

# ASSIMILATING SATELLITE SOIL MOISTURE AND FLOOD EXTENT MAPS INTO A FLOOD PREDICTION MODEL.

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# CONTEXT

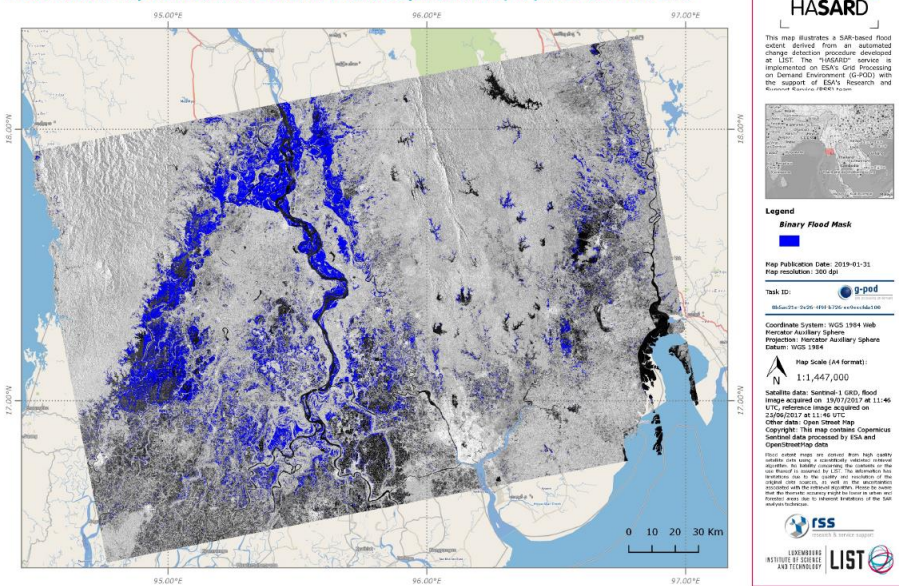


A powerful tool for flood management and prediction: hydrological modelling

Need for observations to set up, calibrate and evaluate these models. Issues:

- Traditional observations are punctual (pb of representativeness).
- Observations are scarcely distributed and observation networks tend to be further reduced (e.g. stream gauges)
- Ground observations not always reliable during flood events.

Flood extent map derived from Sentinel-1 data acquired on 19/07/2017 at 11:46 UTC



=> Need for new observation techniques : good candidates : satellite SAR flood images, and satellite derived soil moisture products

# MORE AND MORE READILY AVAILABLE (RADAR) OBSERVATIONS

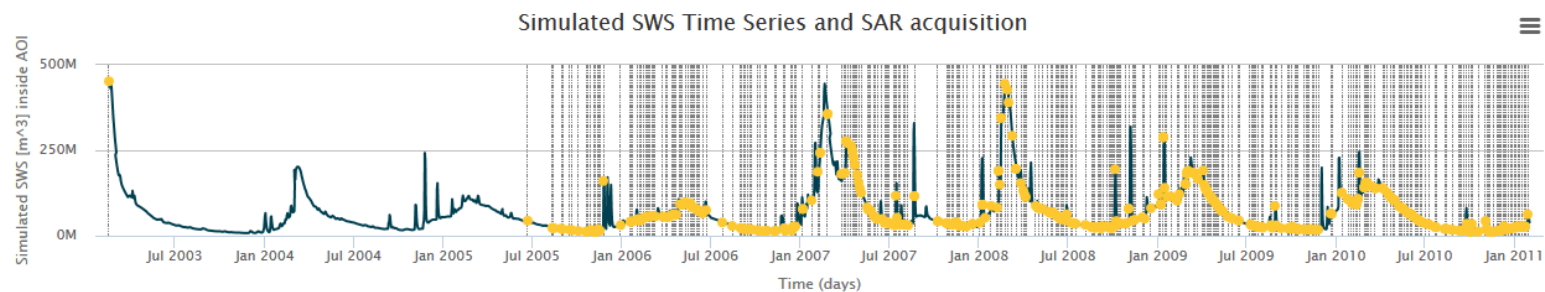
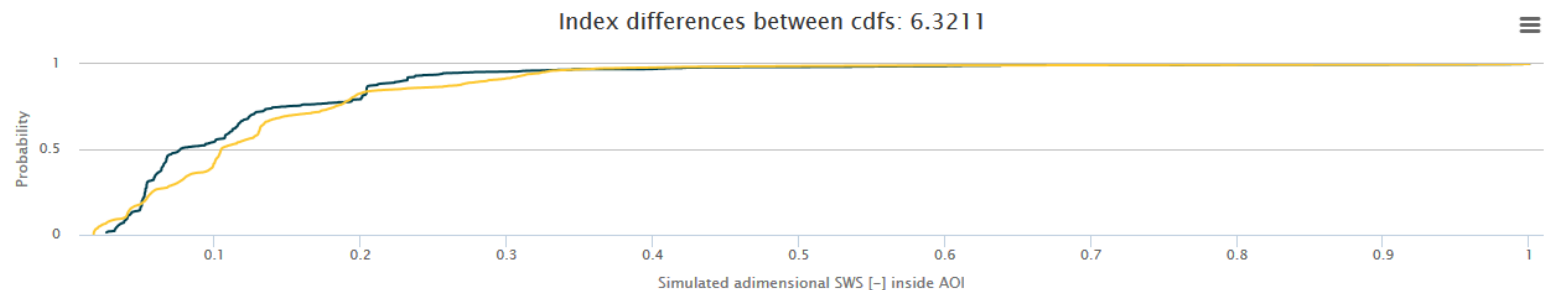


