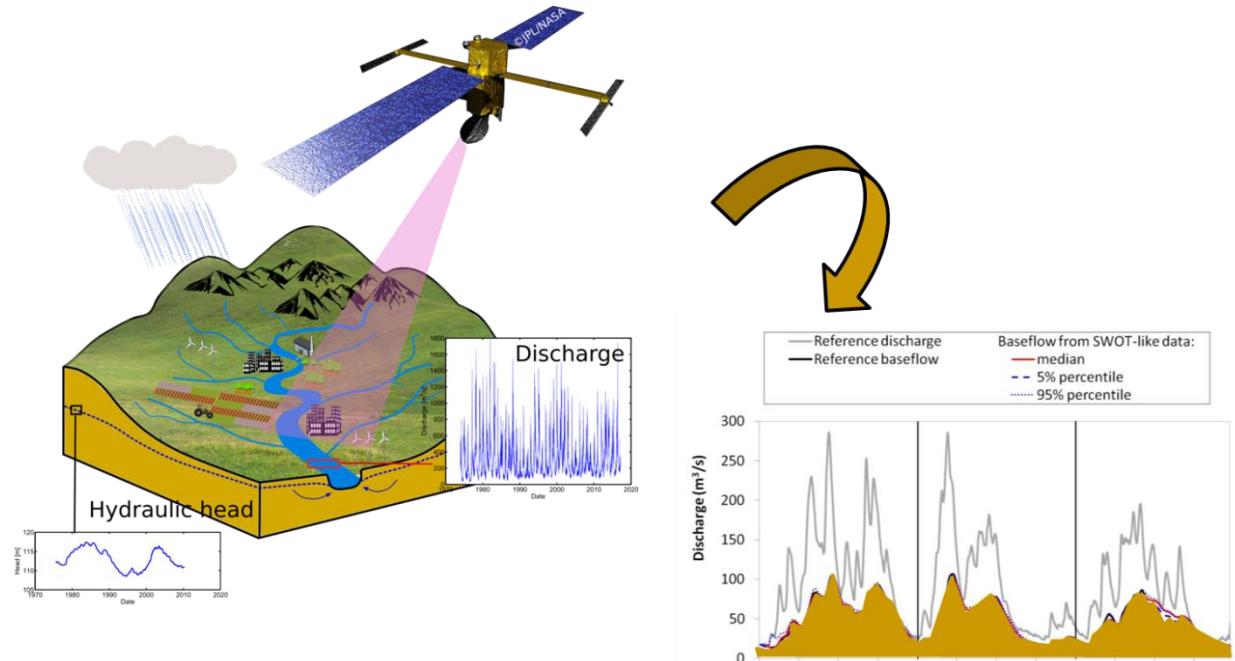


Retrieving baseflow of large rivers from space with the future SWOT mission



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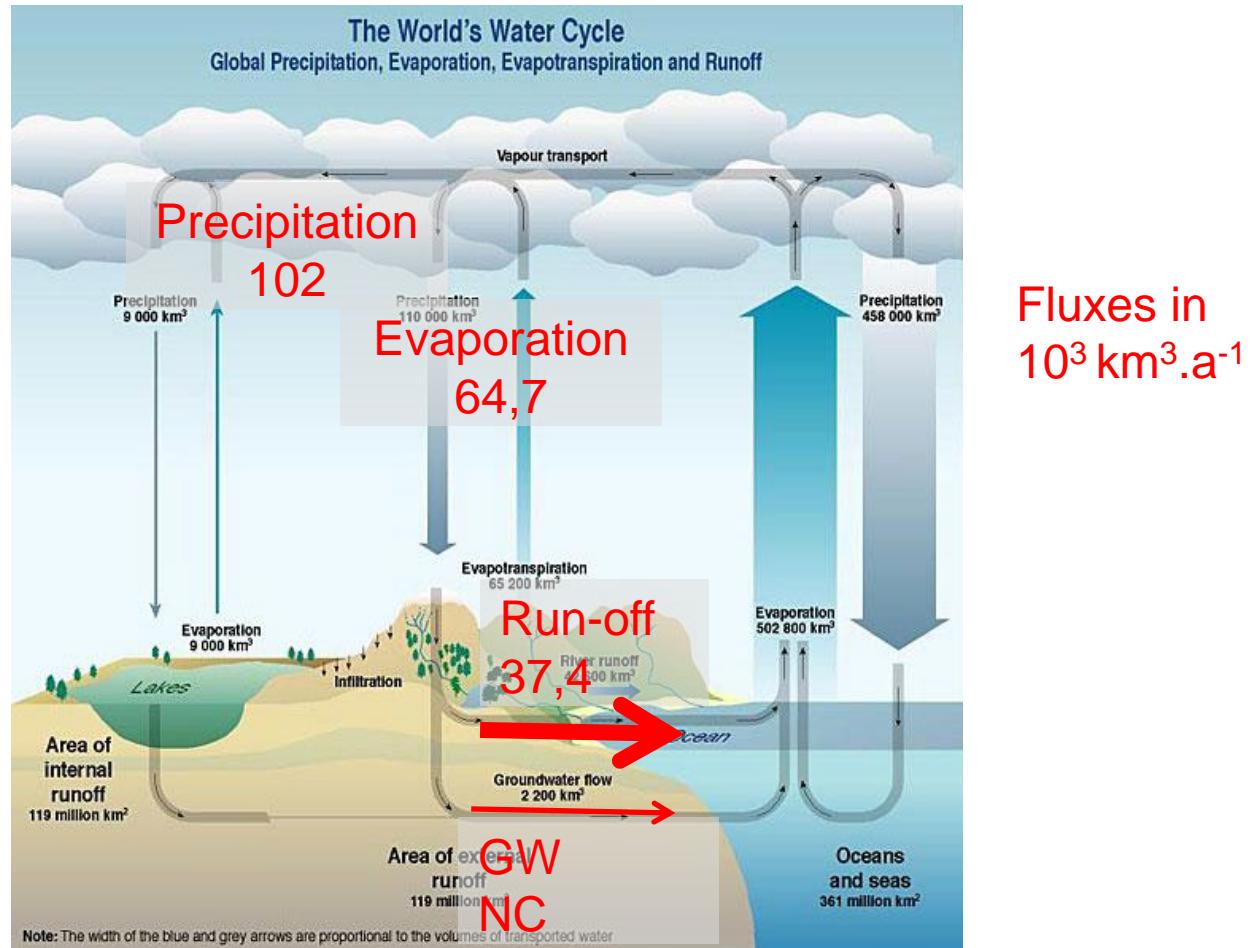


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Water Cycle

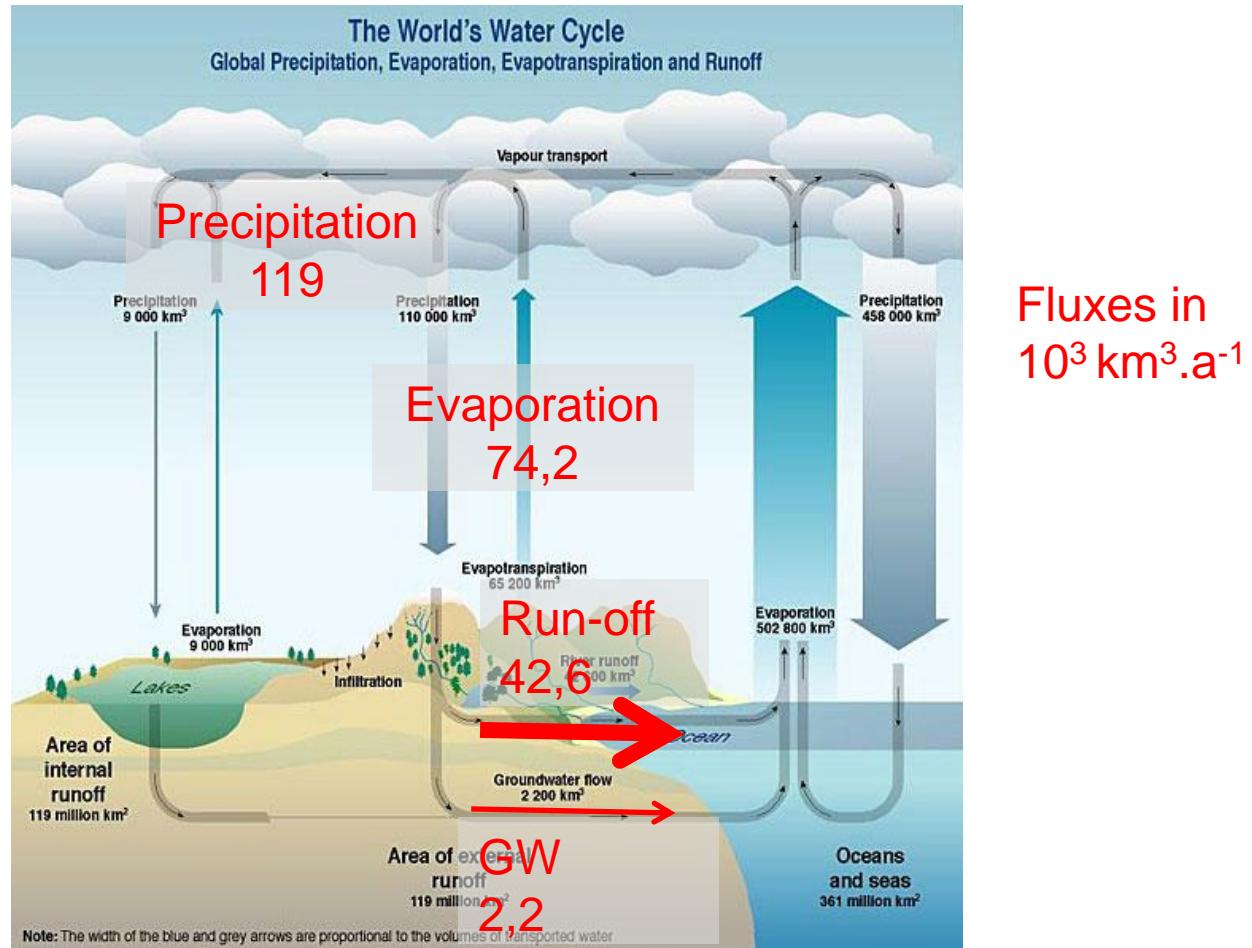
- From observations and coarse approximations¹ ...



¹Synthesis by Lvovitch 1978 World Water Balance

Water Cycle

- From observations and coarse approximations¹ ...

