



# A generic hydrological forecasting system using existing and future altimetry missions : an OSSE study over the Niger basin

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(2)



**Hydromatters**  
Prendre la mesure de l'eau From space to society

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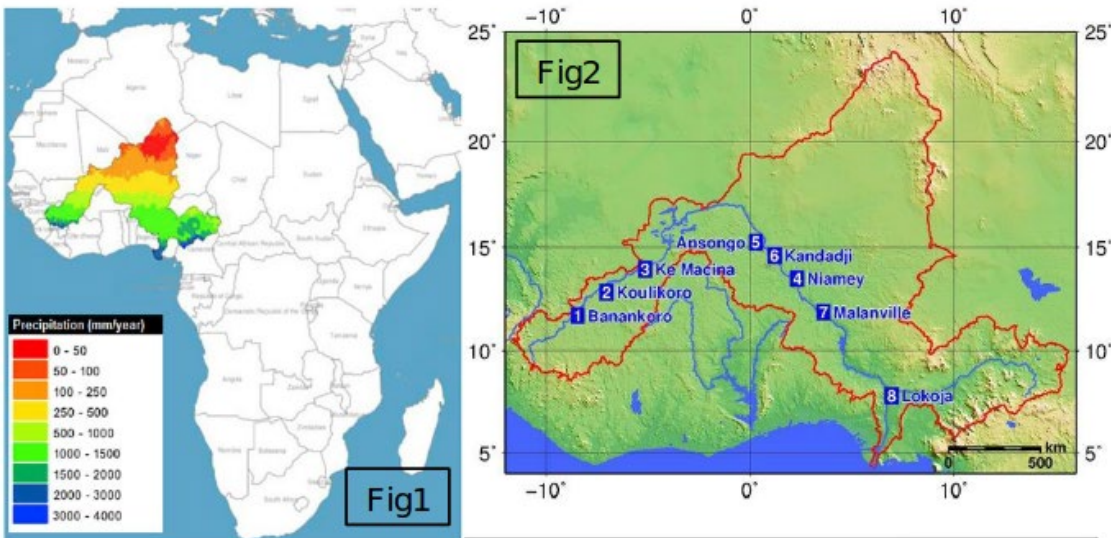


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## Flow forecasting in the Niger river basin : a key issue



- Length : 4200km
- Origin: Guinean Fouta-Djalou Plateau
- flows into Atlantic Ocean

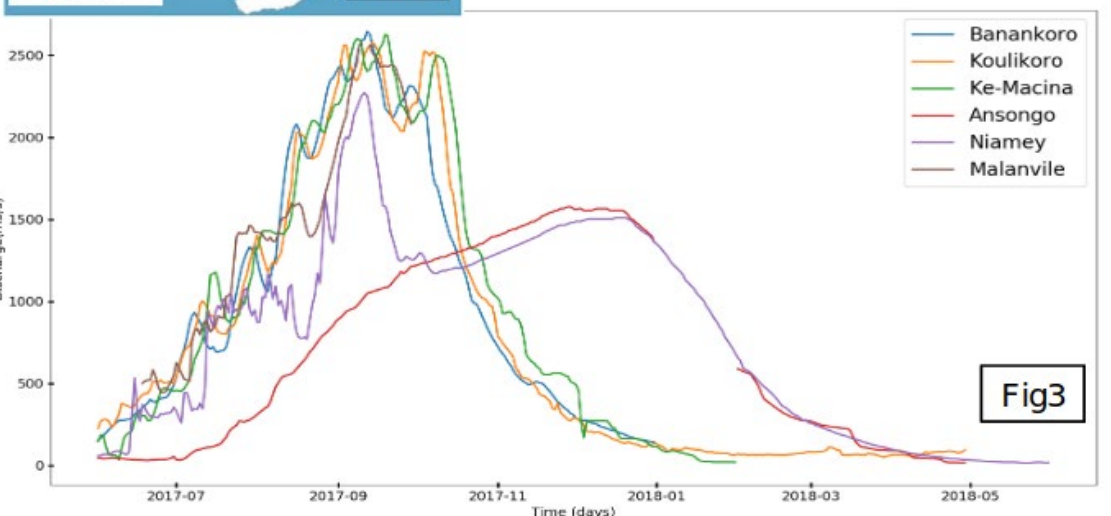
Active basin shared by 9 countries : Benin, Burkina Faso, Cameroon, Ivory Coast, Guinea, Mali, Niger, Nigeria and Chad

Specificities that make simulation difficult :

- Inner delta composed of vast floodplains in Mali
  - great source of evaporation
  - impact on river discharge seasonal outline (see Fig3)
- Flash floods in Niamey area due to small tributaries contribution

Flow modeling in the NRB crucial for :

- monitoring floods and droughts
- water allocation for agriculture
- water management decision making



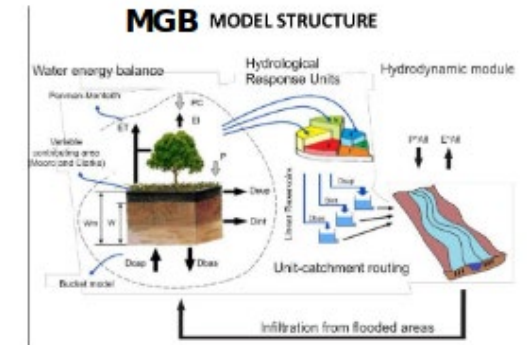


## Requirements for basin scale hydrological forecasting

- large observation datasets : at near to real time  
High temporal and spatial resolution



- low computational cost modeling solutions to account for the complexity of the domain



- Real time meteorological forcing (precipitation, temperature, ...)



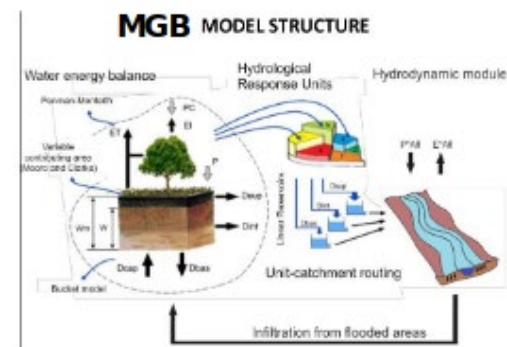


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Need for a monitoring and forecasting system that coordinates the tasks and interactions between the hydrological model and the real time data released on dedicated services





I - The Hydrological Forecasting system with Altimetry Assimilation (HYFAA)

II - OSSE

1. Methodology
2. Results

III - OSE

1. Preliminary results on Niger and Congo basins



## I - The Hydrological Forecasting system with Altimetry Assimilation (HYFAA)

### II - OSSE

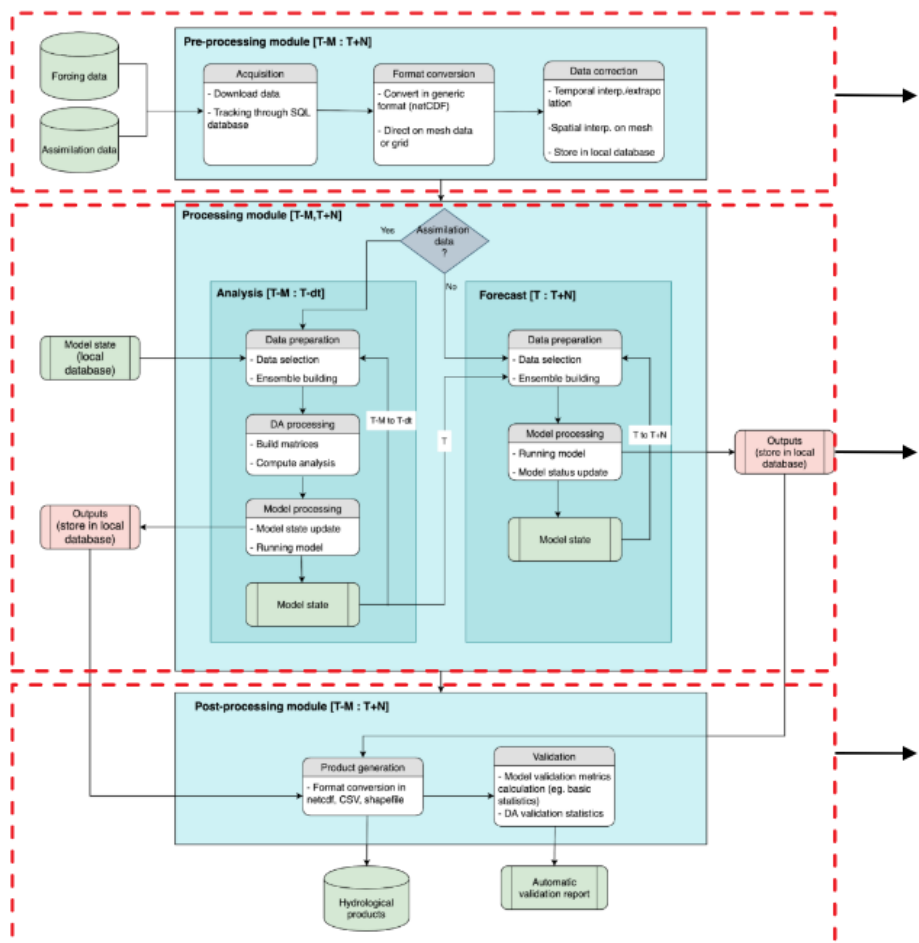
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## HYFAA : Hydrological Forecasting system with Altimetry Assimilation



Pre-processing : Update local forcing and assimilation database from external sources, and keep track of changes in data.