grid processing on demand

European Space Agency

Day–Night–All weather acquisitions

Synoptic view of large areas



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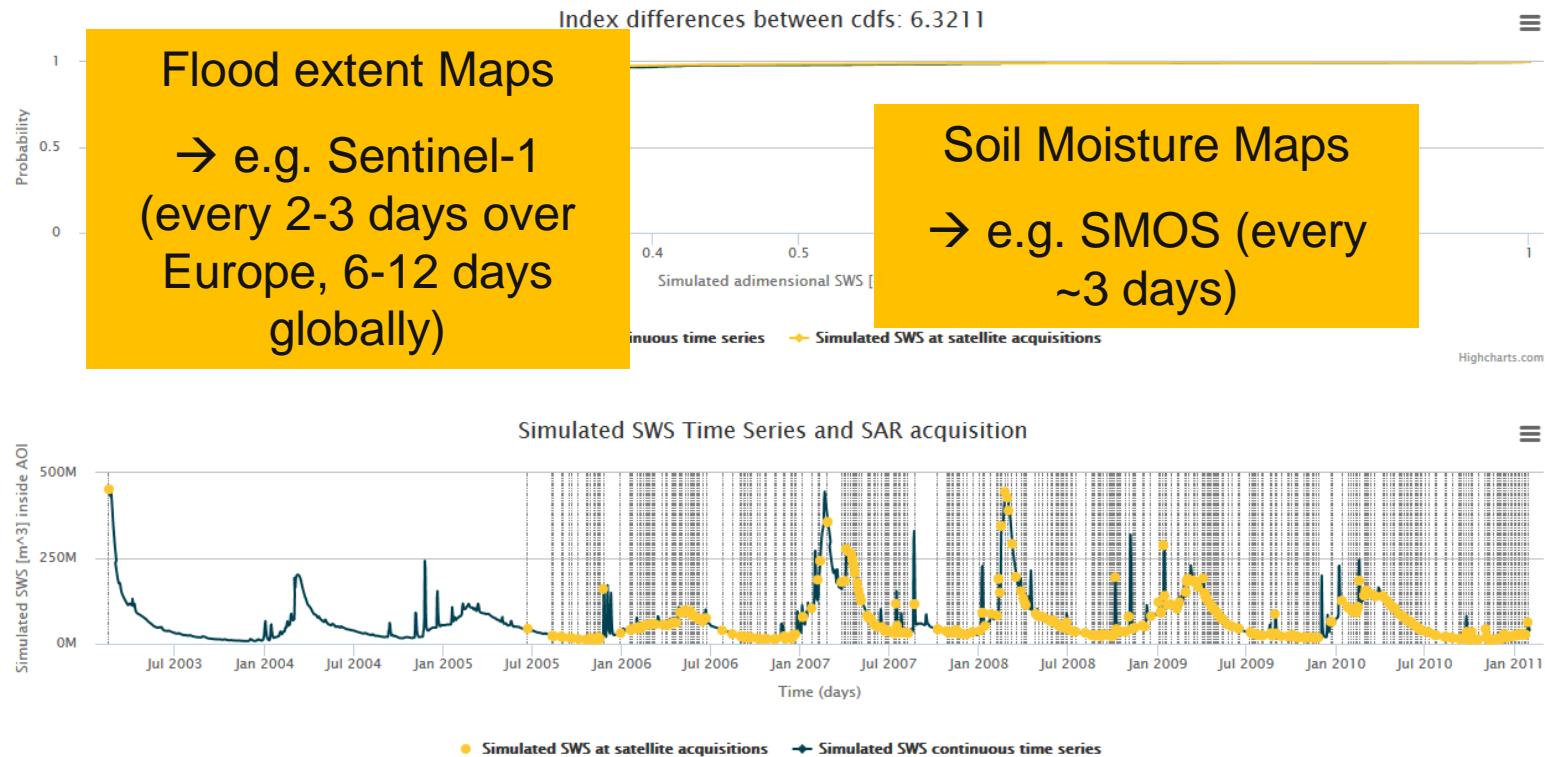


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**Research question: Are these EO datasets sufficient for calibrating a distributed conceptual hydrological model ?**



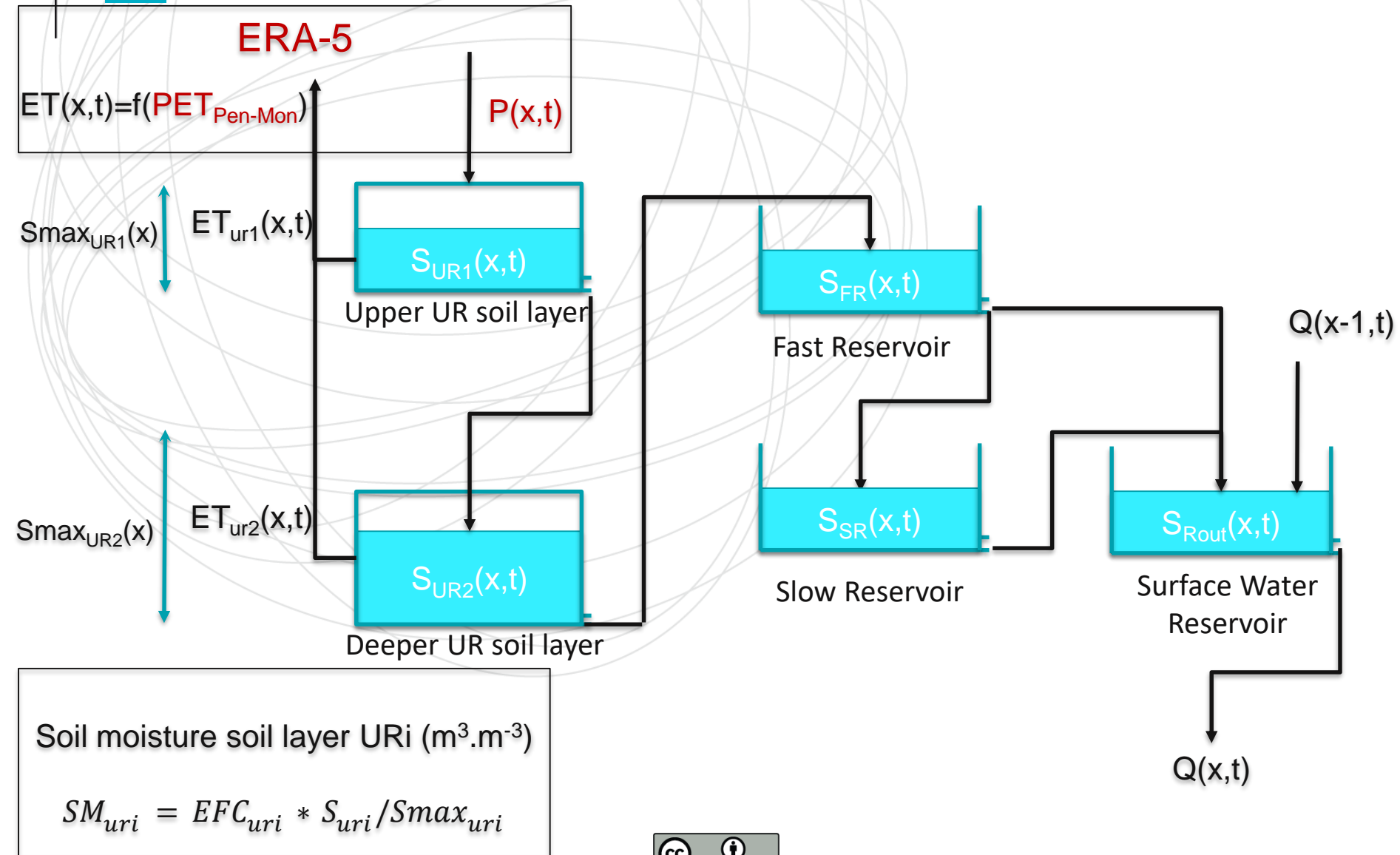
# THE CONCEPTUAL HYDROLOGICAL MODEL



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# THE MODEL STRUCTURE



# THE MODEL STRUCTURE

The diagram illustrates the model structure for simulating flood extent. It shows a sequence of model components and their interactions, leading to the final simulated flood extent map.