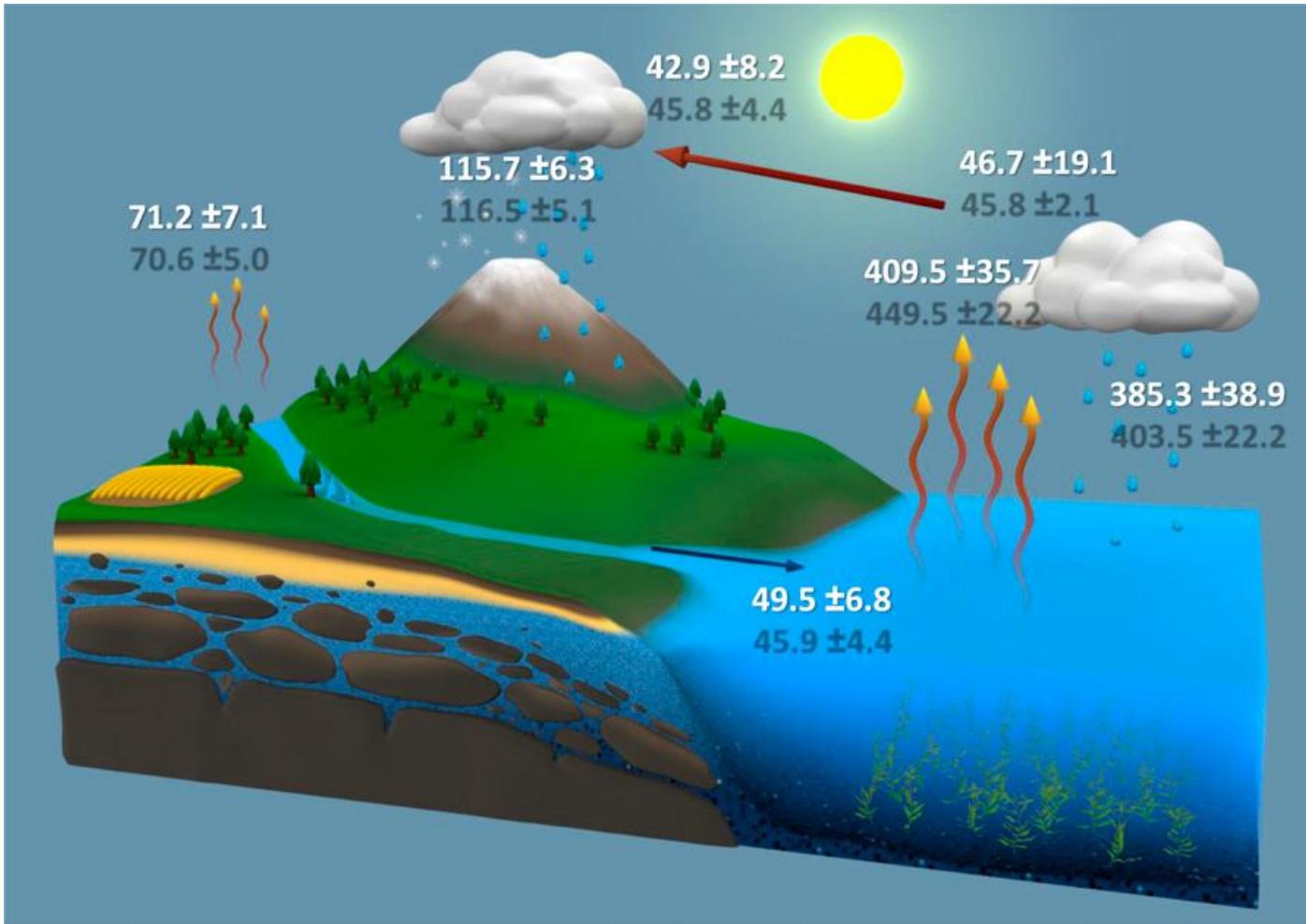


¹Synthesis by Shiklomanov 1994 for UNESCO based on earlier studies

Water Cycle

Fluxes in
 $10^3 \text{ km}^3 \cdot \text{a}^{-1}$

- To remote sensing and models²

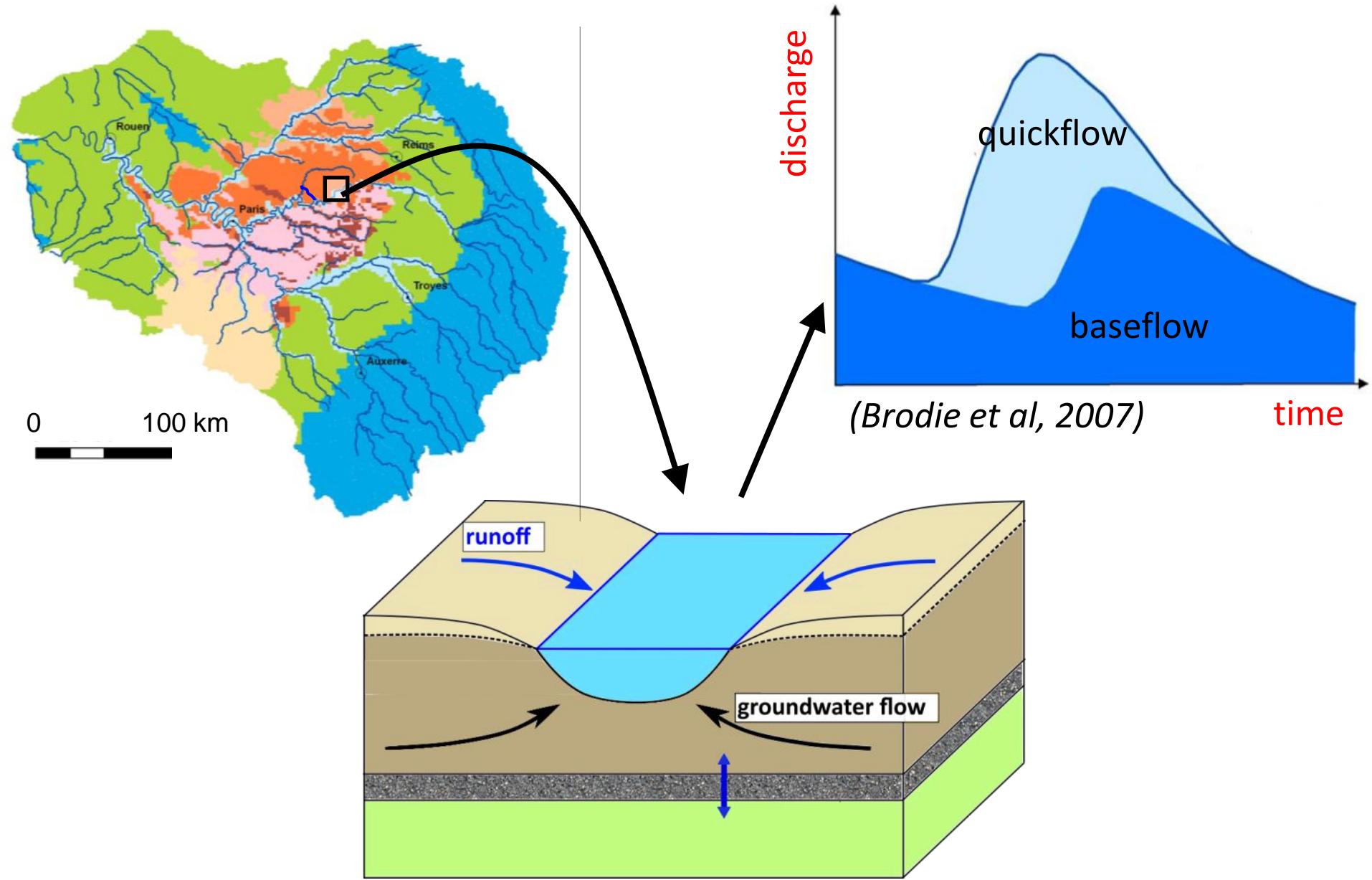


²Synthesis by Rodell et al. 2015 Journal of Climate

How to assess groundwater contribution to river flow?

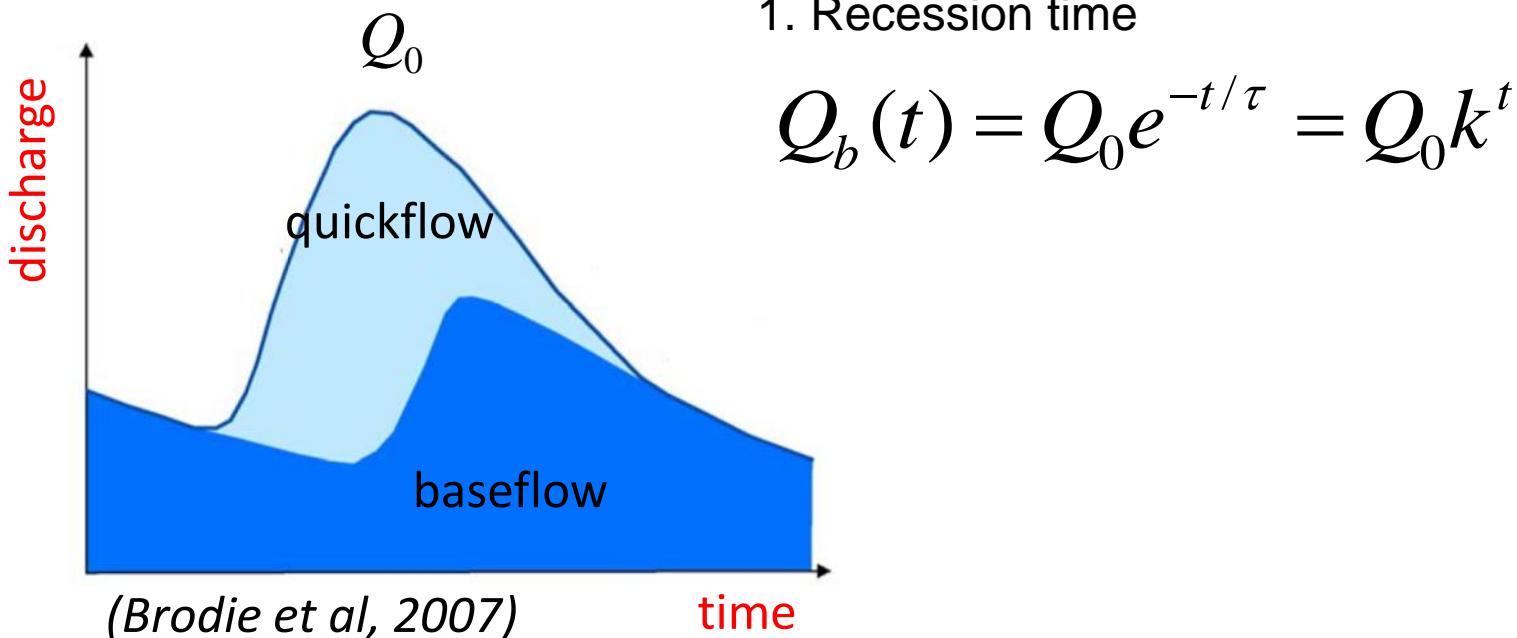
- **Estimate surface-groundwater exchanges**
 - No direct measurements in most of the cases even at the local scale
 - Seepage meter
- **Remote sensing : gravimetry with GRACE (Tapley et al, 2004, GRL)**
 - Difficult to deconvolute various processes (lake, soil, river floodplain, aquifer units) but feasible for large basins → Amazon River Basin (Tourian et al., 2018, WRR)
- **~~Model needs a proper conceptualisation and validation~~**
 - Flux data is the most valuable information for GW system calibration (Hunt et al 2006, JH)