Processing :

- launches ensemble hydrological simulations
- manages coupling between the hydrological model and the EnKF filter
- updates model state database
- manages short to lon term hydrological forecasts.

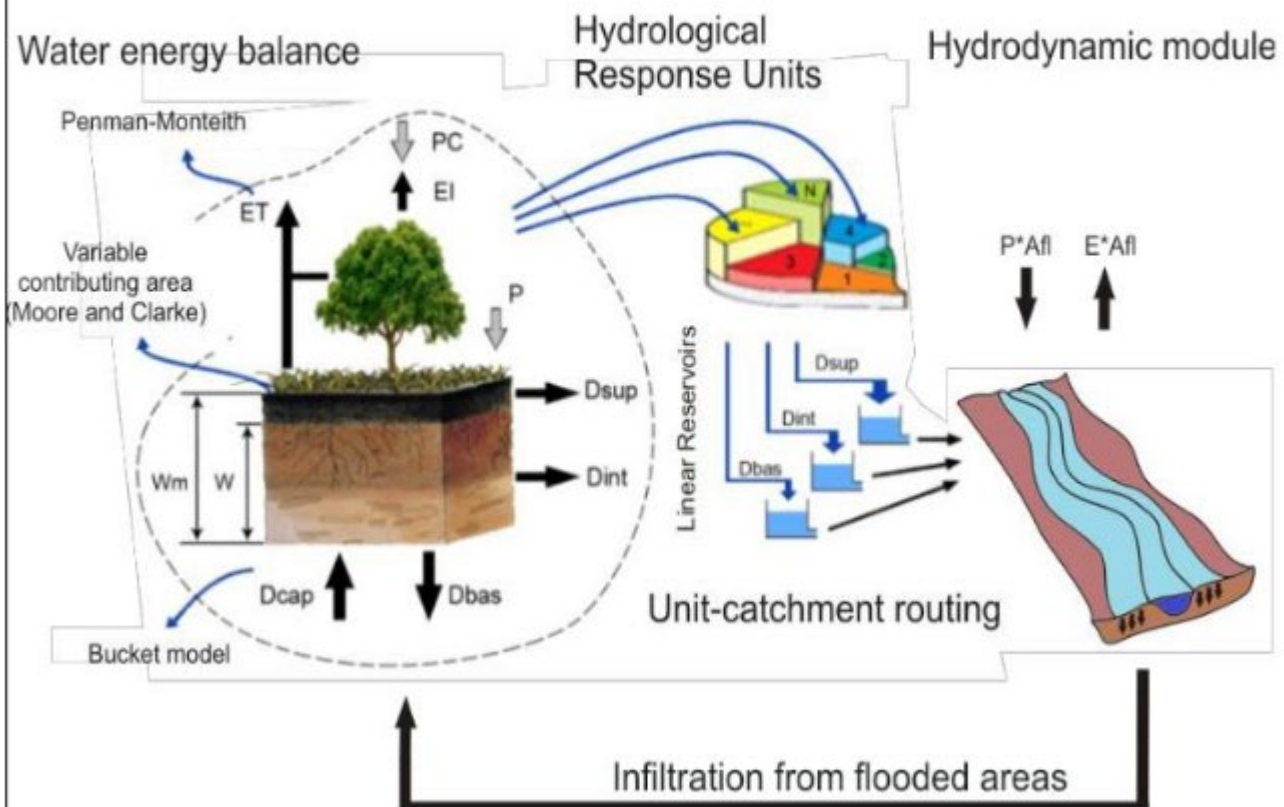
Post-processing : Generate value added products, and a report containing run diagnostics.





## The MGB hydrological model : a distributed large-scale model

### MODEL STRUCTURE



-vertical hydrologic balance (hydrologic module)

-flood propagation (hydrodynamic module)

Floodplains

bifurcations

2D flow in inner delta

Backwater

-Two-way coupling

Infiltration from flooded areas into soil

Dynamic evapotranspiration (soil & open water)

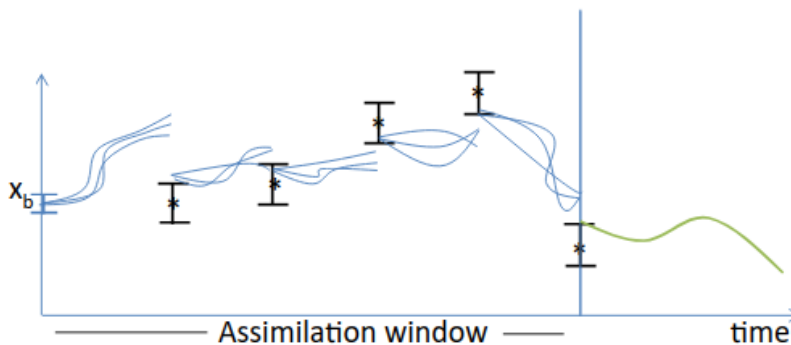
The model was calibrated over the period 1999-2014 against in-situ discharge observations (Fleischmann et al., 2018)



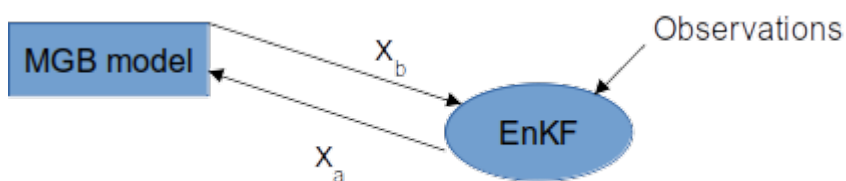


## The Ensemble Kalman Filter (EnKF)

- Stochastic method (Evensen, 2003) chosen for its ease of implementation
- Sequential : model is updated each time an observation is acquired.



- Externalized : communicates with MGB model through the scheduler



- $X_b$  : background matrix (ensemble of background vectors)
- $X_a$  : analyzed matrix (ensemble of analyzed vectors)

- EnKF gain matrix :  $K = P^f H^T [HP^f H^T + R]^{-1}$  with  $P^f$  as background covariance error matrix,  $R$  as observation error matrix,  $H$  observation operator.