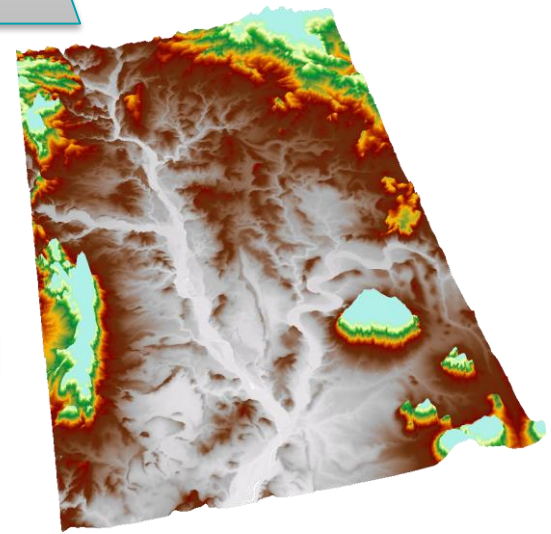
The model structure is organized into three rows of diagrams, each representing a different stage or component of the simulation:

- Top Row:** Three identical diagrams showing the initial model structure. Each diagram includes a box labeled  $ET_{(1-HPET_{max})}$  and  $PMAS$ , a box labeled  $PMAS$ , and a box labeled  $Q_{(1-HPET_{max})}$ . Arrows indicate the flow of data between these components.
- Middle Row:** Three identical diagrams showing the model structure after the first stage of simulation. Each diagram includes a box labeled  $ET_{(1-HPET_{max})}$  and  $PMAS$ , a box labeled  $PMAS$ , and a box labeled  $Q_{(1-HPET_{max})}$ . Arrows indicate the flow of data between these components.
- Bottom Row:** Two identical diagrams showing the model structure after the second stage of simulation. Each diagram includes a box labeled  $ET_{(1-HPET_{max})}$  and  $PMAS$ , a box labeled  $PMAS$ , and a box labeled  $Q_{(1-HPET_{max})}$ . Arrows indicate the flow of data between these components.

Arrows indicate the flow of data between the diagrams, showing the progression of the simulation. A large blue arrow points from the bottom row of diagrams to the final simulated flood extent map.