River baseflow in a river basin



River baseflow

- Use of the Chapman filter as a proof of concept (Chapman 1999, HP) based on Lyne and Hollick (1979)



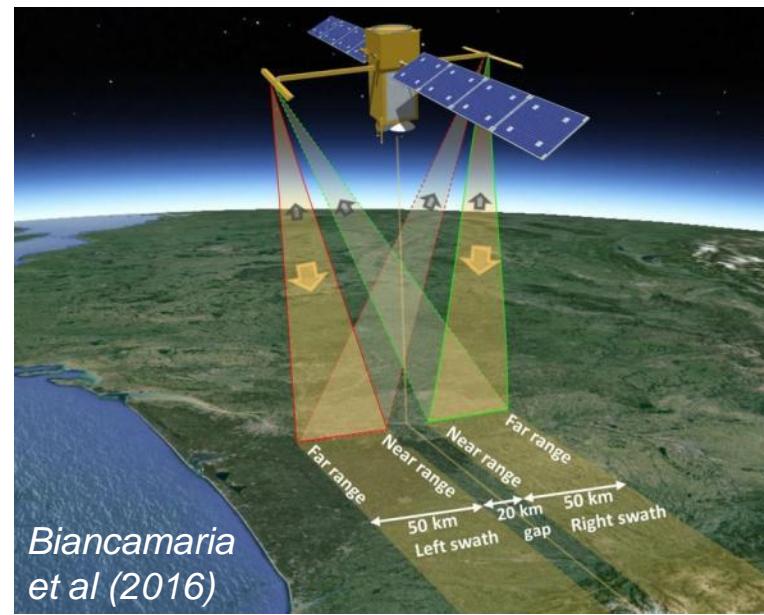
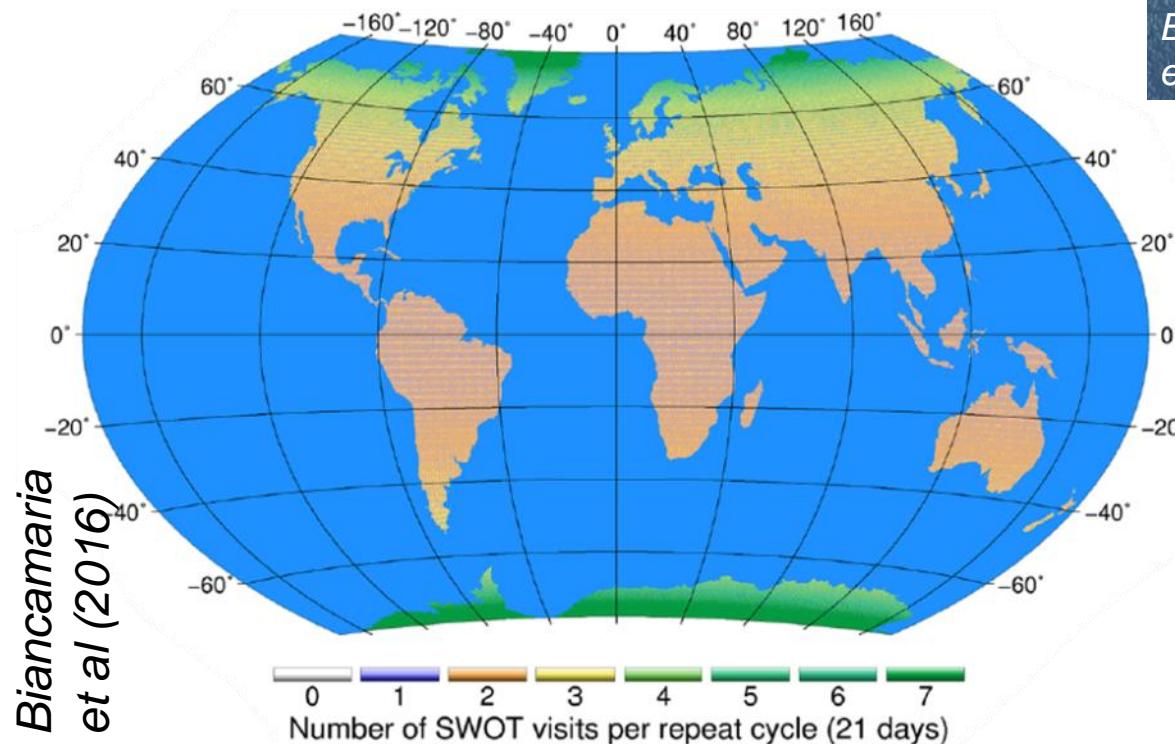
- 2. Recursive filter

$$Q_b(t) = \frac{k}{2-k} Q_b(t-1) + \frac{1-k}{2-k} Q(t)$$

River discharge from SWOT

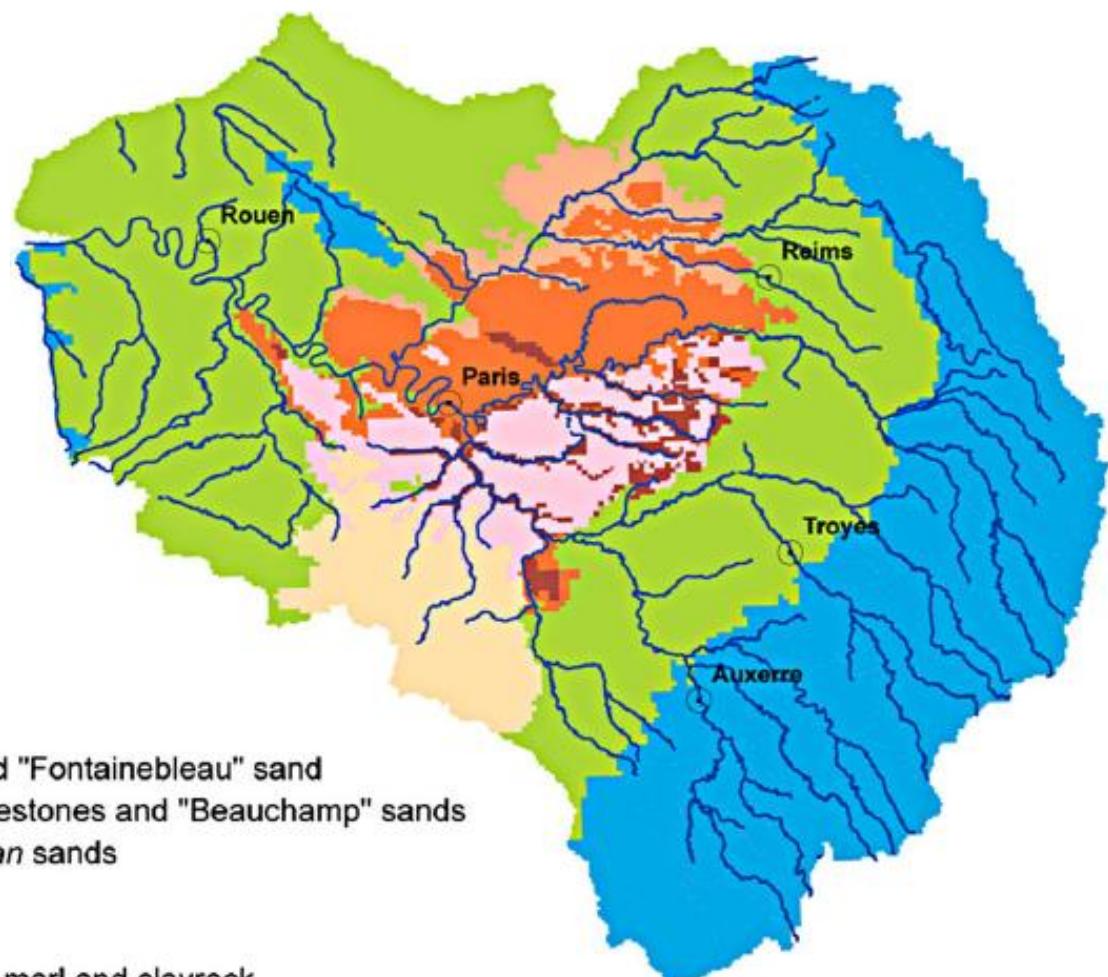
SWOT (Surface Water and Ocean Topography)

- oceans - land surface water topography
- **river discharge**
- river width > 100 m
- improves GRDC observation extent



→ Can **baseflow** be estimated at **global** scale from **SWOT** observations during the mission lifetime?