- Can be easily adapted/plugged in to other modeling systems



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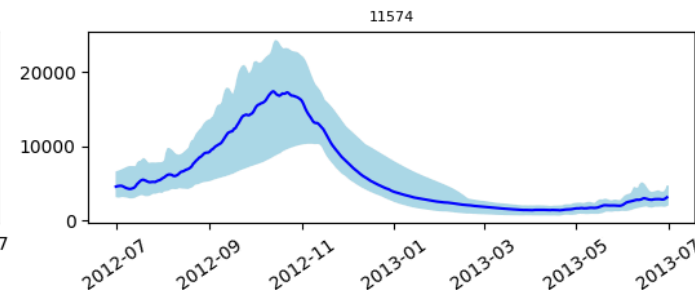
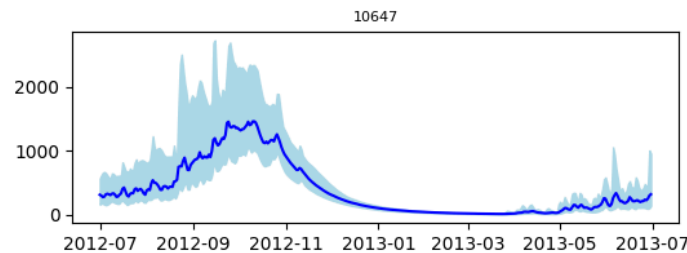
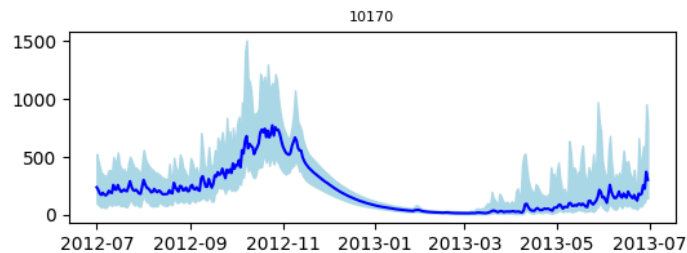
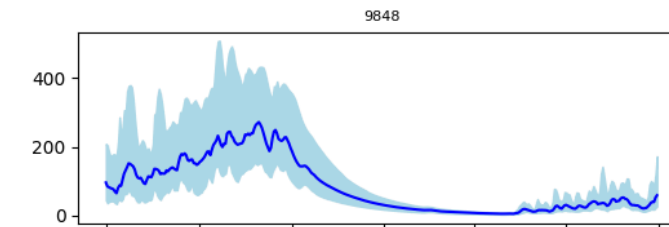
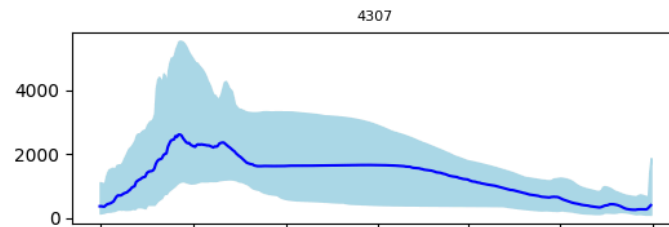
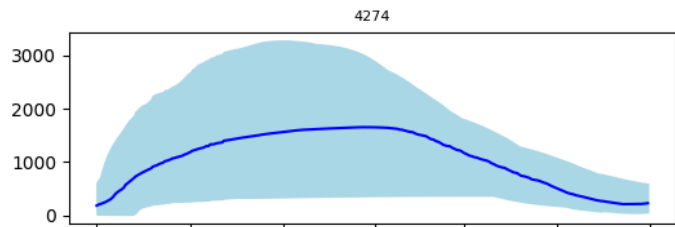
1. Preliminary results on Niger and Congo basins



## Observing Simulation System Experiment (OSSE)

Aim : validate the method by placing oneself in a framework where the uncertainties are known

- Generate perturbed ensemble of model realizations
  - perturbation applied to precipitation and hydrodynamics parameters (Manning, river width and depth)
- Pick one ensemble member to be the « truth »
- Simulate virtual observations  $Q_{obs} = Q_{model} + \varepsilon$  where  $\varepsilon$  is the observation error





## Observation systems

Daily in-situ discharge measurements provided by the Niger Basin Authority



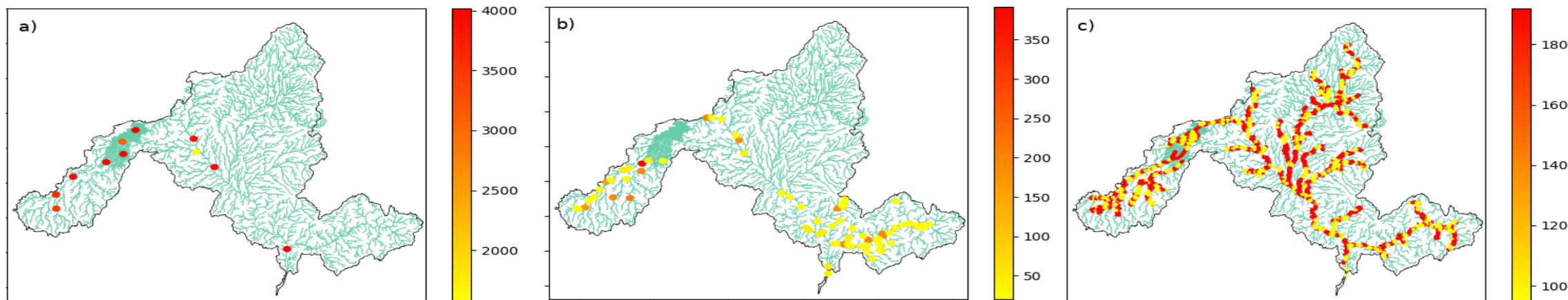
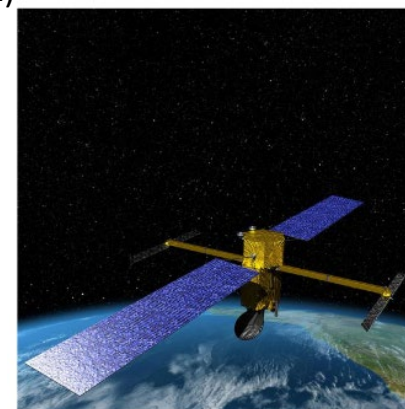
Hydroweb water levels released operationnaly on the Theia Land service



SWOT mission : will deliver high resolution products of water levels and discharge over rivers

Launch planned for end of 2022

SWOT track simulator (Biancamaria et al., 2011)

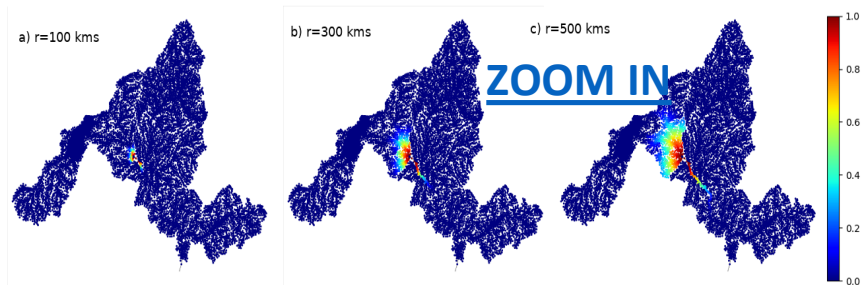


a) In-situ, b) Hydroweb and c) SWOT data coverage for the period July 2012 - December 2017.



## Added EnKF functionalities

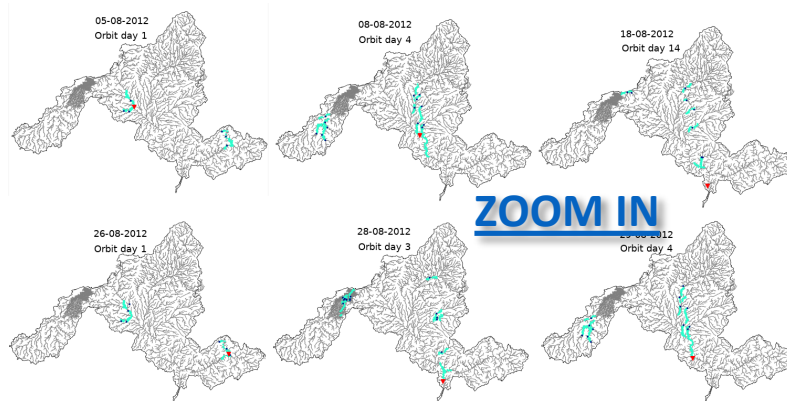
### Localization algorithm



Localization matrix in Niamey for  $r=100\text{kms}$  (a),  $r=300\text{kms}$  and  $r=500\text{kms}$ (c).

- Compensate for spurious correlations over long spatial distances

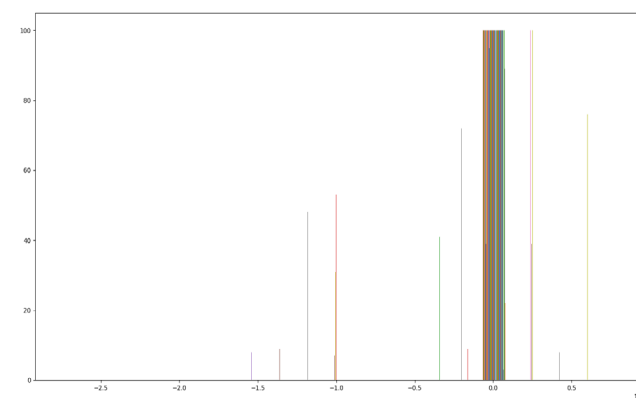
### Selection method



Example maps of data selection applied to SWOT observations. The maximum variance location is represented by a red triangle. SWOT observations for each day of acquisition are in light blue and selected observation are in dark blue.

- Nb of obs  $\gg$  ensemble size
- leads to inversion pbs
- > obs selection method that only keeps uncorrelated obs

### Quality control



Innovation distribution in the case of real Hydroweb data

→ innovation clipping

- Statistical outliers in observation errors leads to divergence issues



## Questions addressed in this study

Considering discharge data assimilation as a way to improve hydrological forecasting :

- What should be the characteristics of a generic system that assimilates these data from different satellite missions and on several basins ? For example, which functionalities work best on a given basin or with a given observation system ?
- What are the assets and limitations of such forecasting systems ?



