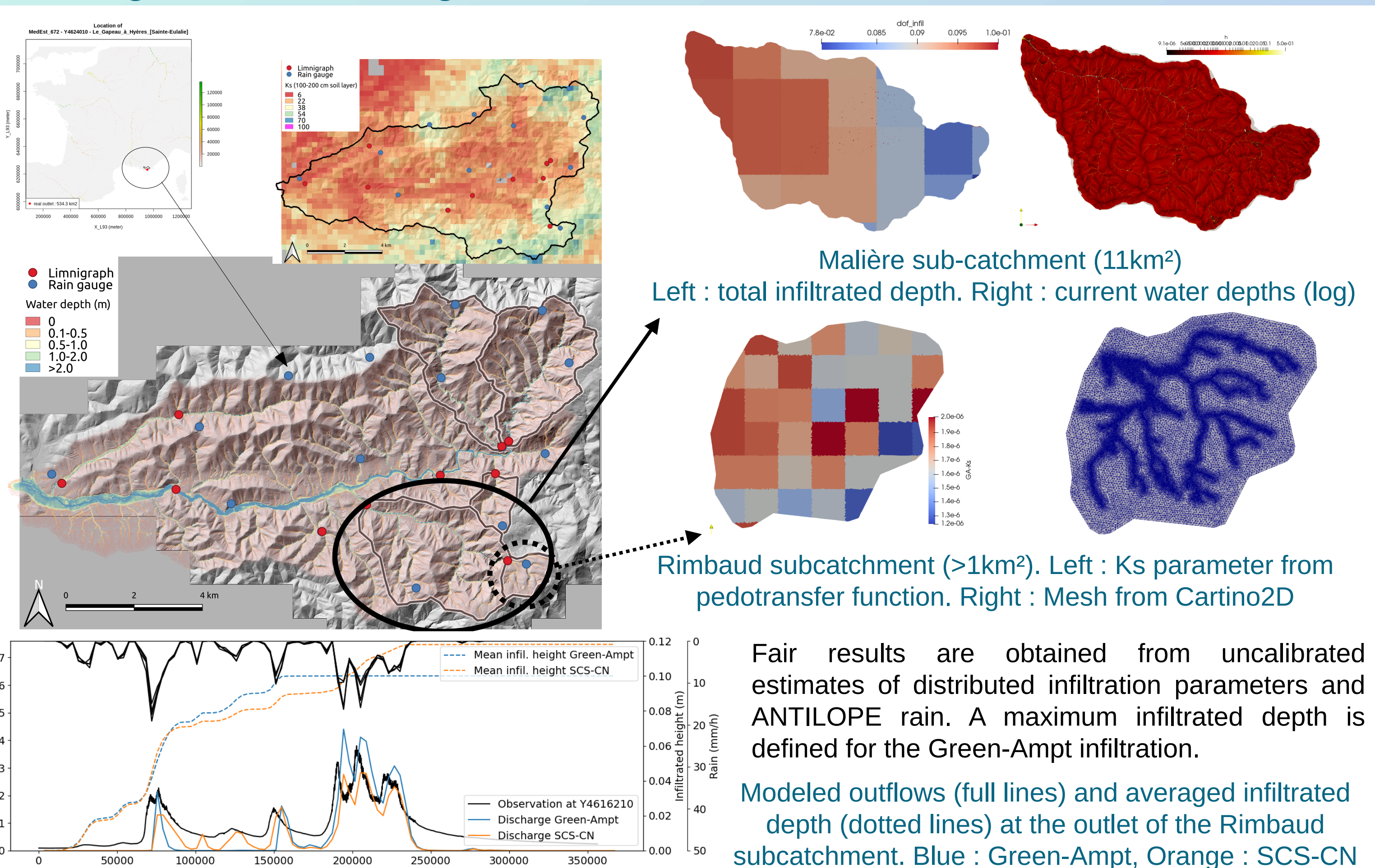
Different parameters have similar effects on observed hydraulic signatures at certain space-time scales.



Inference of distributed parameters using water depth and downstream discharge observations

Signatures of distributed parameters may not reflect their distributed character, leading to potential issues of signature attribution and spatial equifinality.

### Large scale modeling of flash floods : the Réal-Collobrier catchment



Fair results are obtained from uncalibrated estimates of distributed infiltration parameters and ANTILOPE rain. A maximum infiltrated depth is defined for the Green-Ampt infiltration.

Modeled outflows (full lines) and averaged infiltrated depth (dotted lines) at the outlet of the Rimbaud subcatchment. Blue : Green-Ampt, Orange : SCS-CN

## Pattern-aware inferences

- MPR is fully implemented and differentiable, adjoint model is obtained.
- Next steps include calibration of distributed parameters and comparison to "simpler" approaches through performance at gauges, initialisation of hydrological states in the hydraulic model through remote-sensed data (e.g. from SMAP) or from hydrological models (e.g. SMASH from the DassHydro toolchain).

### Other current studies with DassFlow :

- Application on urban flash floods on the Riviera-Palmerie catchment in Abidjan : challenging assimilation problem given high spatial and temporal variability of hydrodynamic phenomenons and less observations sources (no satellite since events are under 5h, less developed in situ station network). Planned leveraging of observed surface velocity fields from optical analysis of multi-source videos.
- Multi-scale modeling of large scale river networks and selected high-resolution flooding areas (based on [Pujol, 2022]). Planned leveraging of both satellite data, in situ stations and surface velocities in multi-source assimilation approach.