The Seine River Basin

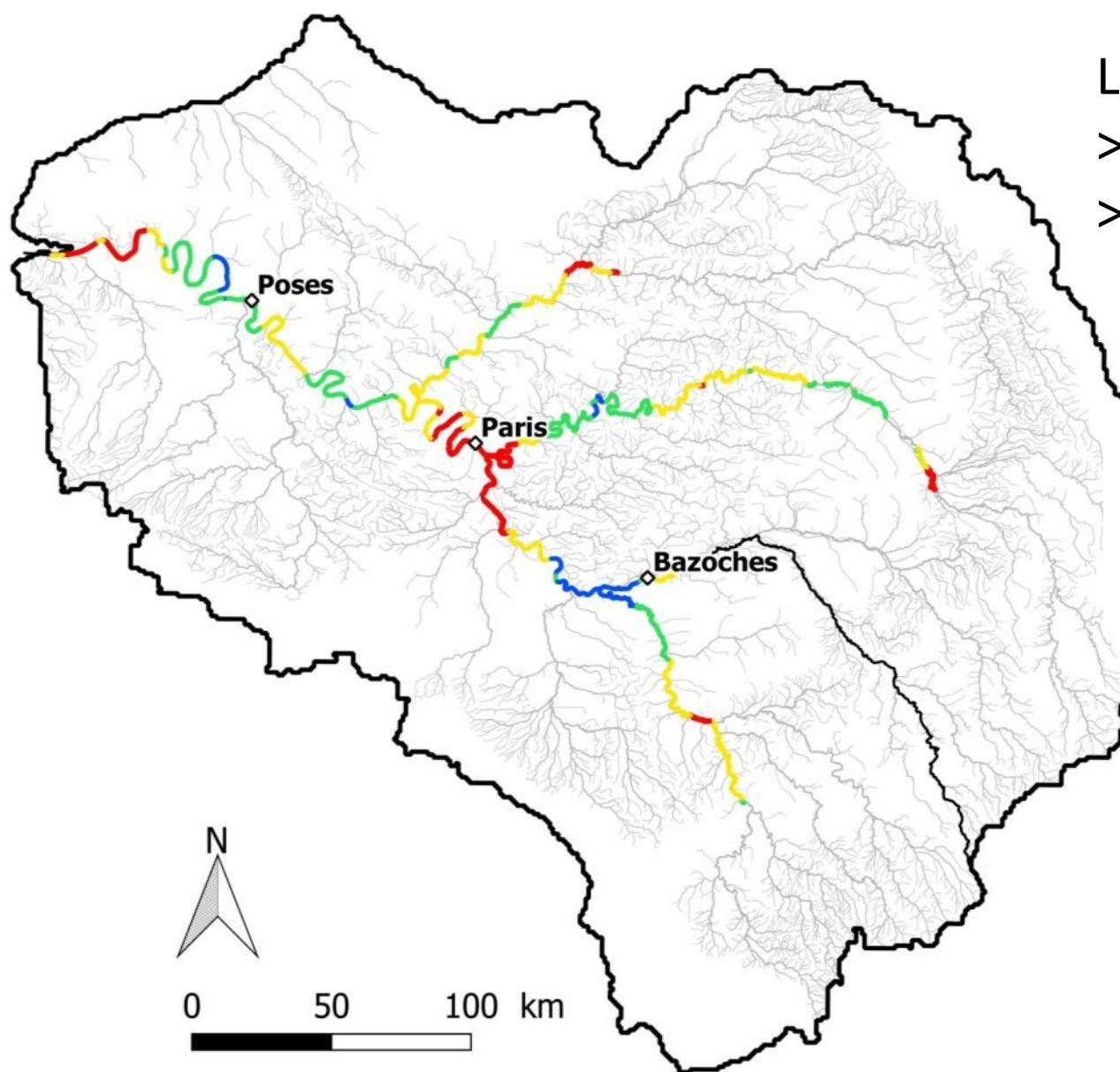


Geology

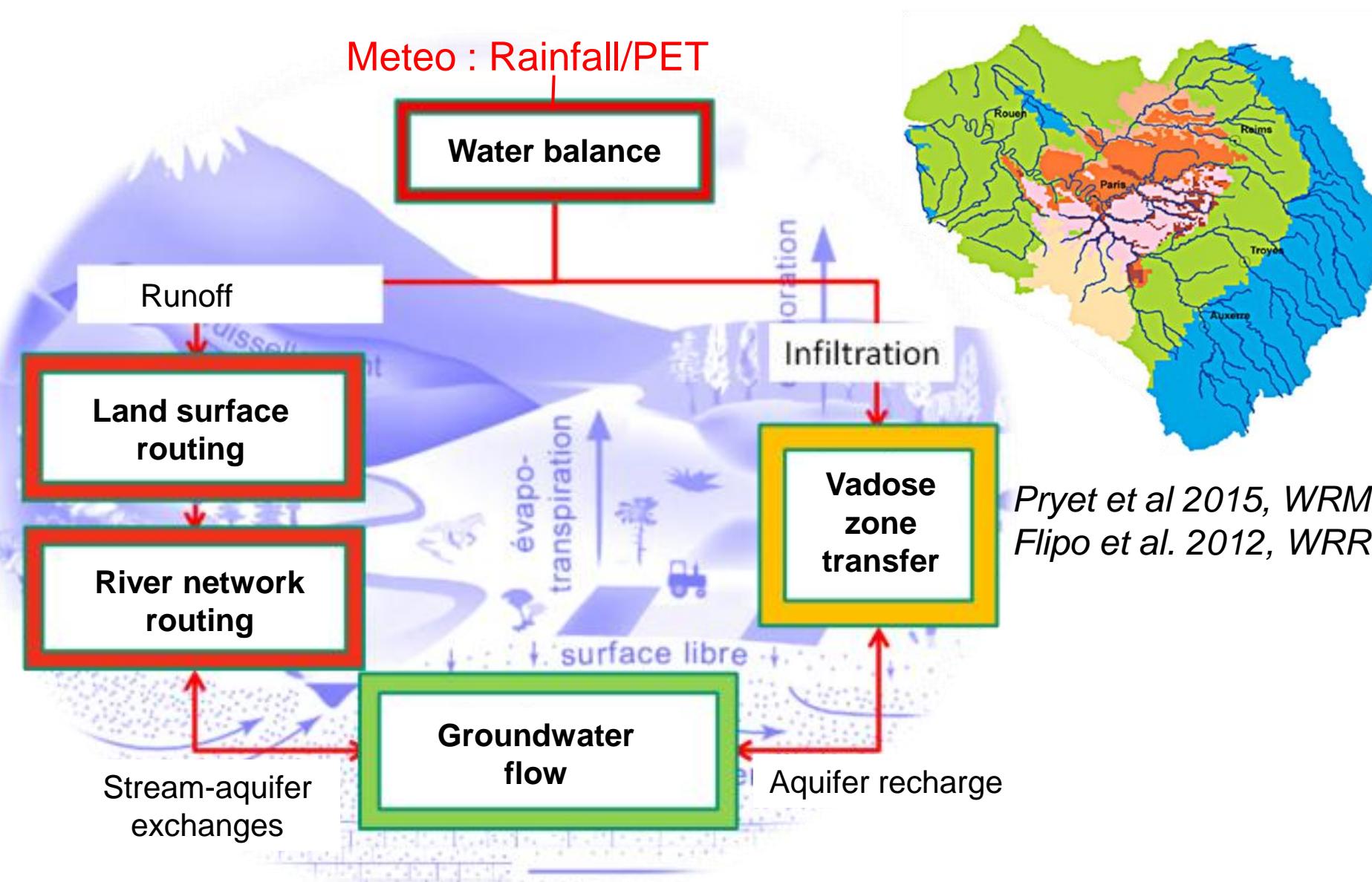
- Rupelian limestones
 - Priabonian "Brie" limestones and "Fontainebleau" sand
 - Upper-Eocene "Champigny" limestones and "Beauchamp" sands
 - Lutecian limestones and Ypresian sands
 - Thanetian limestones
 - Upper-Cretaceous chalk
 - Lower-Cretaceous and Jurassic marl and clayrock
- Simulated hydrographic network
- Main cities