## Study configuration

### General configuration

#### Model

- Simulation TAPEER 2012-2017
- Climate forcing (except rain) [climatologies] air T°C, wind, humidity, atm pression, solar radiation, vegetation data, albedo, LAI, tree height, superficial resistance
- Calibration : against obs
  - Model parameters (soil, flow, hydrodynamic propagation) and forcing (vegetation)
  - Geometry: (floodplains)
- Evaluation/validation : against obs In situ data from Niger Basin Authority (discharge)

#### EnKF

Ensemble size : 100  
Observed variable : discharge

### Experiments

#### EnKF varying configuration

- Background vector :
  - all state variables (soil moisture, baseflow, subsurface and surface water, canopy interception, runoff, chanel water depth, river water storage, flooded are, flow interconnections)
  - parameters : river width, river depth, Manning coefficient
- Functionalities
  - localization
  - observations selection
  - observations quality control

#### OSSEs

- Simulated observations  
simplified framework where all uncertainties are known
- 3 observing systems : in-situ, Hydroweb, SWOT
- Reference simulation : 28th member of the ensemble

#### OSEs

- Real observations
- Hydroweb
- Extension to the Congo river basin





## General overview

Exp1			
Exp2			
Exp3			
Exp4			
Exp5			
Exp6			
Exp7			
Exp8			
Exp9			
Exp10			

10 experiments



## General overview

	Observing system		
Exp1	In situ		
Exp2			
Exp3	Hydroweb		
Exp4			
Exp5			
Exp6	SWOT		
Exp7			
Exp8	SWOT river		
Exp9	SWOT		
Exp10			

10 experiments  
3 observing systems



## General overview

	Observing system	Control vector	Filter
Exp1	In situ	sv	
Exp2		sv+par	
Exp3	Hydroweb	sv	
Exp4		sv+par	
Exp5		sv	
Exp6	SWOT	sv+par	
Exp7		sv+par	
Exp8	SWOT river	sv+par	
Exp9	SWOT	sv+par	
Exp10		sv+par	

10 experiments

3 observing systems

Control vector : state variables (sv), parameters (par)



## General overview

	Observing system	Control vector	Filter
Exp1	In situ	sv	EnKF
Exp2		sv+par	EnKF
Exp3	Hydroweb	sv	EnKF
Exp4		sv+par	EnKF
Exp5		sv	LEnKF
Exp6	SWOT	sv+par	EnKF
Exp7		sv+par	EnKF
Exp8	SWOT river	sv+par	EnKF
Exp9	SWOT	sv+par	OS+EnKF
Exp10		sv+par	OS+LEnKF

10 experiments

3 observing systems

Control vector : state variables (sv), parameters (par)

Filter : EnKF

Localized EnKF (LEnKF)

Observations' selection (OS)



## General overview

	Observing system	Control vector	Filter
Exp1	In situ	sv	EnKF
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Exp4		sv+par	EnKF
Exp5		sv	LEnKF
Exp6	SWOT	sv+par	EnKF
Exp7		sv+par	EnKF
Exp8	SWOT river	sv+par	EnKF
Exp9	SWOT	sv+par	OS+EnKF
Exp10		sv+par	OS+LEnKF

10 experiments

3 observing systems

Control vector : state variables (sv), parameters (par)

Filter : EnKF

Localized EnKF (LEnKF)

Observations' selection (OS)

$$NRMSE = \left[ \sqrt{\frac{\sum_{i=1}^n (y_i - \hat{y})^2}{n}} \right] / \sigma$$