**Simulated flood extent map**

## Topography information



## Simulated flood extent map

# ASSIMILATION DESIGN

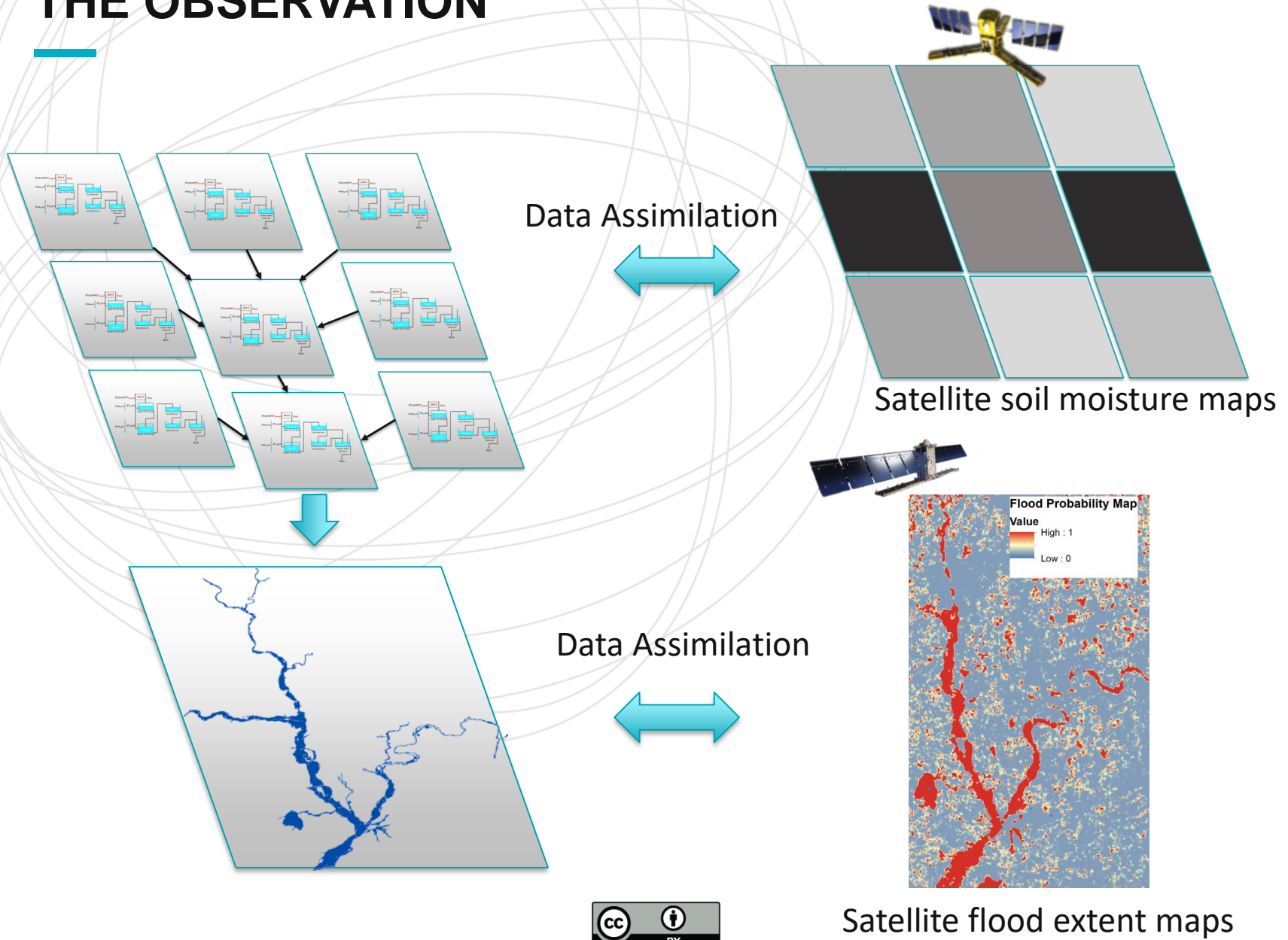


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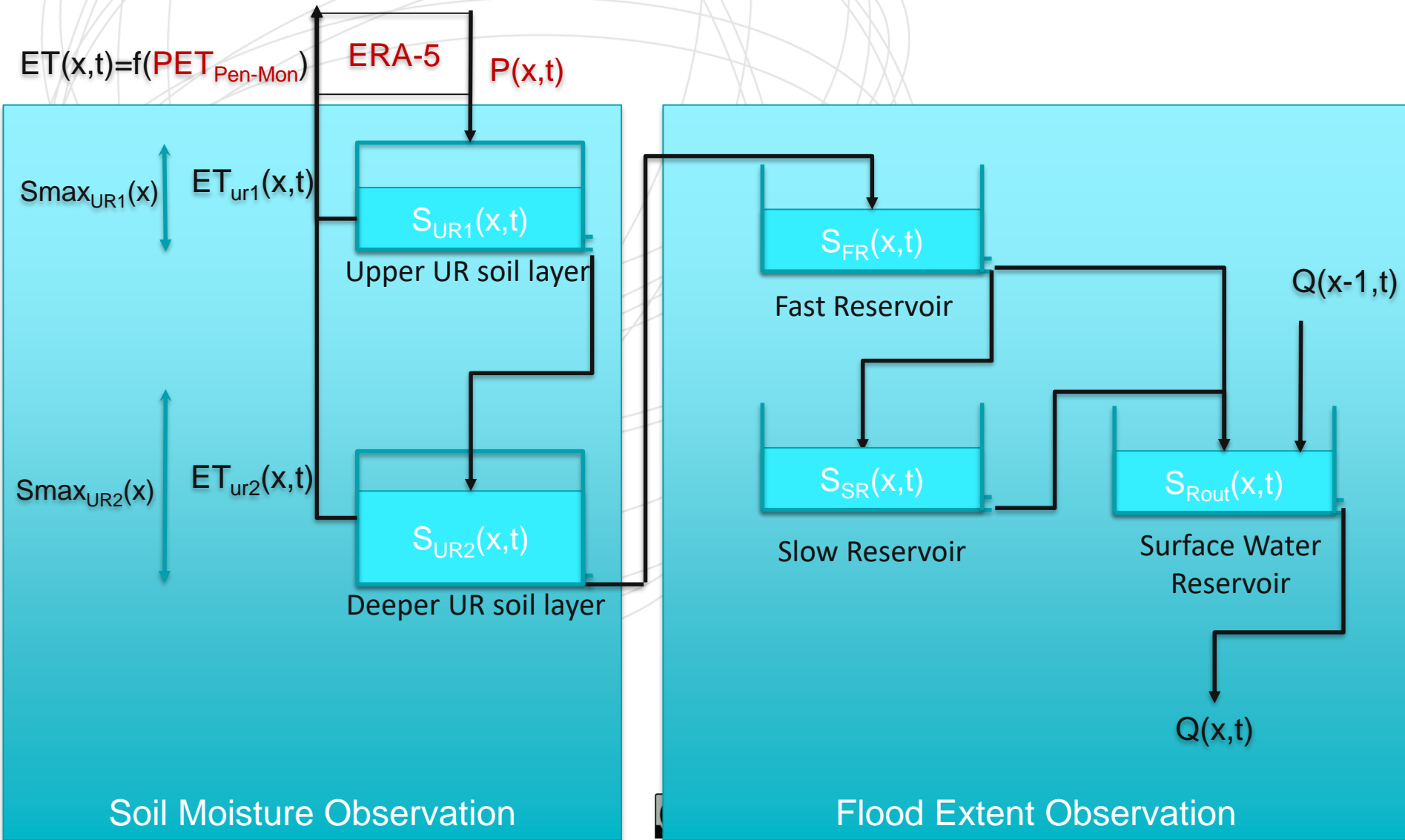




# ASSIMILATION DESIGN: THE OBSERVATION



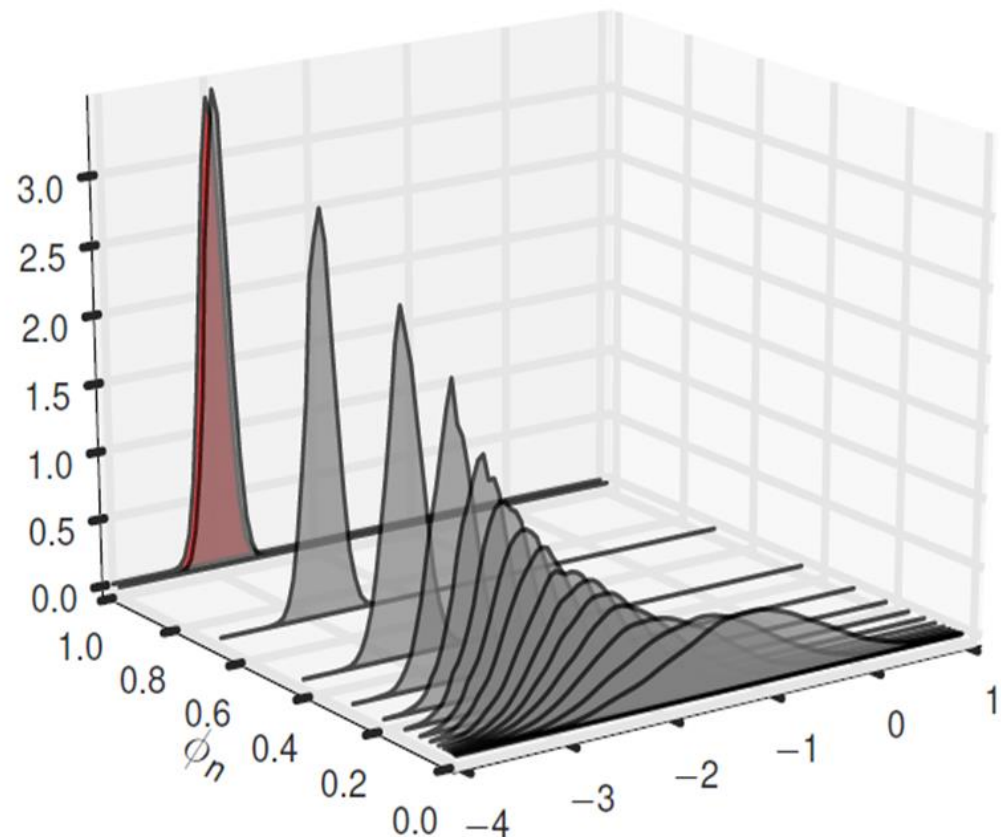
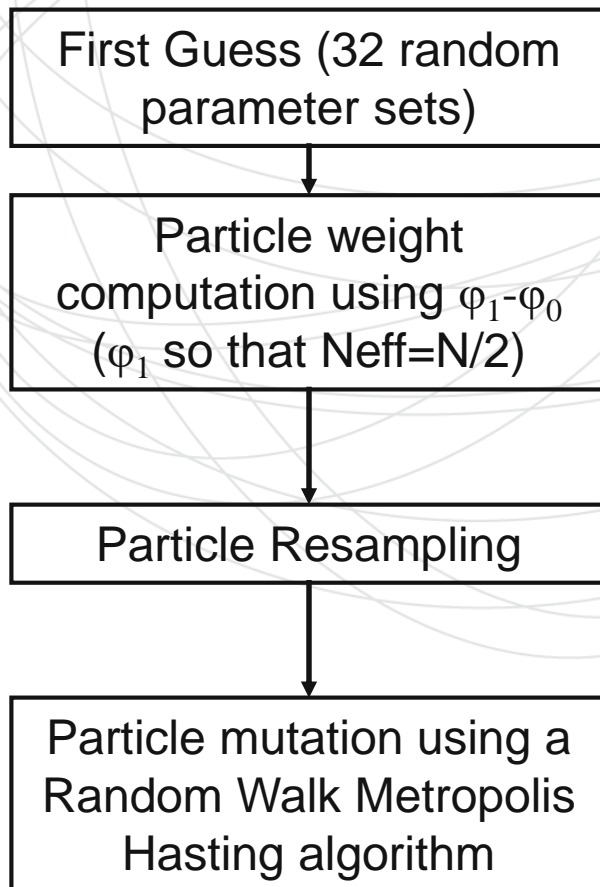
# THE ASSIMILATION DESIGN PARAMETER UPDATING STRATEGY