0 25 50 100 150 200 Km

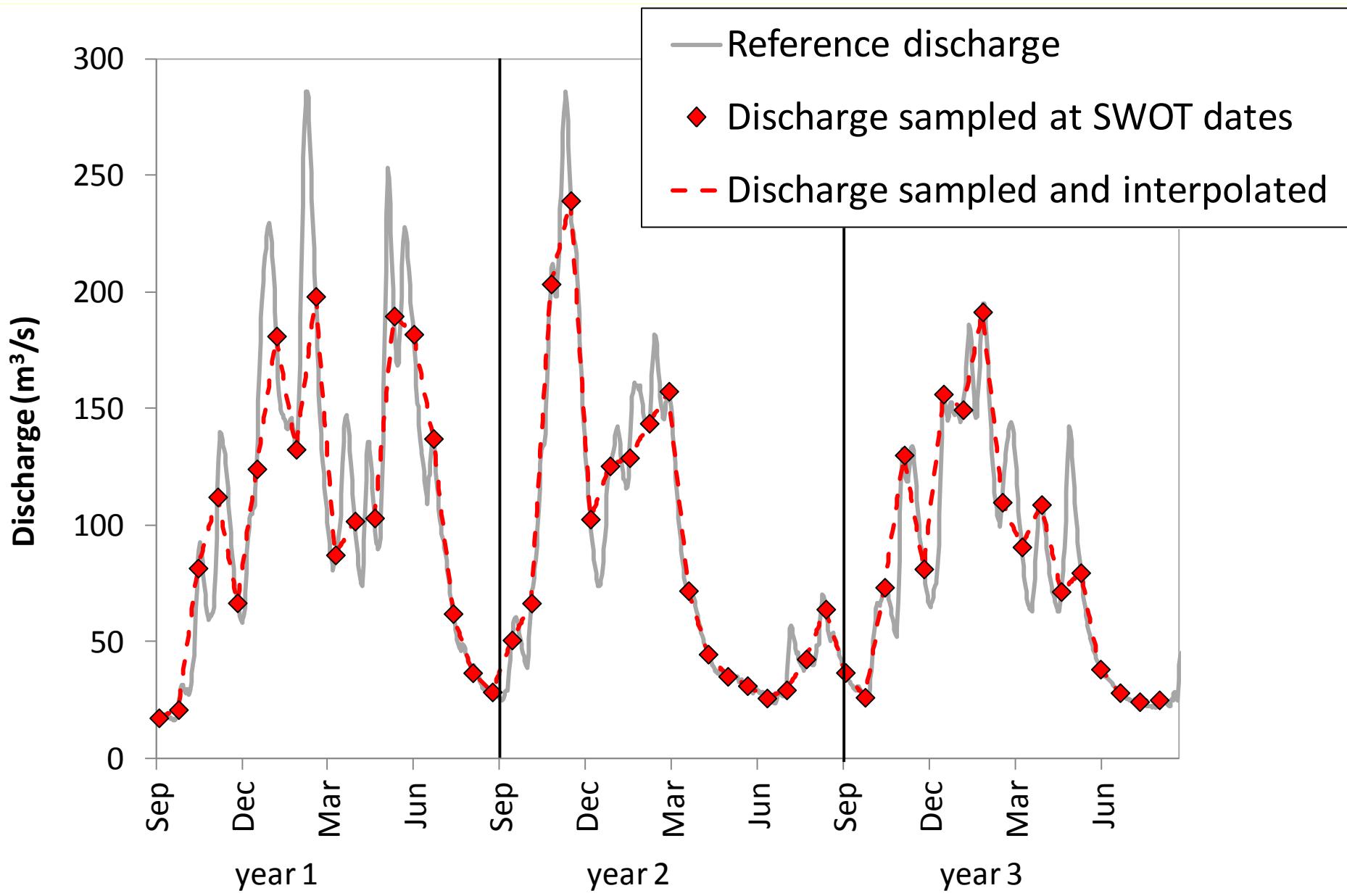
Observation by SWOT



Data simulation with the CaWaQS model



SWOT sampling effect

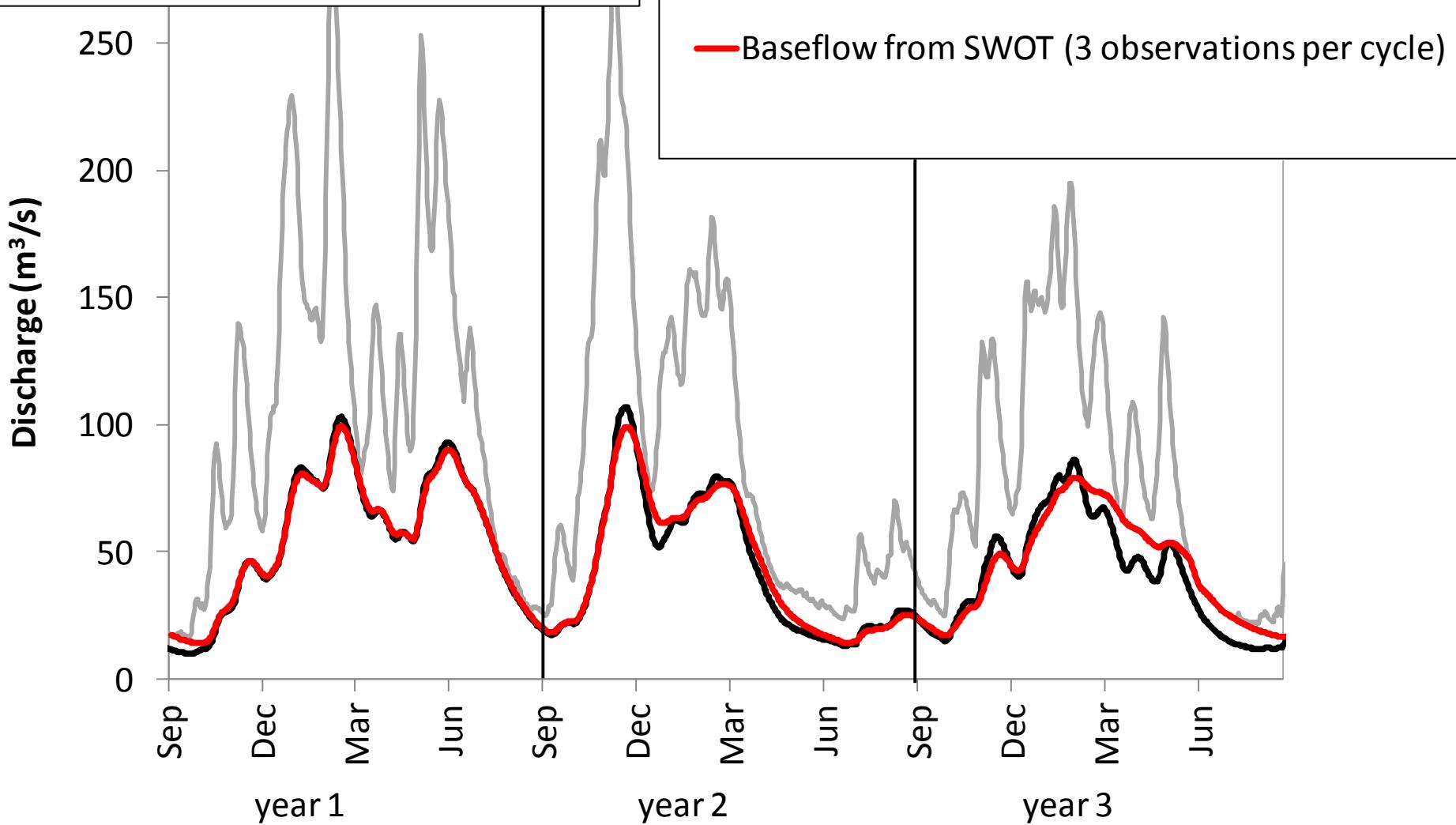


Baseflow estimate – SWOT sampling effect

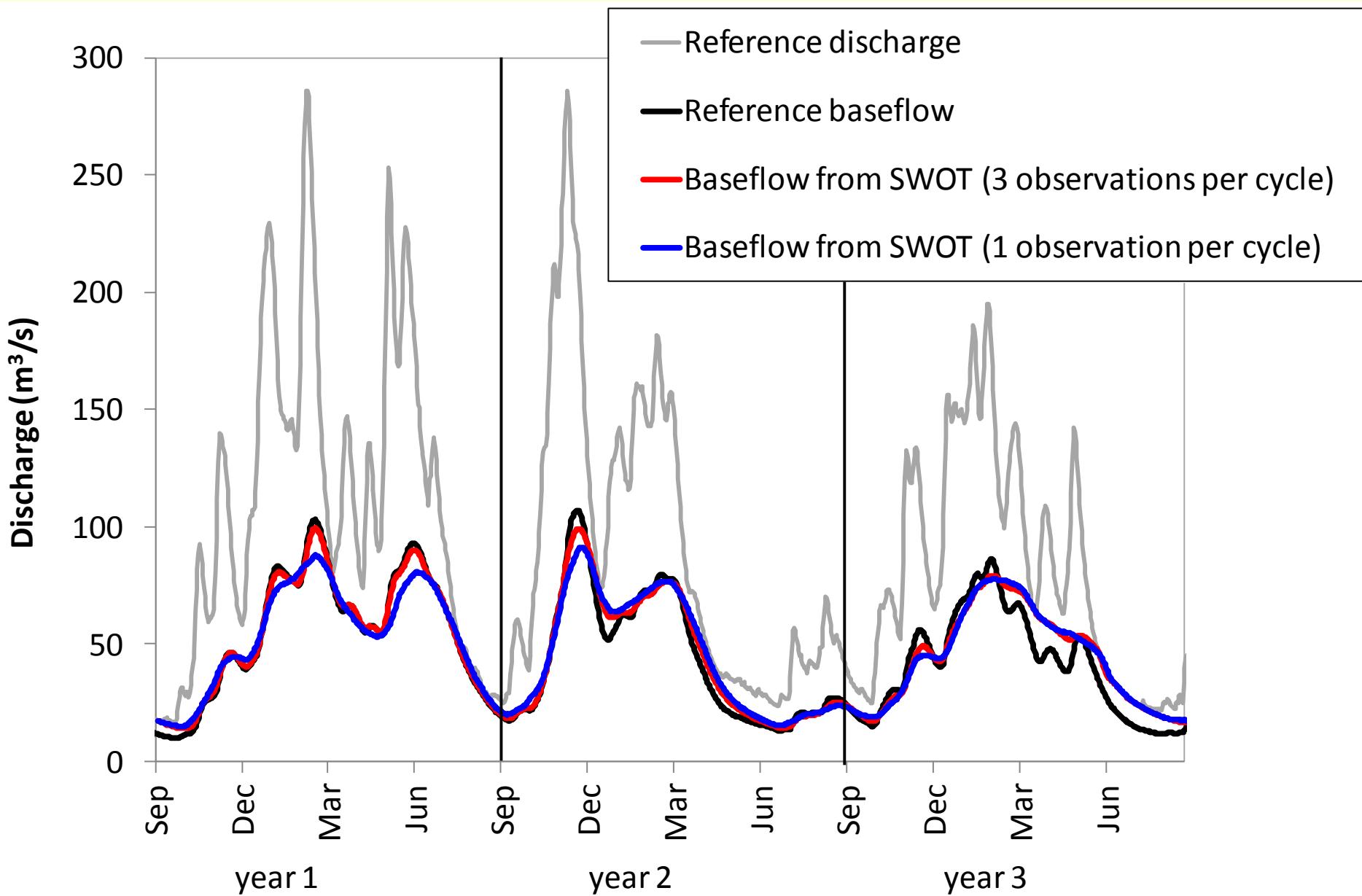
Relative bias
Nash-Sutcliffe

3.5 %
0.95

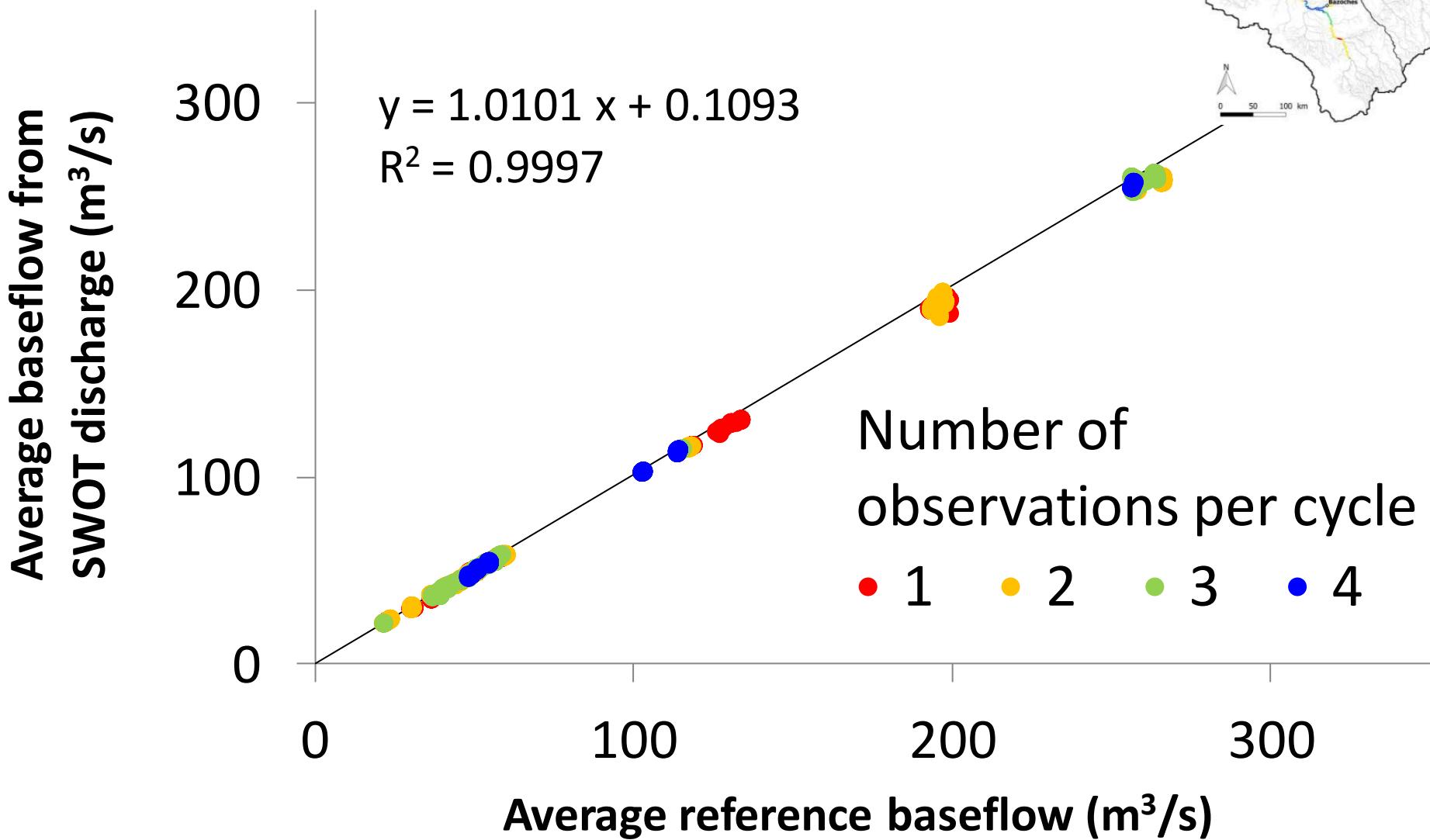
- Reference discharge
- Reference baseflow
- Baseflow from SWOT (3 observations per cycle)