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1. Methodology

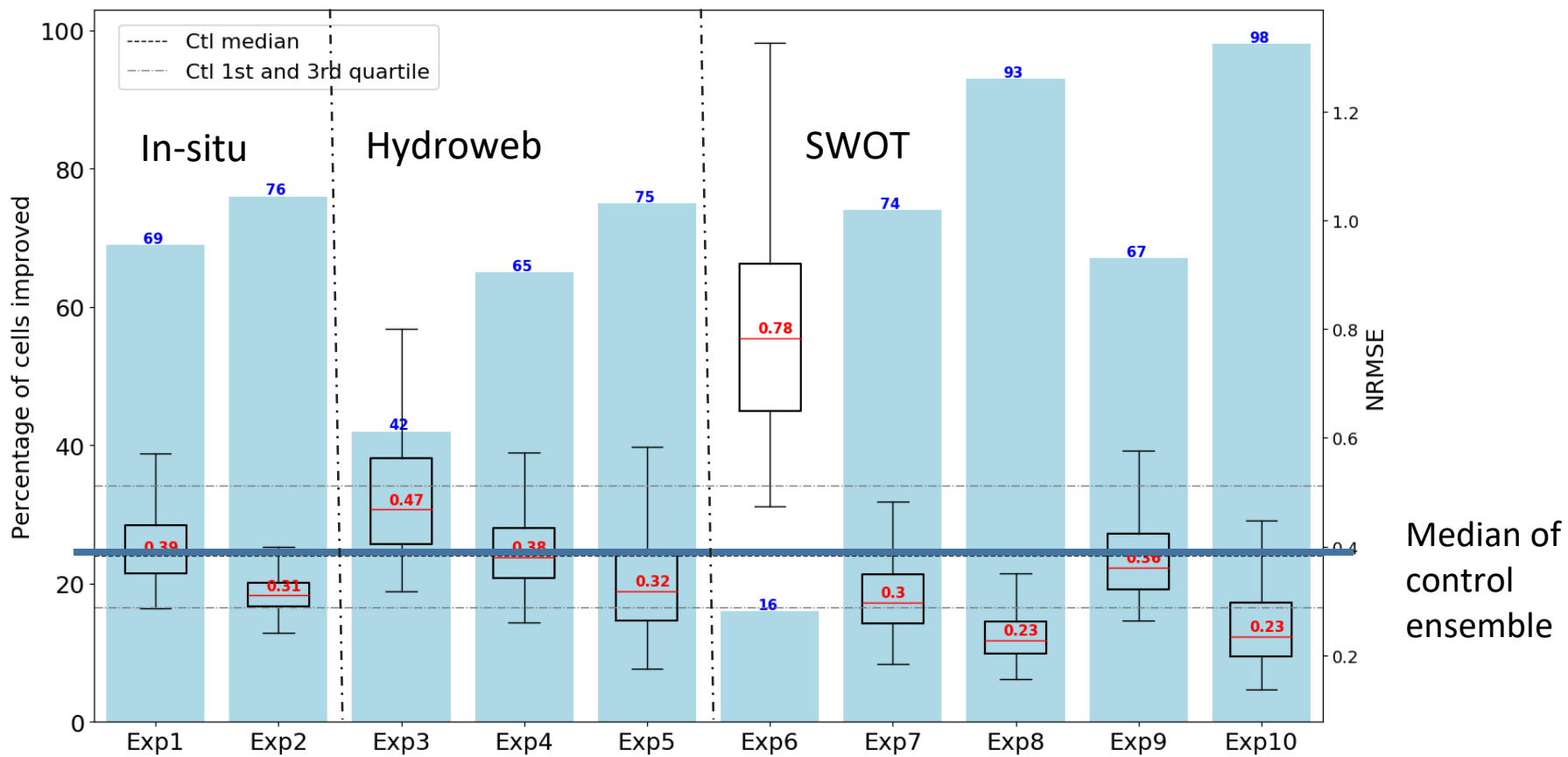
**2. Results**

III - OSE

1. Preliminary results on Niger and Congo basins



## General results



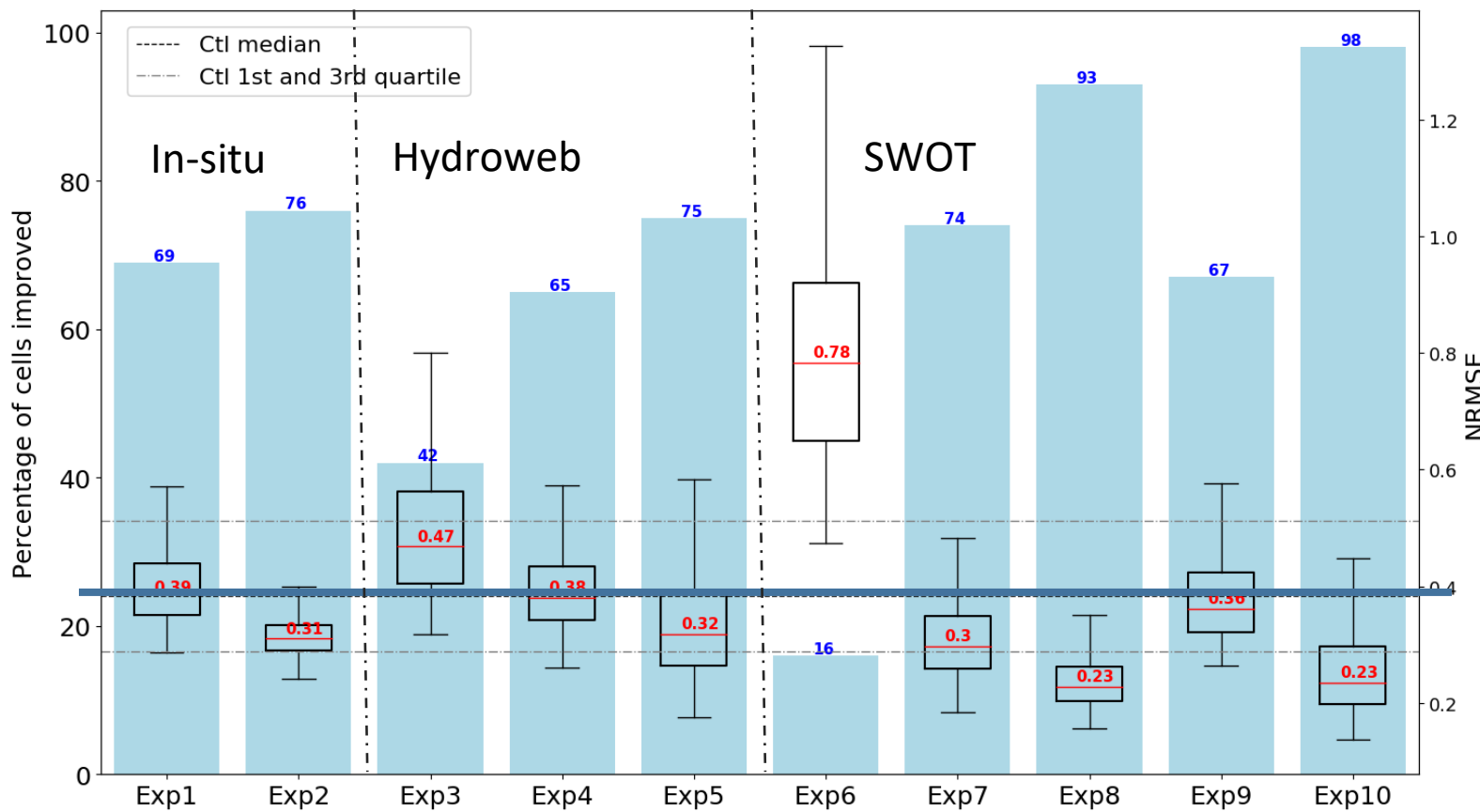
Boxplots of discharge NRMSE over the basin





## General results

Parameters' correction  
consistently improves the  
performance of DA



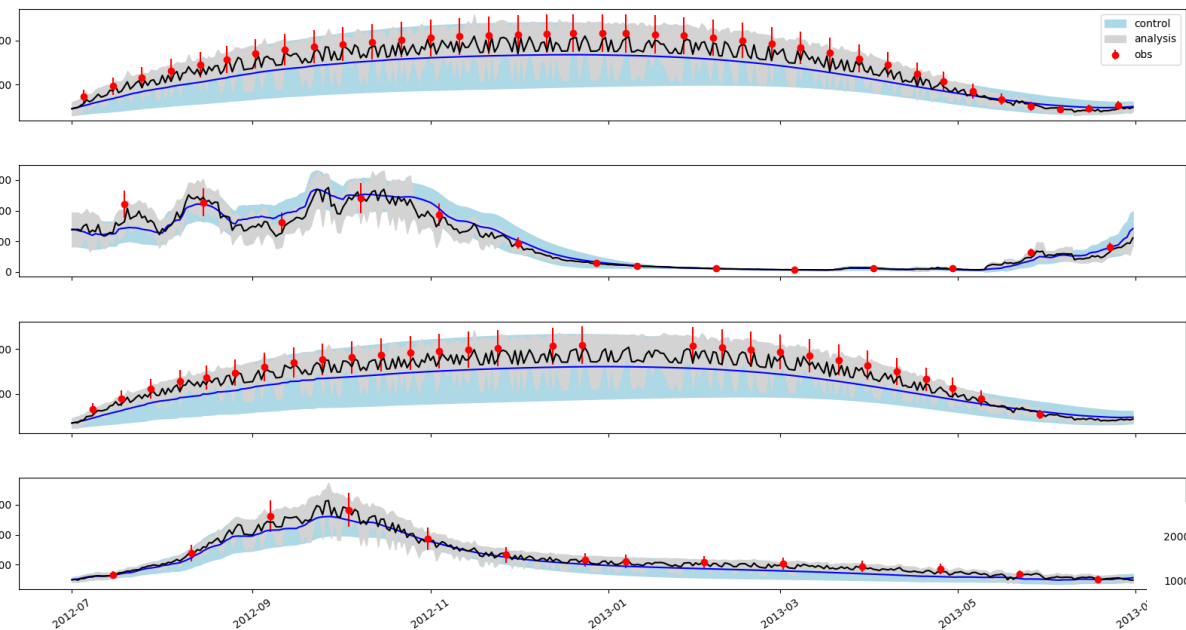
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Median of  
control  
ensemble



# OSSE - Results

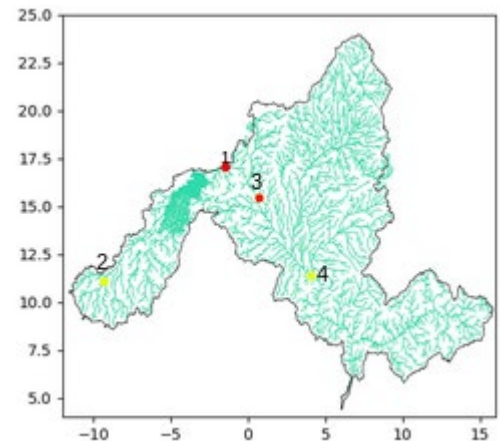
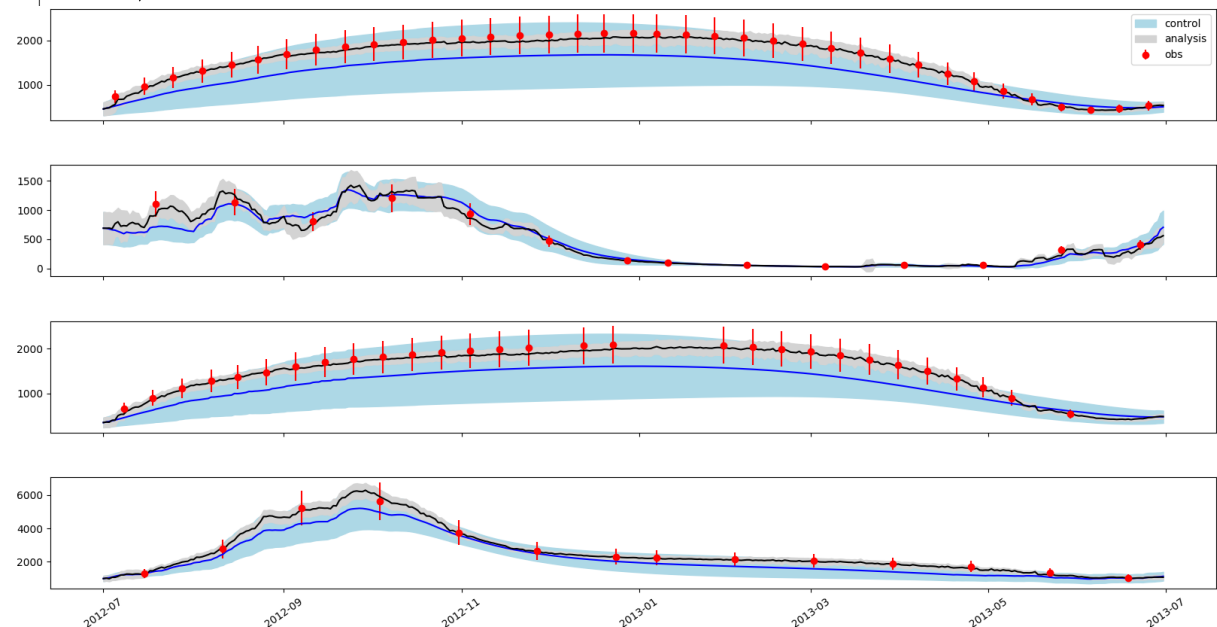
## Without parameters correction



← Correction of state variables only

Correction of state variables + parameters (Manning, river width and depth)

## With parameters correction



→ Correction of parameters stabilizes the signal between two observation dates and reduces model uncertainties