# THE ASSIMILATION DESIGN: A TEMPERED PARTICLE FILTER

Bayes Theorem: 
$$p(\theta|o) = \frac{p(o|\theta)}{p(o)} p(\theta) = \prod_{n=1}^K \frac{p(o|\theta)^{\varphi_n - \varphi_{n-1}}}{p(o)} p(\theta)$$

$$0 = \varphi_0 < \varphi_1 < \varphi_2 < \dots < \varphi_K = 1$$

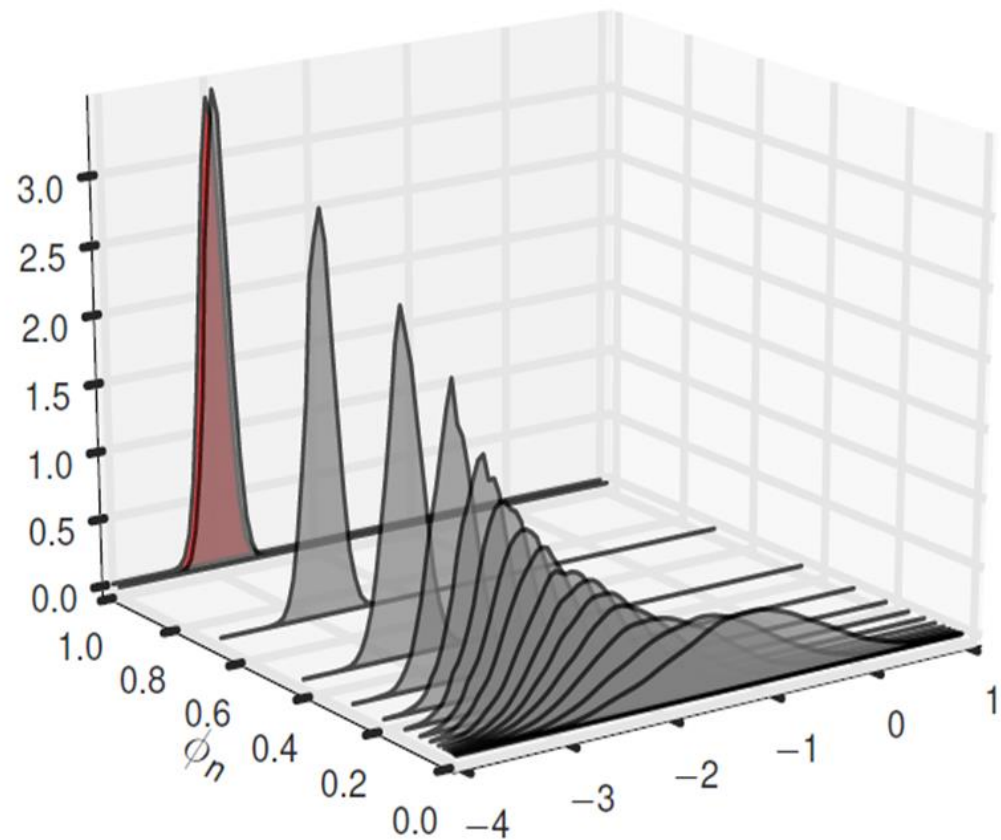
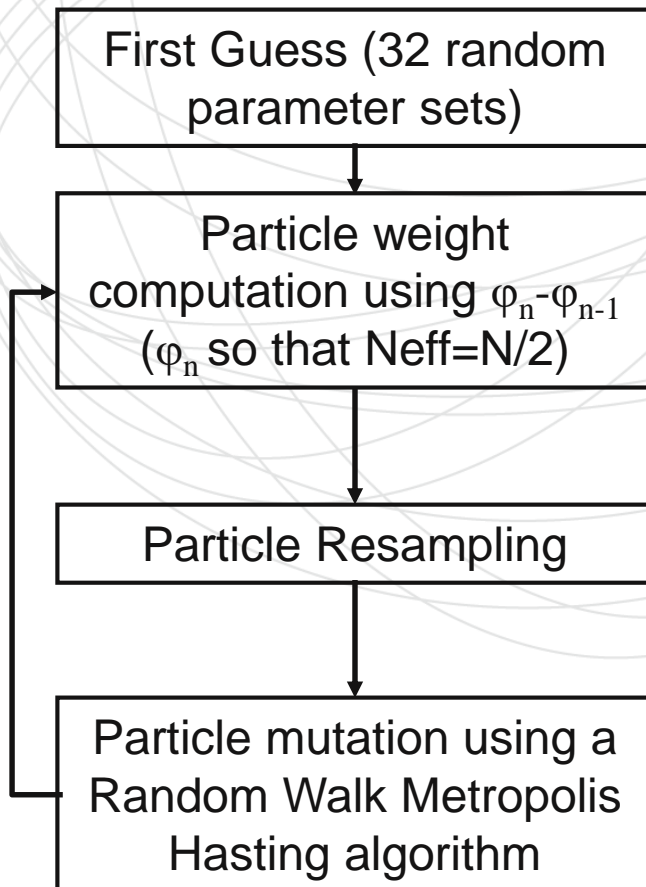


After Herbst et al., 2019

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# SYNTHETIC TWIN EXPERIMENTS



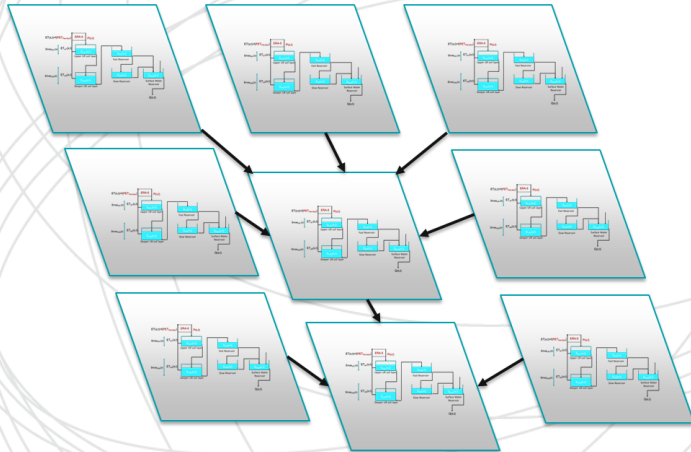
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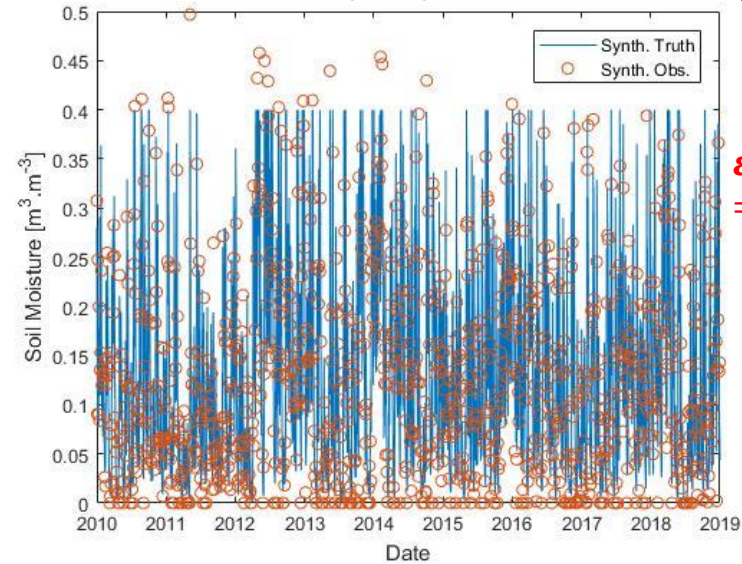


# SYNTHETIC TWIN EXPERIMENTS: SYNTHETIC TRUTH AND OBSERVATION

Model forward run (9 years, hourly)