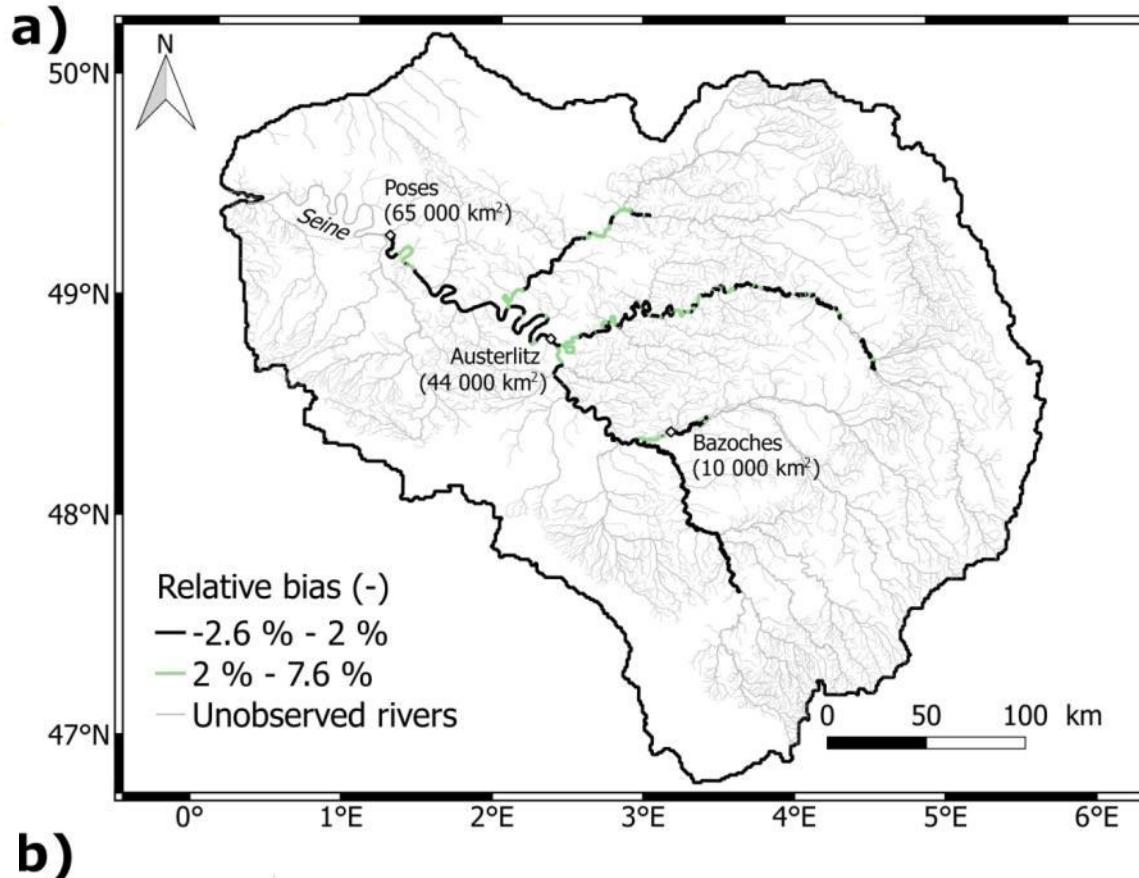
Baseflow estimate – SWOT sampling effect



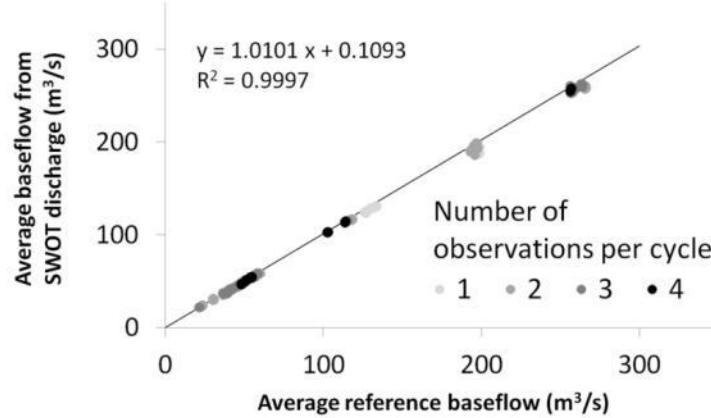
Retrieving river baseflow at the basin scale



- Very good estimates of the average baseflow, even with only 1 observation per cycle

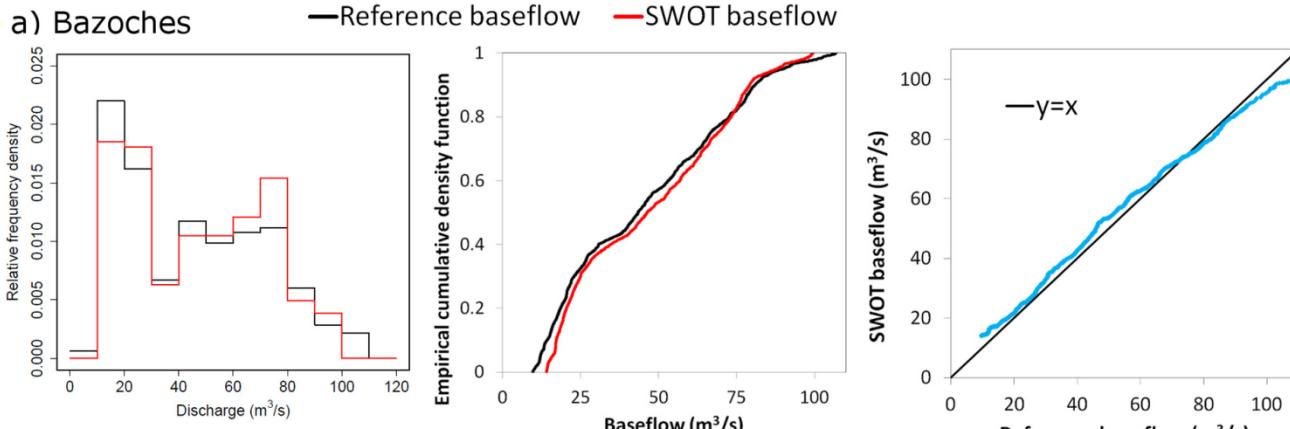


b)

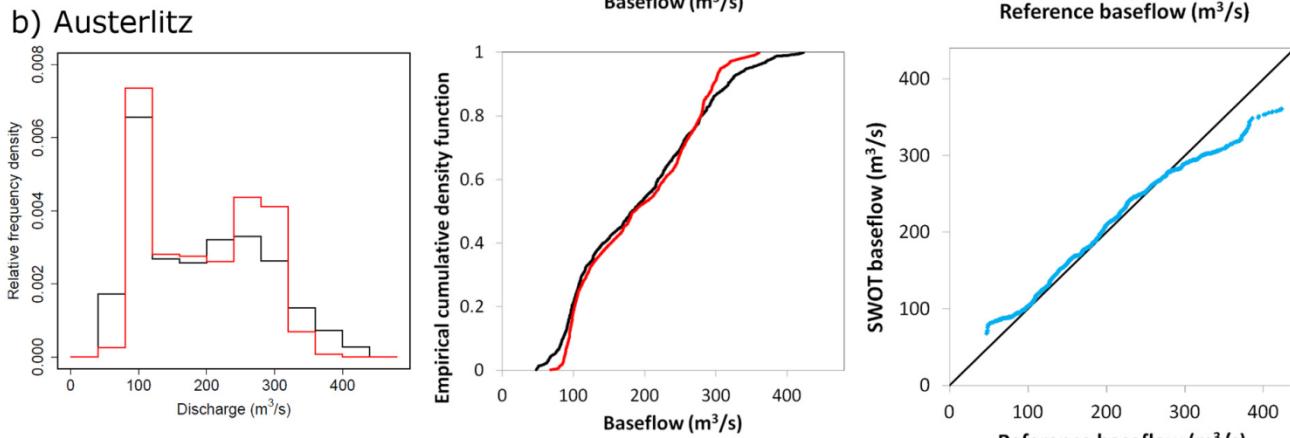


From upstream to downstream

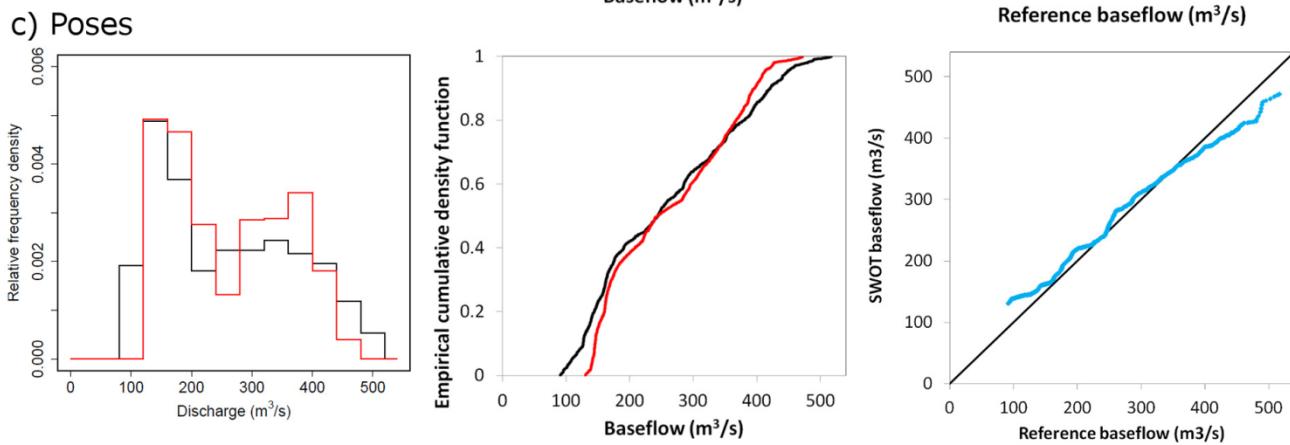
10 000 km²



45 000 km²



65 000 km²



Discharge error propagation

- Cumulates error on discharge estimate and SWOT sampling
- Lognormal error Y (Hagemann, 2017) on discharge estimate Q^* for the Seine from Durand et al., 2016