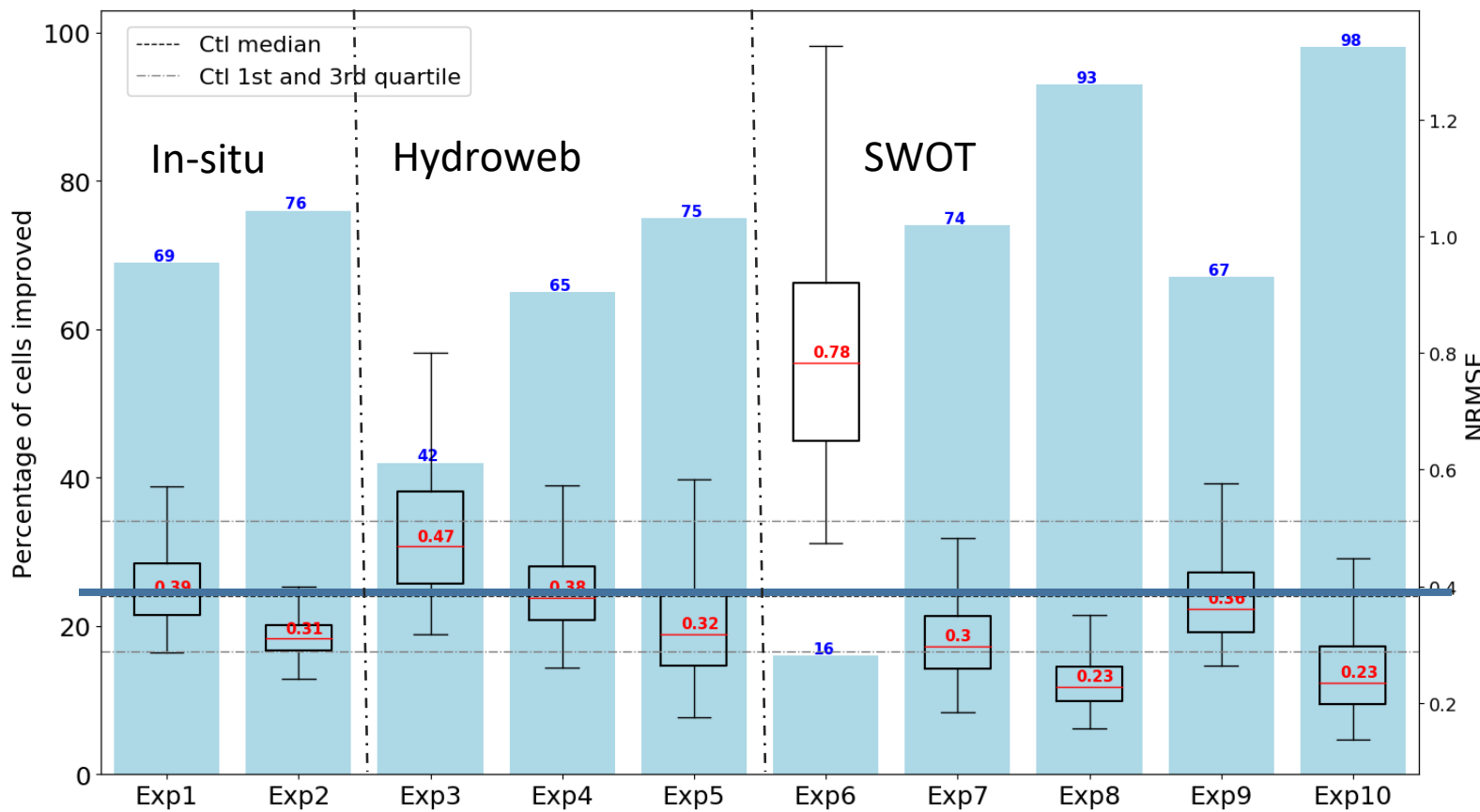
Control/Analyzed (blue/black) discharge and their range of uncertainty, observations (red)

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## General results



Boxplots of discharge NRMSE over the basin

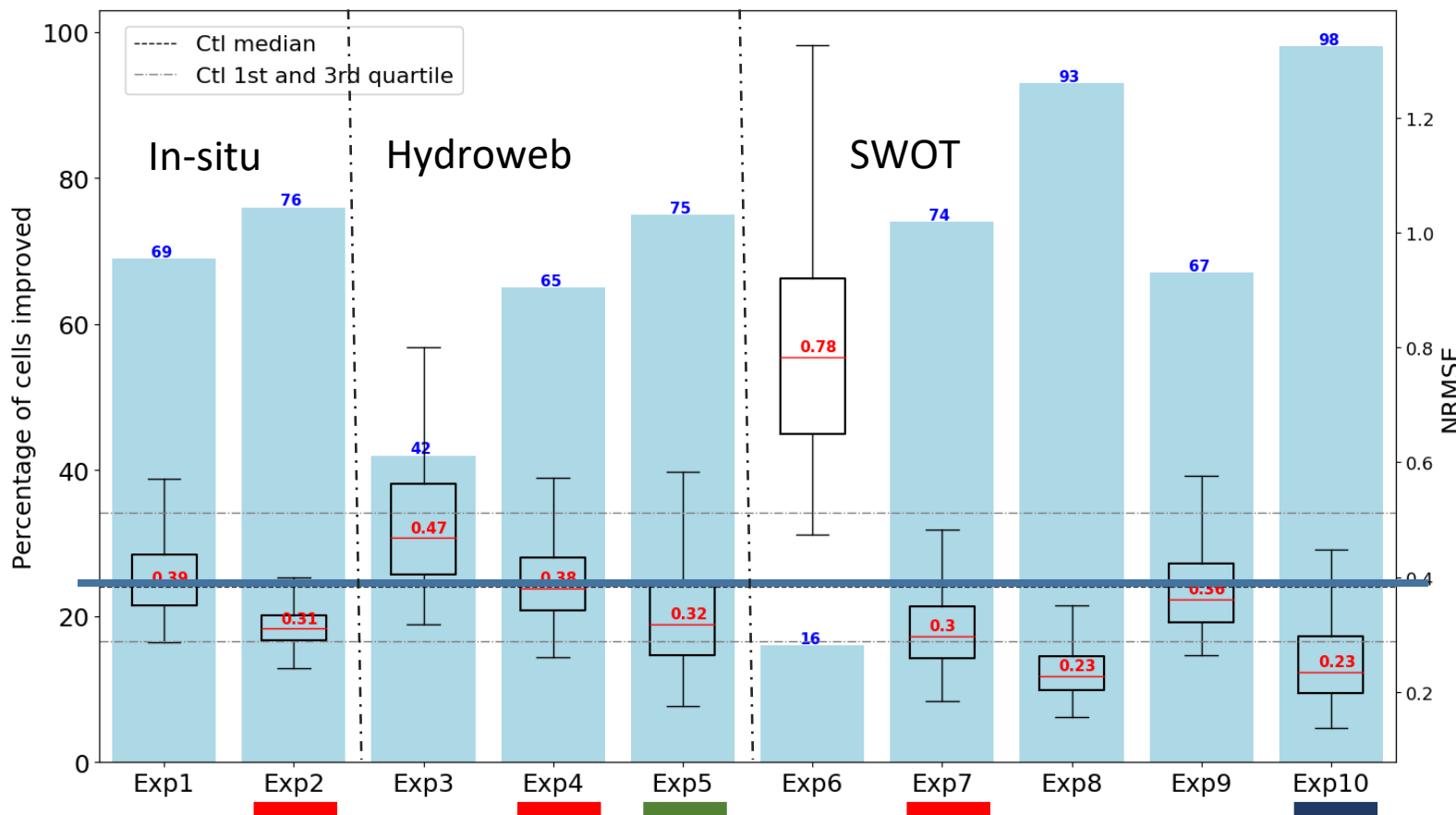
Parameters' correction consistently improves the performance of DA