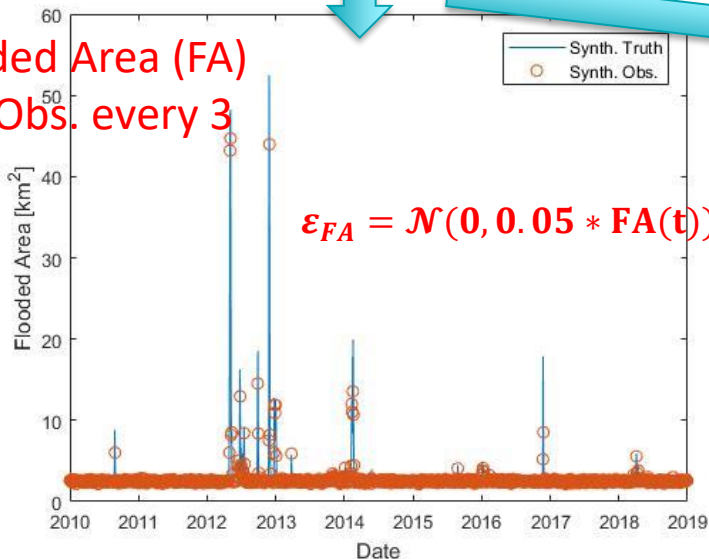


Soil moisture (SM): One Obs. every 3 days

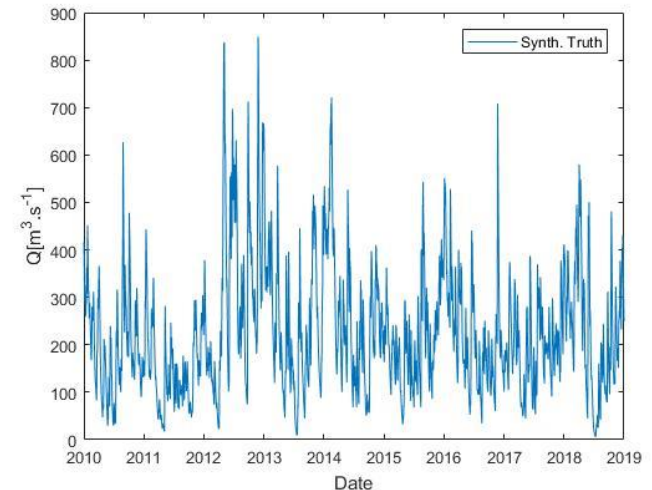


$$\varepsilon_{SM} = \mathcal{N}(0, 0.04)$$

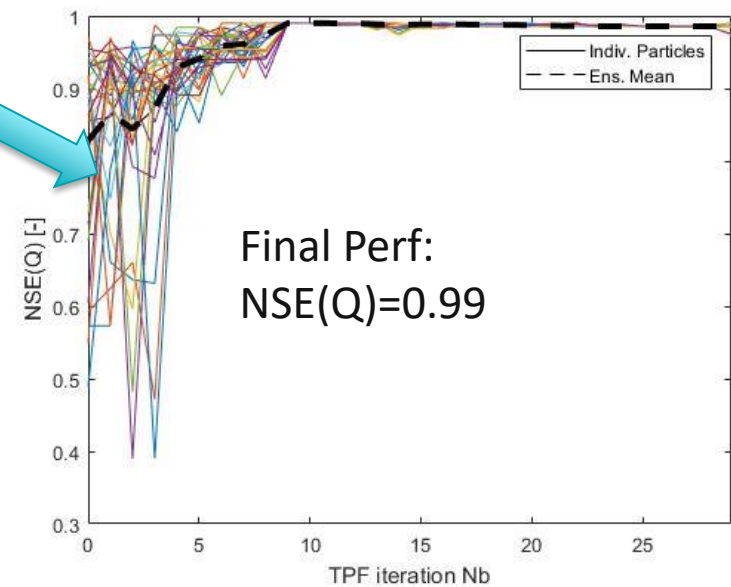
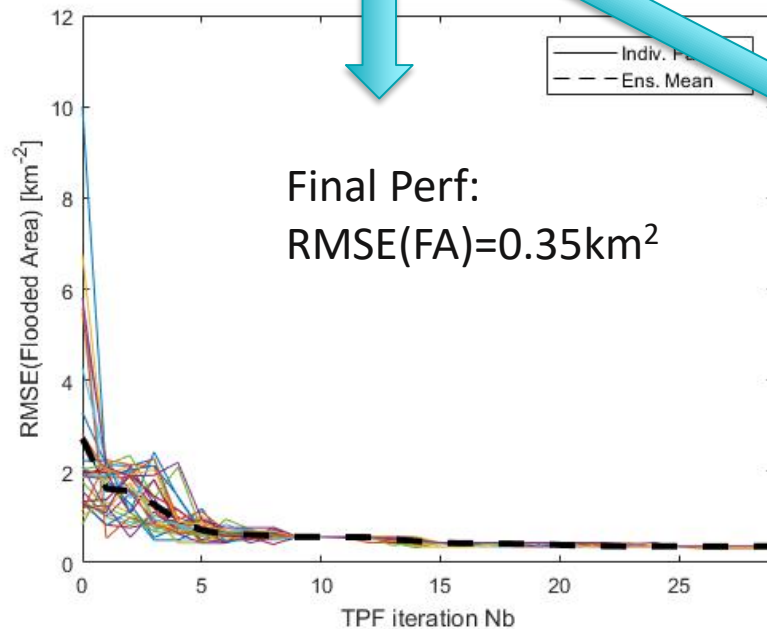
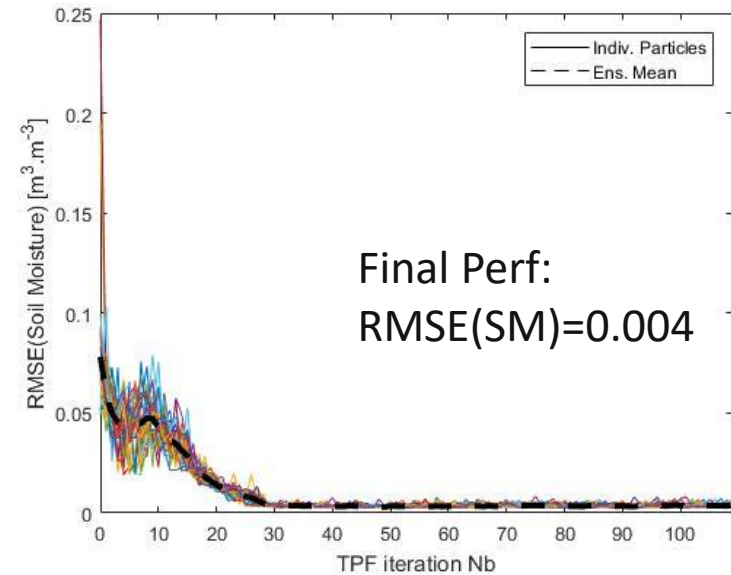
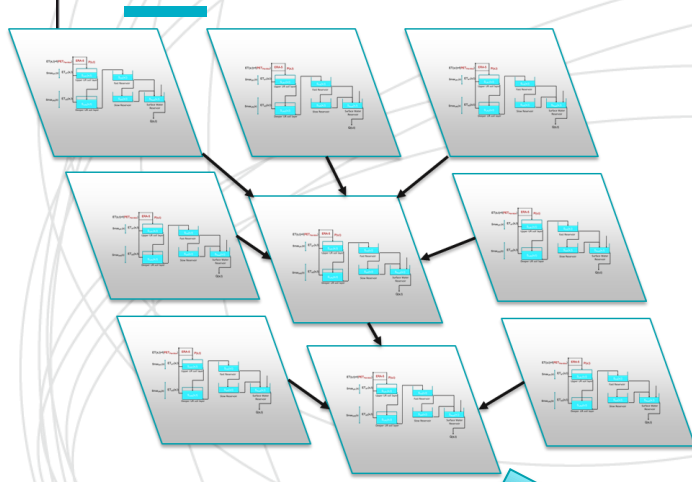
Flooded Area (FA)  
One Obs. every 3  
days