$$Q^*(t) = y(t)Q(t)$$

$$RR = \frac{Q^* - Q}{Q} \longrightarrow$$

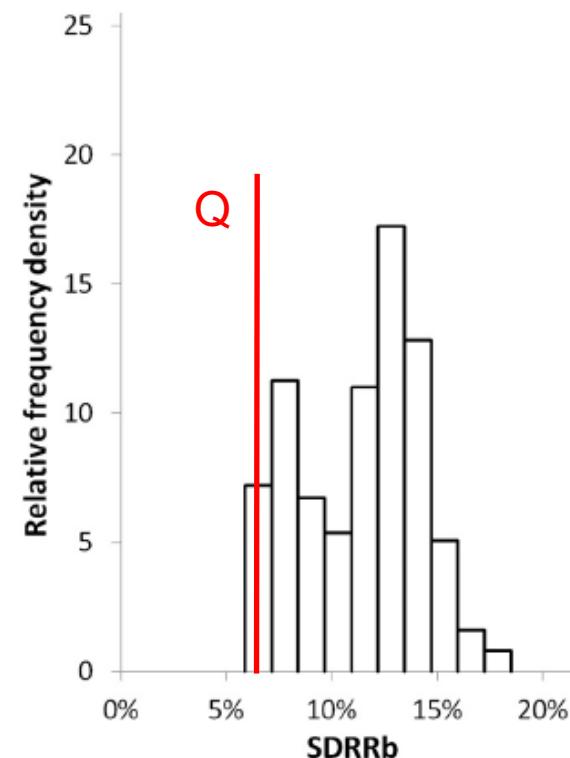
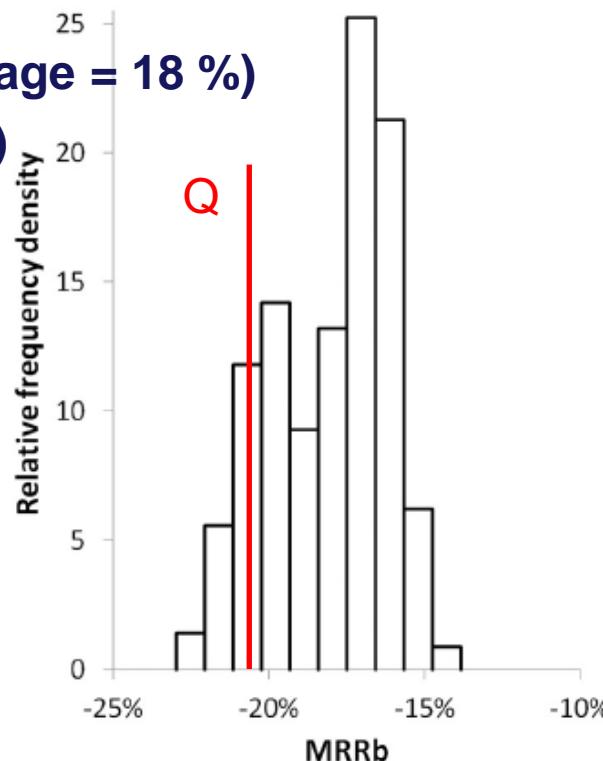
$$y(t) = 1 + RR(t)$$

- Errors on discharge :

- MRR = 21,2%, SDRR = 6.6%

- MRRb is dampened (average = 18 %)

- SDRRb is larger (av=11%)



Baratelli et al., 2018

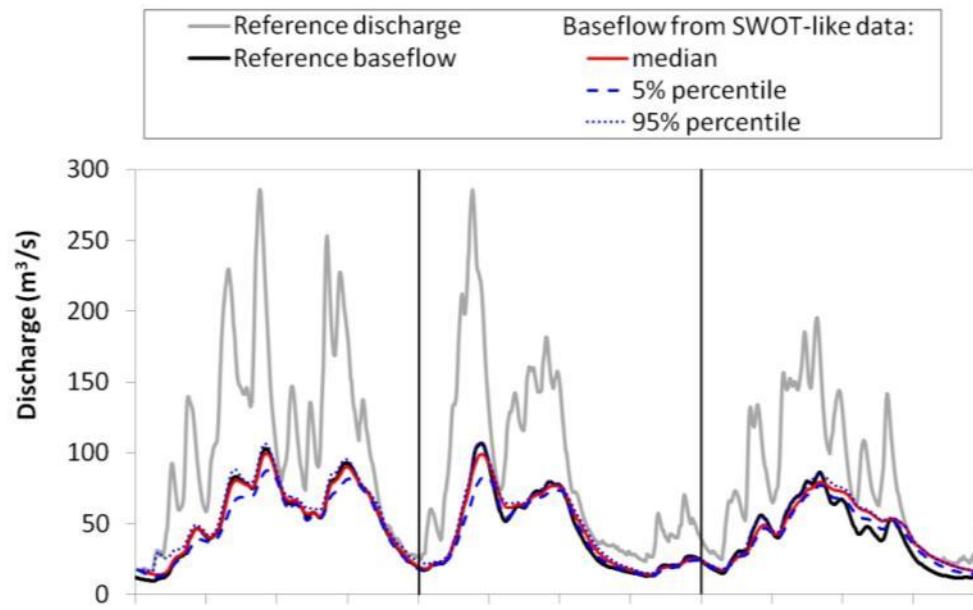


Retrieving river baseflow from SWOT spaceborne mission

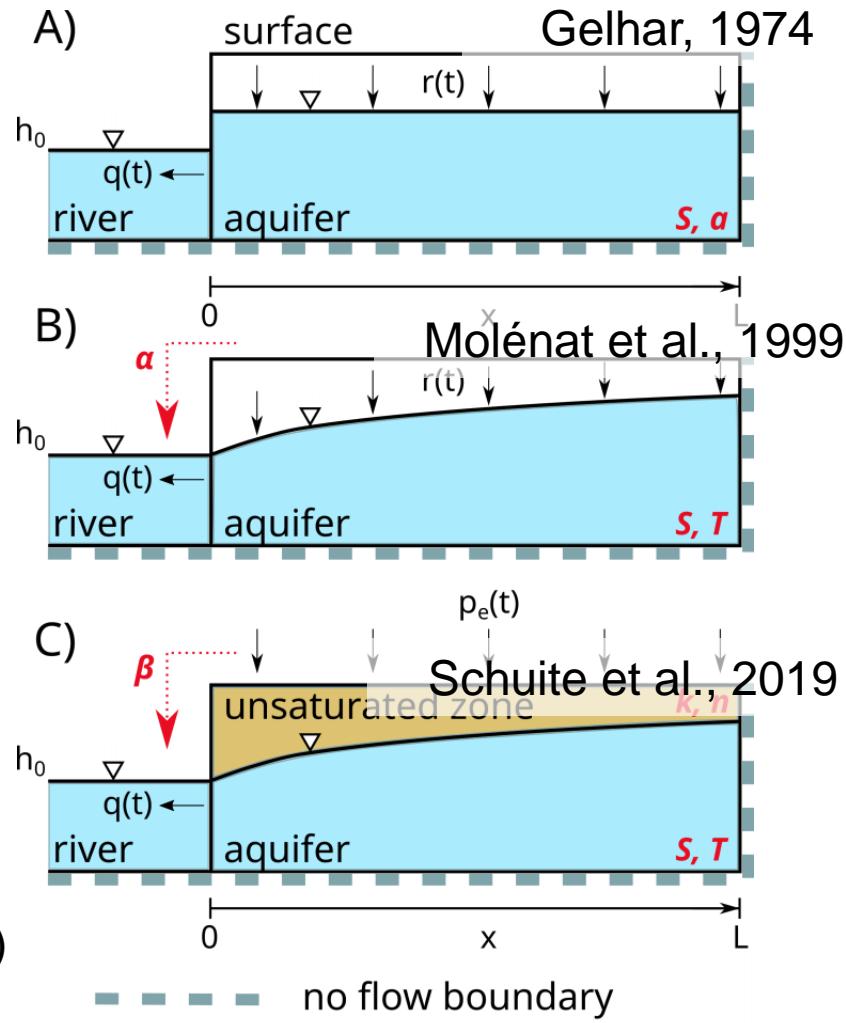
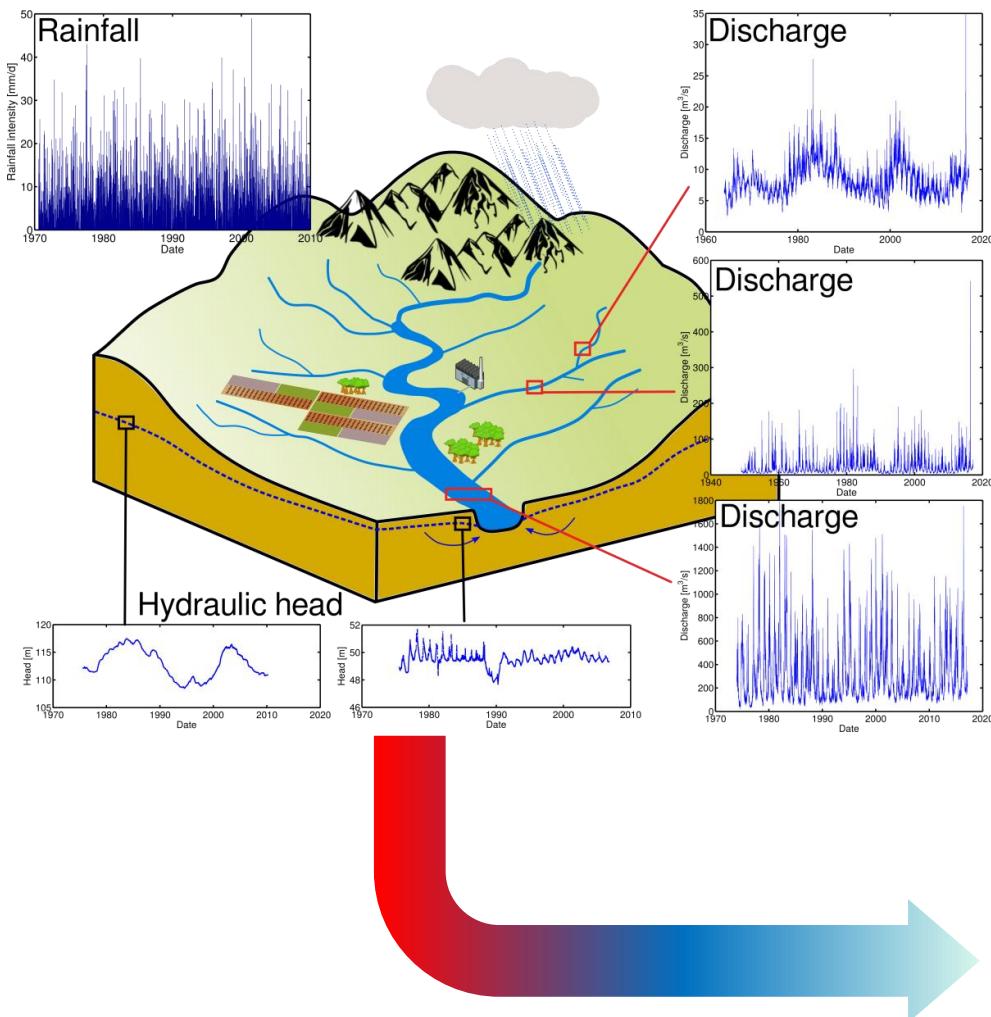
Fulvia Baratelli^{a,*}, Nicolas Flipo^{a,*}, Agnès Rivière^a, Sylvain Biancamaria^b