Hydroweb DA works best with LenKF

Median of control ensemble



## General results



Boxplots of discharge NRMSE over the basin

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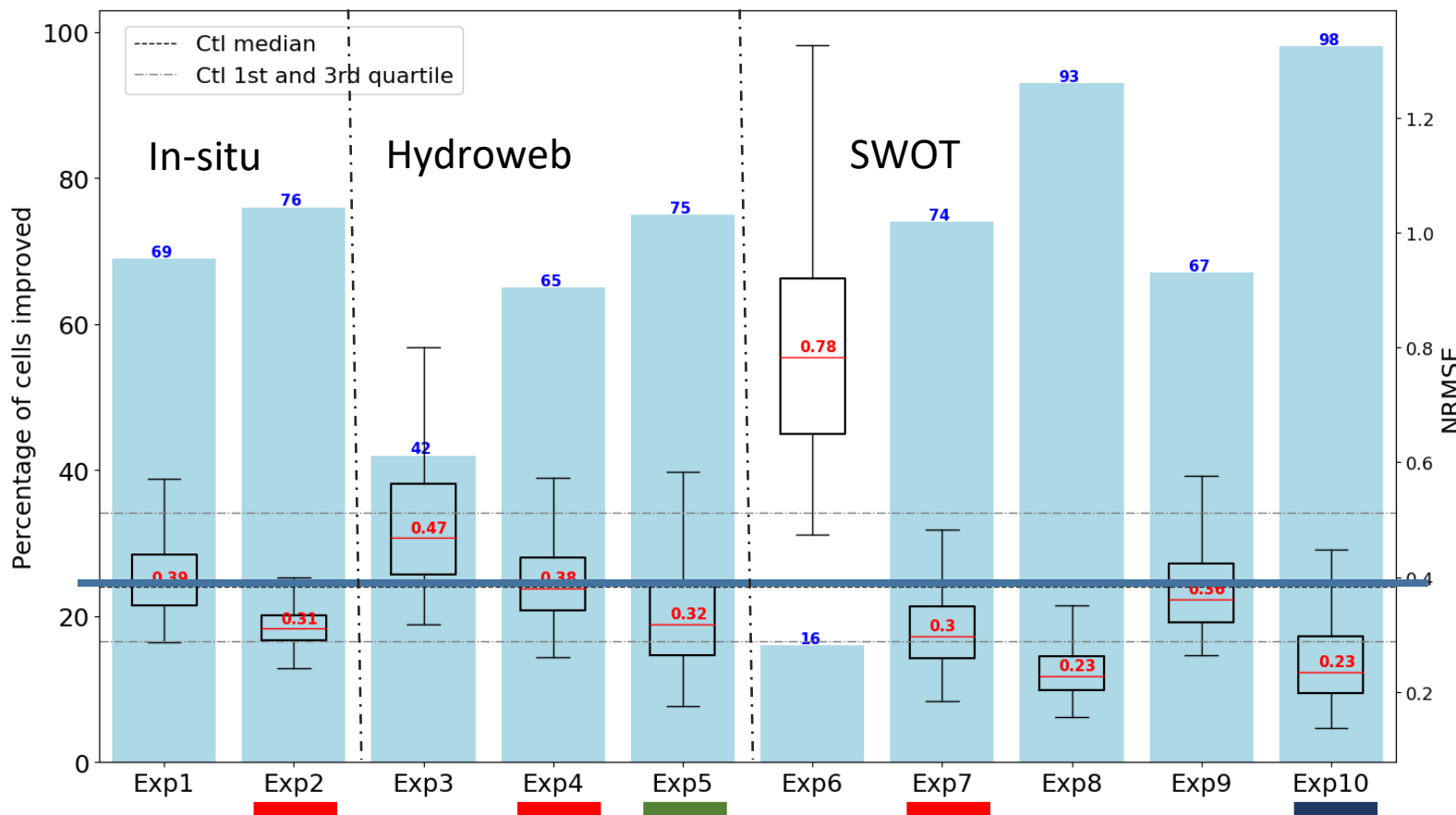
Hydroweb DA works best with LenKF

SWOT DA works best with LenKF and Obs selection

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## General results



Boxplots of discharge NRMSE over the basin

Parameters' correction consistently improves the performance of DA

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Median of control ensemble

Best NRMSE : Exp 8 & 10 (both SWOT)  
Best percentage of cells improved : Exp10 (SWOT)



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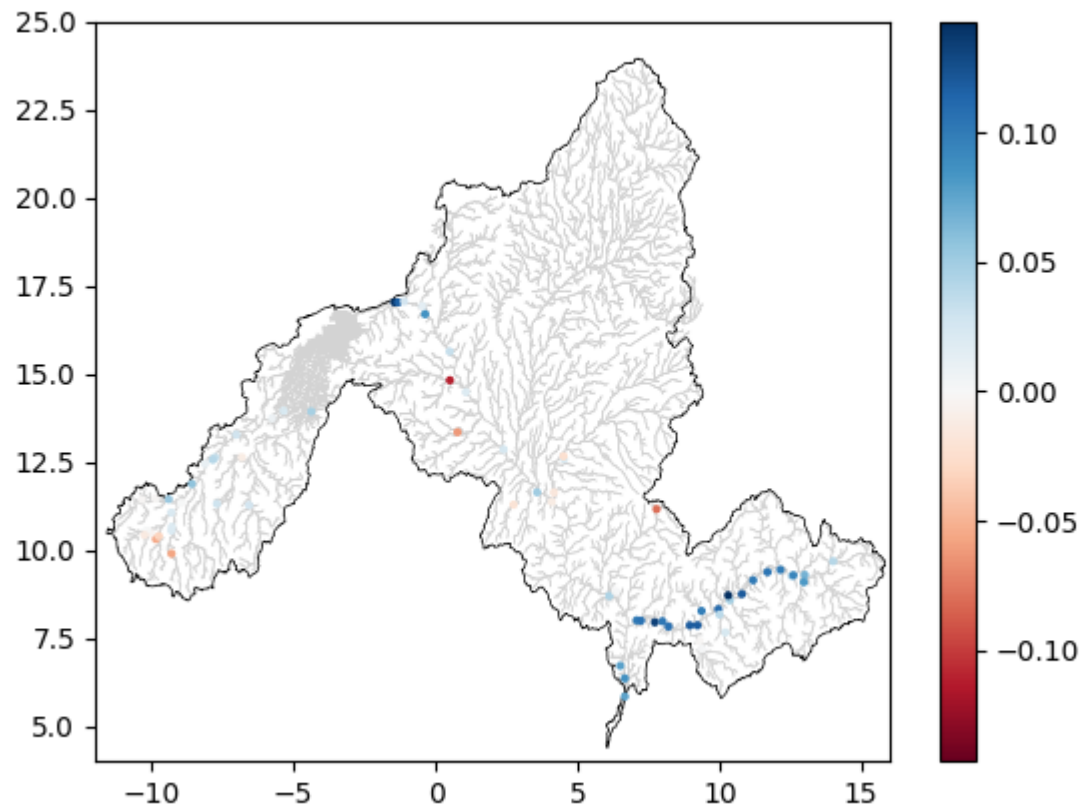
1. Preliminary results on Niger and Congo basins