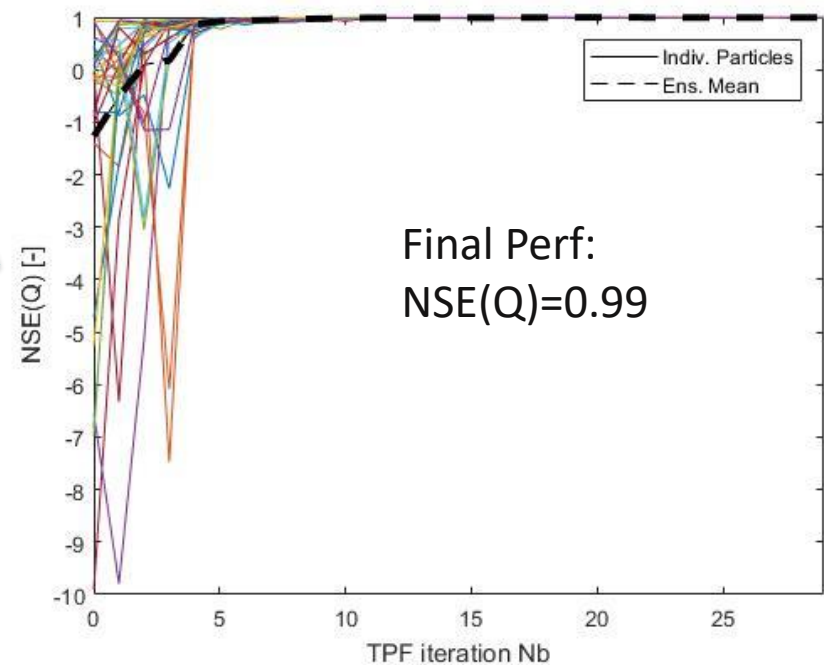
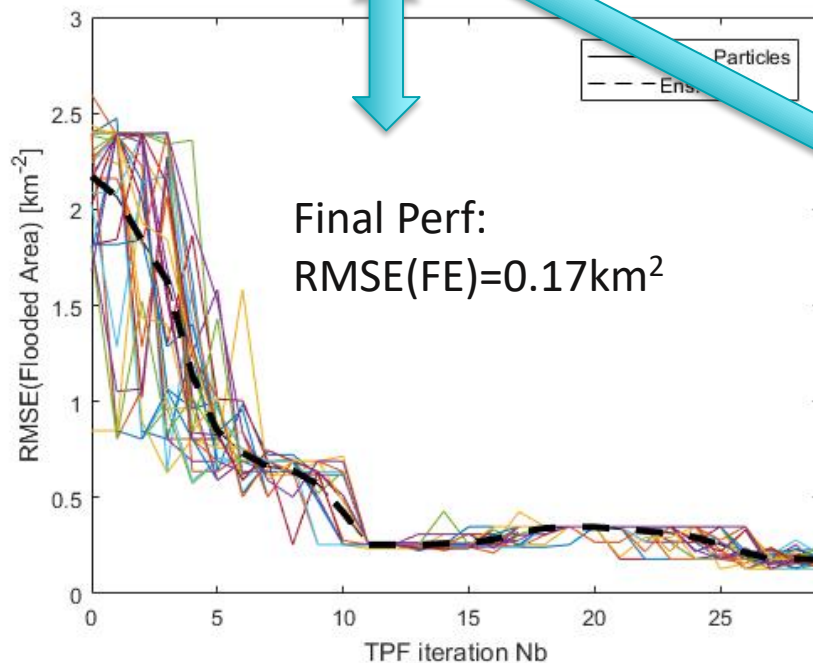
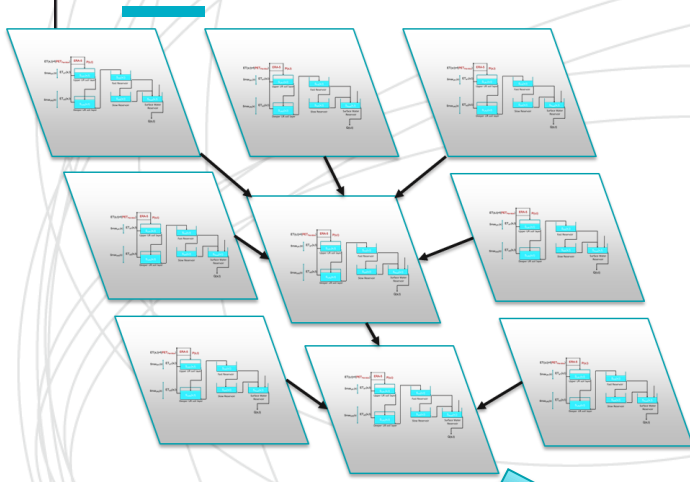
$$\varepsilon_{FA} = \mathcal{N}(0, 0.05 * FA(t))$$



# SYNTHETIC TWIN EXPERIMENTS: MODEL CALIBRATION USING SM+FA

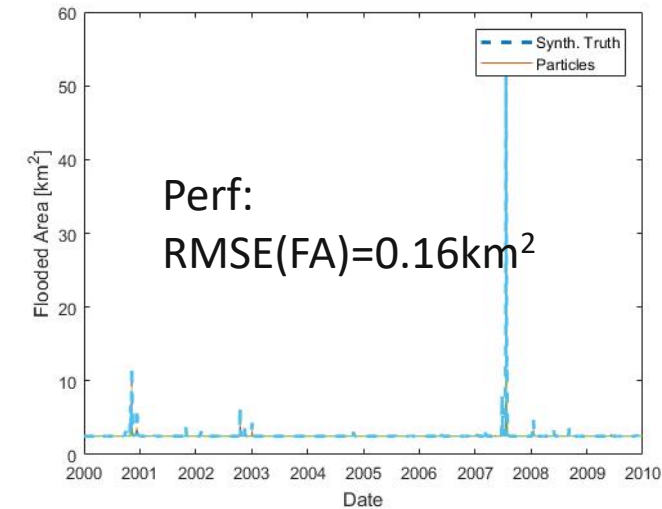


# SYNTHETIC TWIN EXPERIMENTS: MODEL CALIBRATION USING FA ONLY

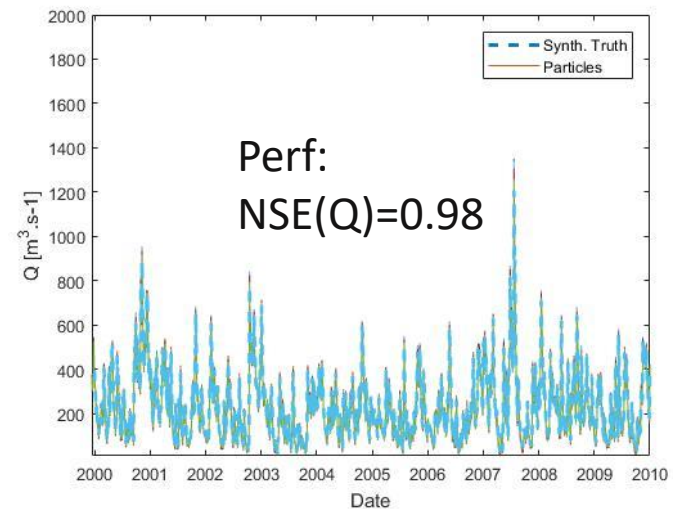
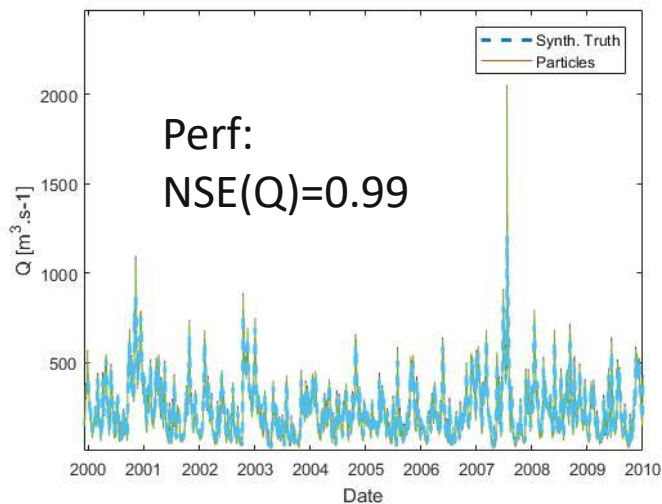
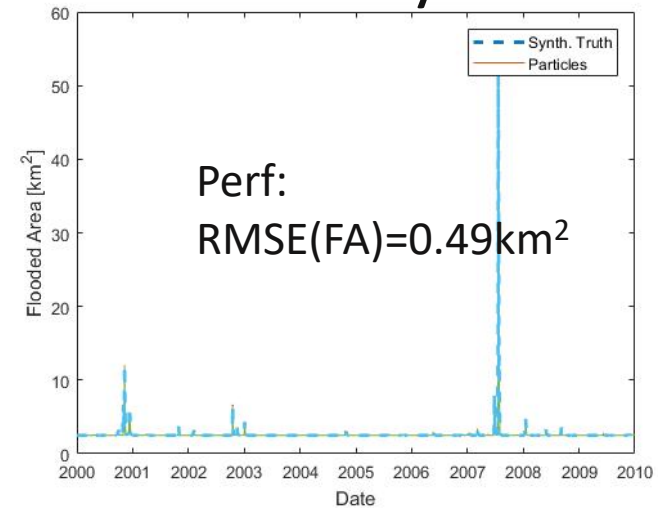


# SYNTHETIC TWIN EXPERIMENTS: CALIBRATED MODEL EVALUATION

## SM+FA



## FA only





# CONCLUSION & NEXT STEPS

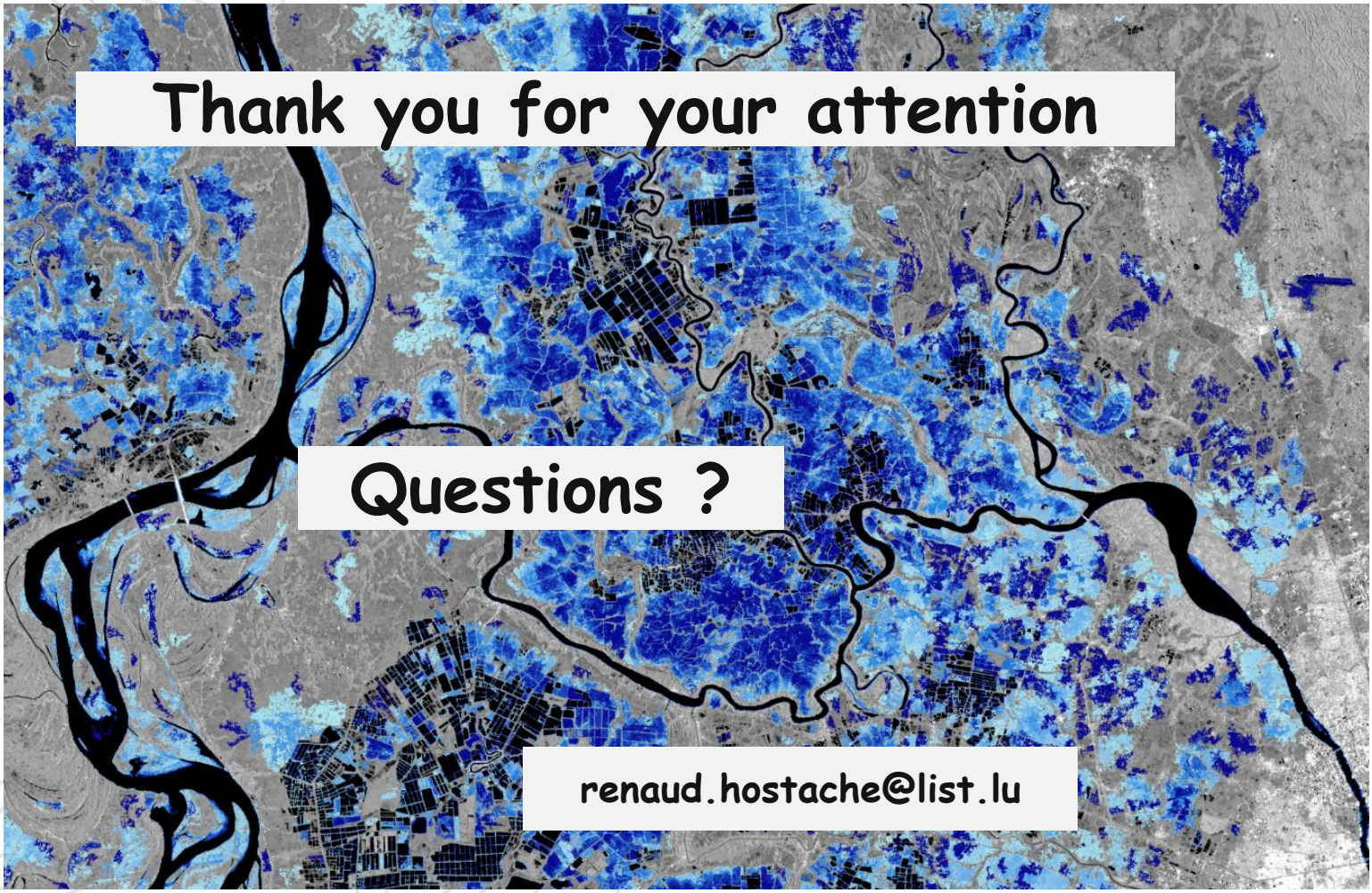
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- We carried out a synthetic experiment using a TPF of the joint assimilation of satellite flooded area and soil moisture observation
- The results are really promising as the calibrated model is predicting surface runoff accurately both during the calibration and the validation periods
- This opens the floor for applications at large scale over poorly gauged areas

## Next steps:

- To further investigate the added value of soil moisture data
- To carry out real test case experiments





Thank you for your attention

Questions ?

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