# OSE - Preliminary results

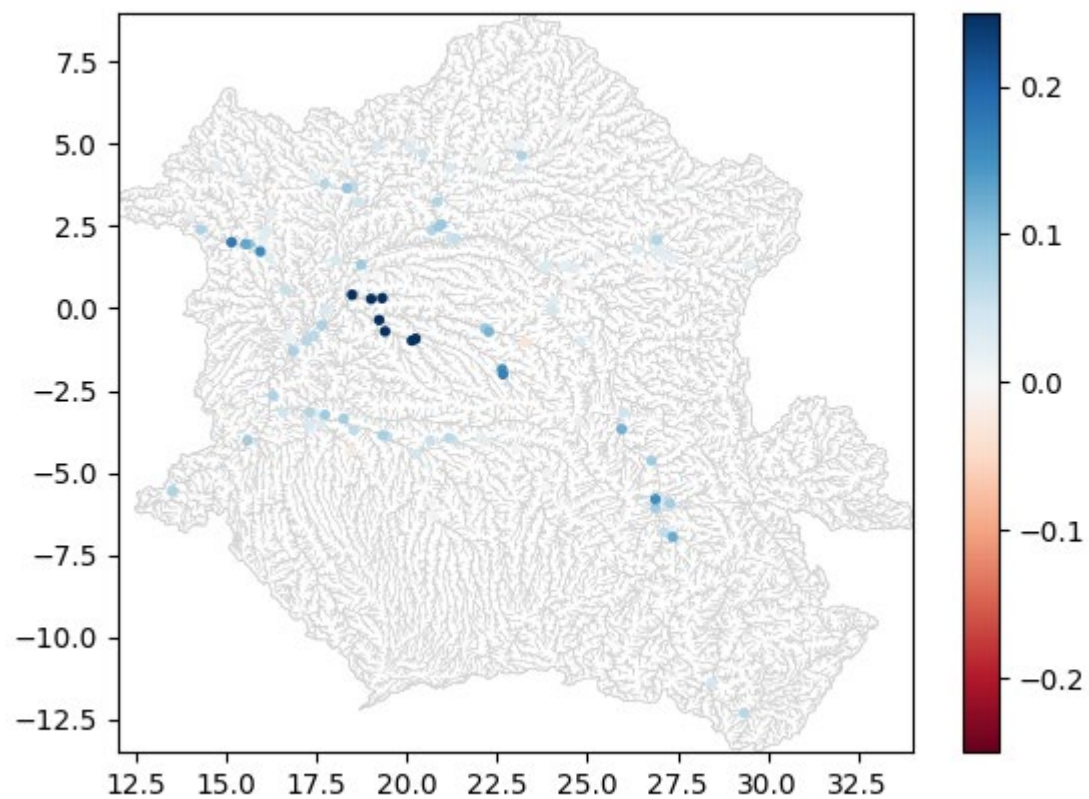
## Niger

NRMSE Ctl - Anl



## Congo

NRMSE Ctl - Anl



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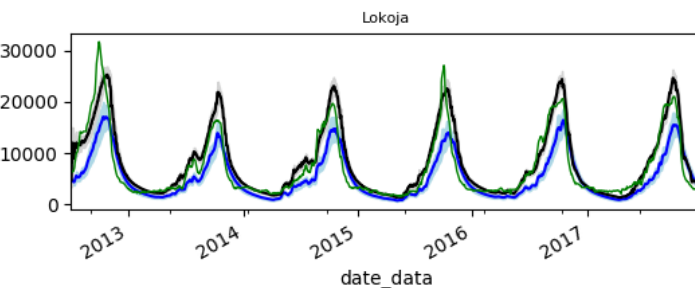
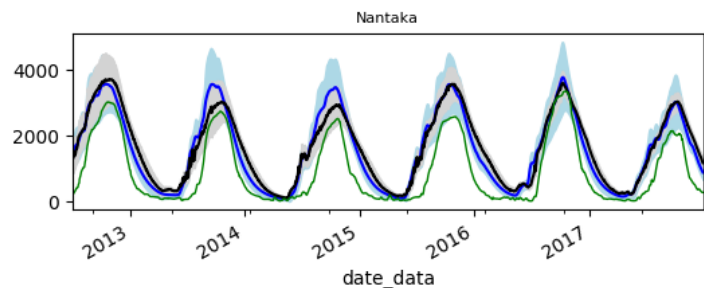
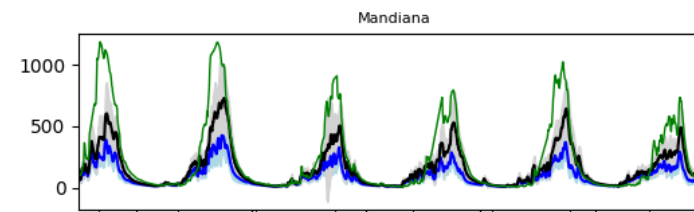
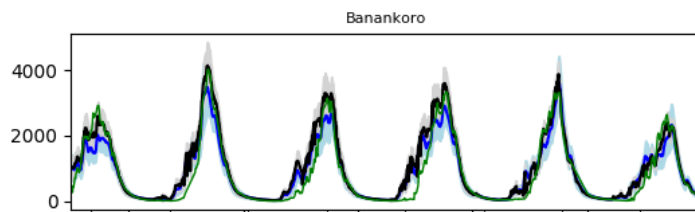
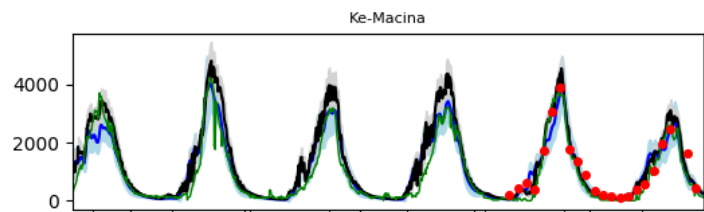
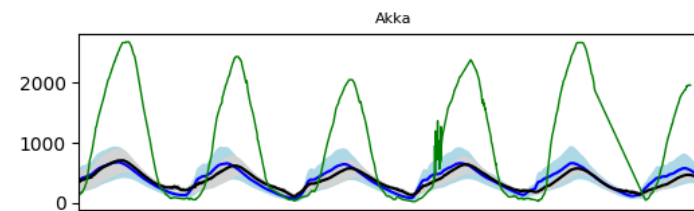
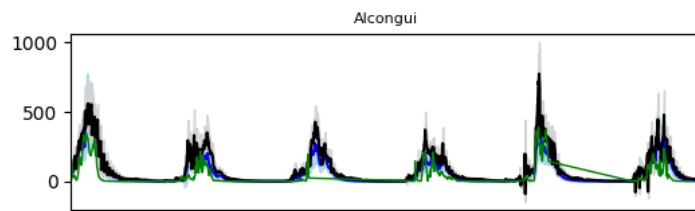
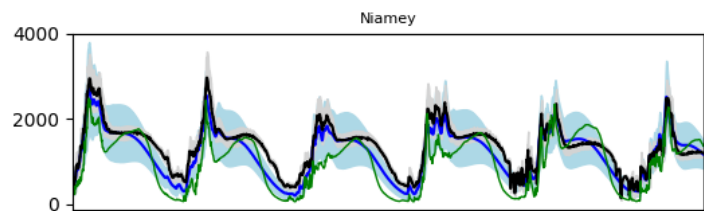
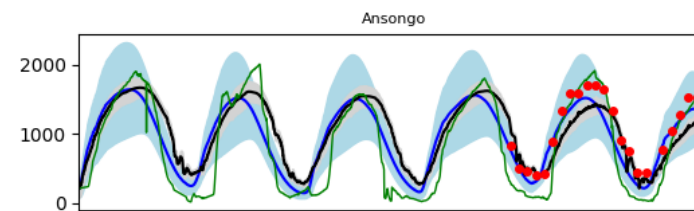
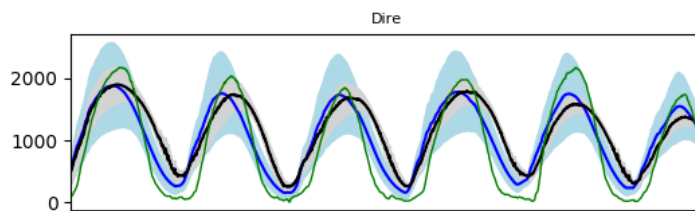
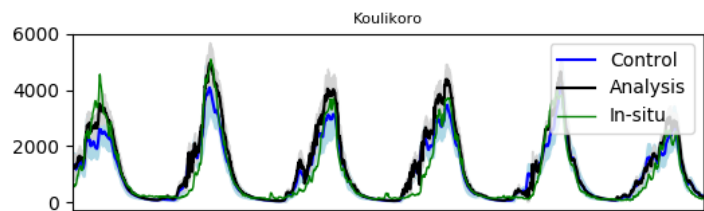




# OSE - Preliminary results

Niger river Discharge

- Control
- Analysis
- In-situ



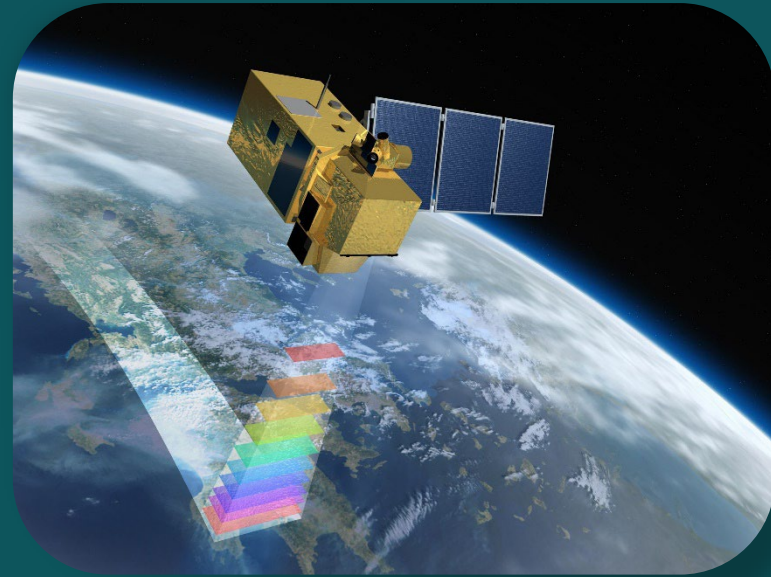


## Conclusions

- Data assimilation helps to improve discharge estimation over the Niger river basin
- The correction of parameters improves the performance of DA by introducing longer term corrections (could be forcing biases)
- The DA system's functionalities can be adapted to the characteristics of the observing systems
- The assimilation of real observations needs to be improved, in particular the management of outliers can be an issue
- SWOT data could contribute to improve large scale hydrological forecasting



Thank you for your  
attention.



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