- **SWOT** spaceborne mission will provide uncertain **river discharge at global scale**
- We **estimate baseflow** applying a filter to SWOT-like river discharge
- Baseflow is retrieved from SWOT-like data in the Seine river basin **with good accuracy**
- **Uncertainties** on baseflow estimates are always slightly **lower than those on discharge**
- SWOT will potentially provide baseflow estimates with unprecedented **global coverage**

**Remote Sensing of Environment 218
(2018) 44–54**

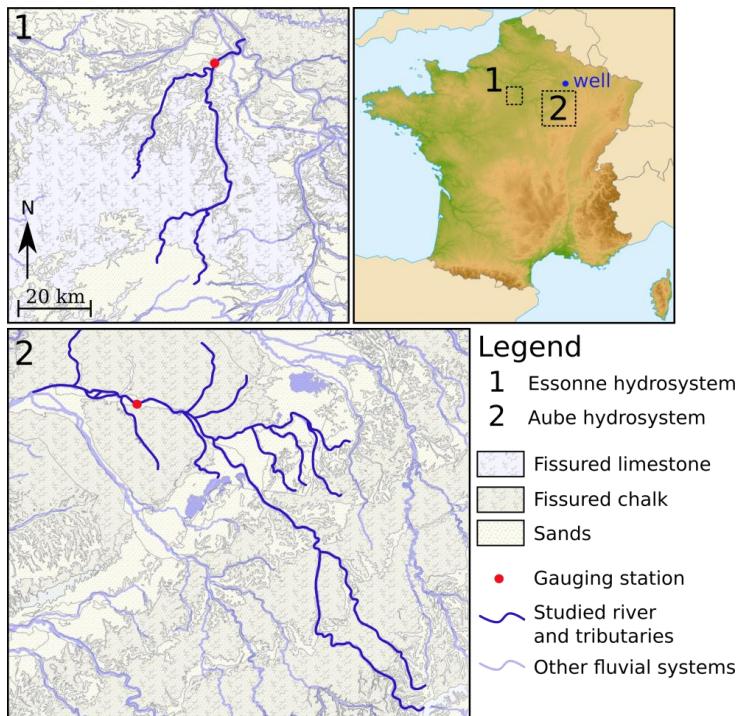


Hydrological Minimalist Transfer function HYMIT

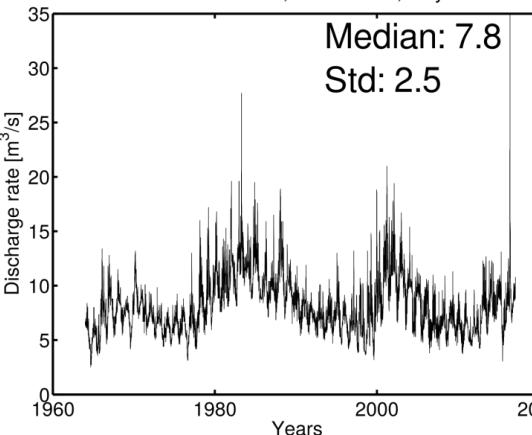


Hydrological Minimalist Transfer function HYMIT

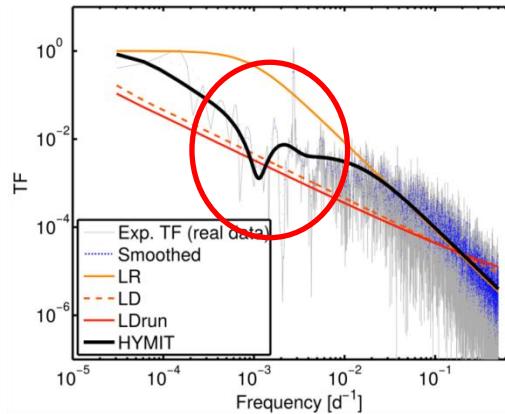
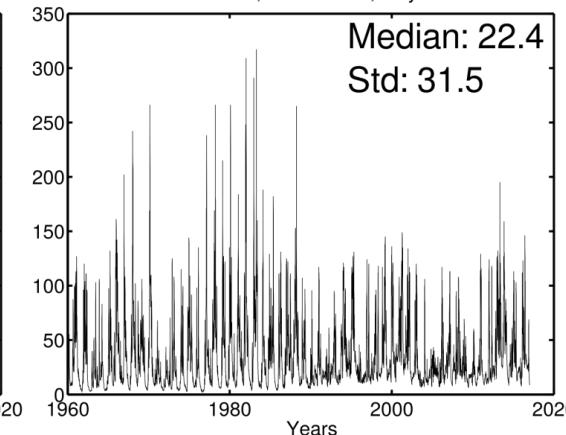
- Retrieve parameters for the interpretation of discharge data (Schuite et al., 2019, WRR)



Essonne: ~1900 km²

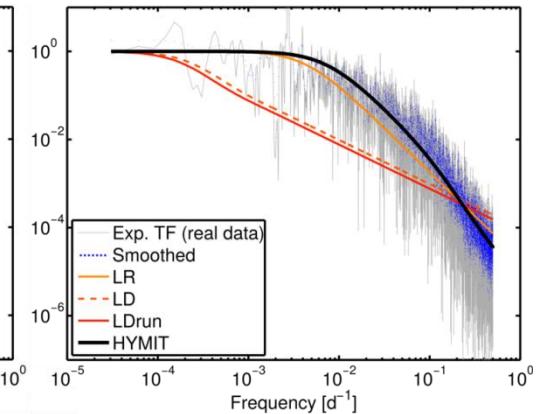


Aube: ~3600 km²



•Very low hydraulic diffusivity
•Thin vadose zone
•Low fraction of overflow (7%)

Thick vadose zone



•Very high hydraulic diffusivity
•Thin vadose zone
•High fraction of overflow (24%), very diffuse

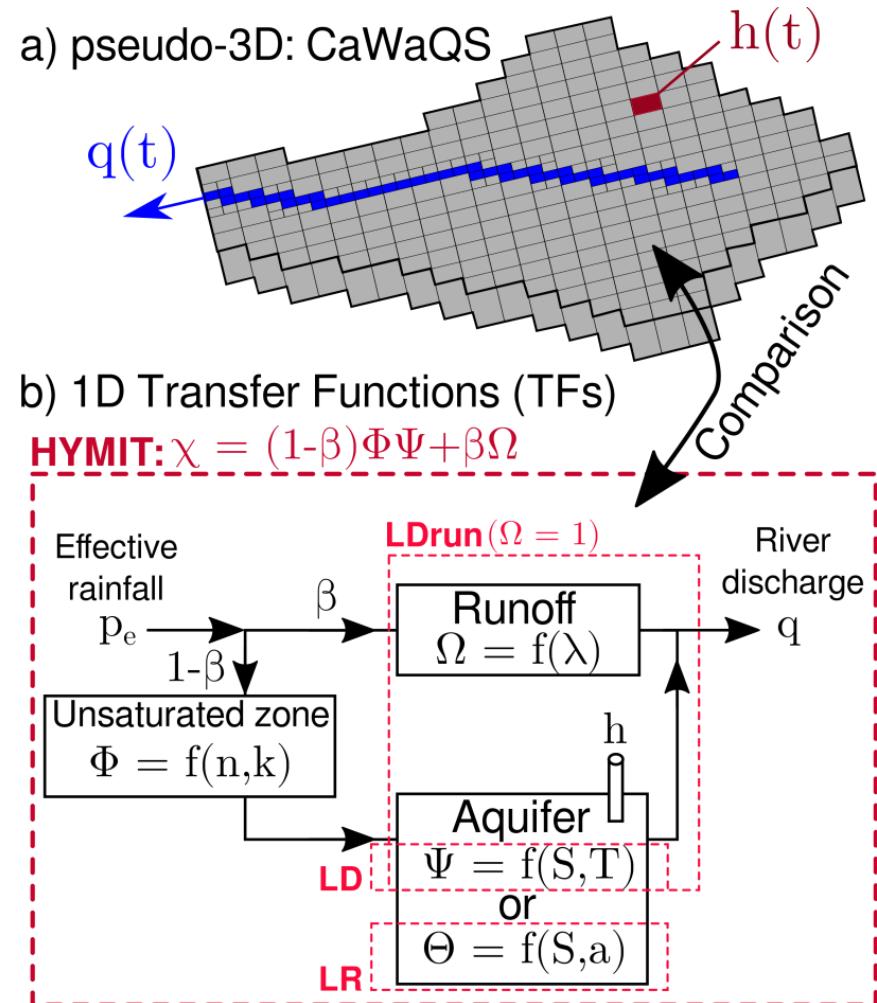
Improving the Spectral Analysis of Hydrological Signals to Efficiently Constrain Watershed Properties

Key Points:

- Hydrological responses to climatic forcing can be described more

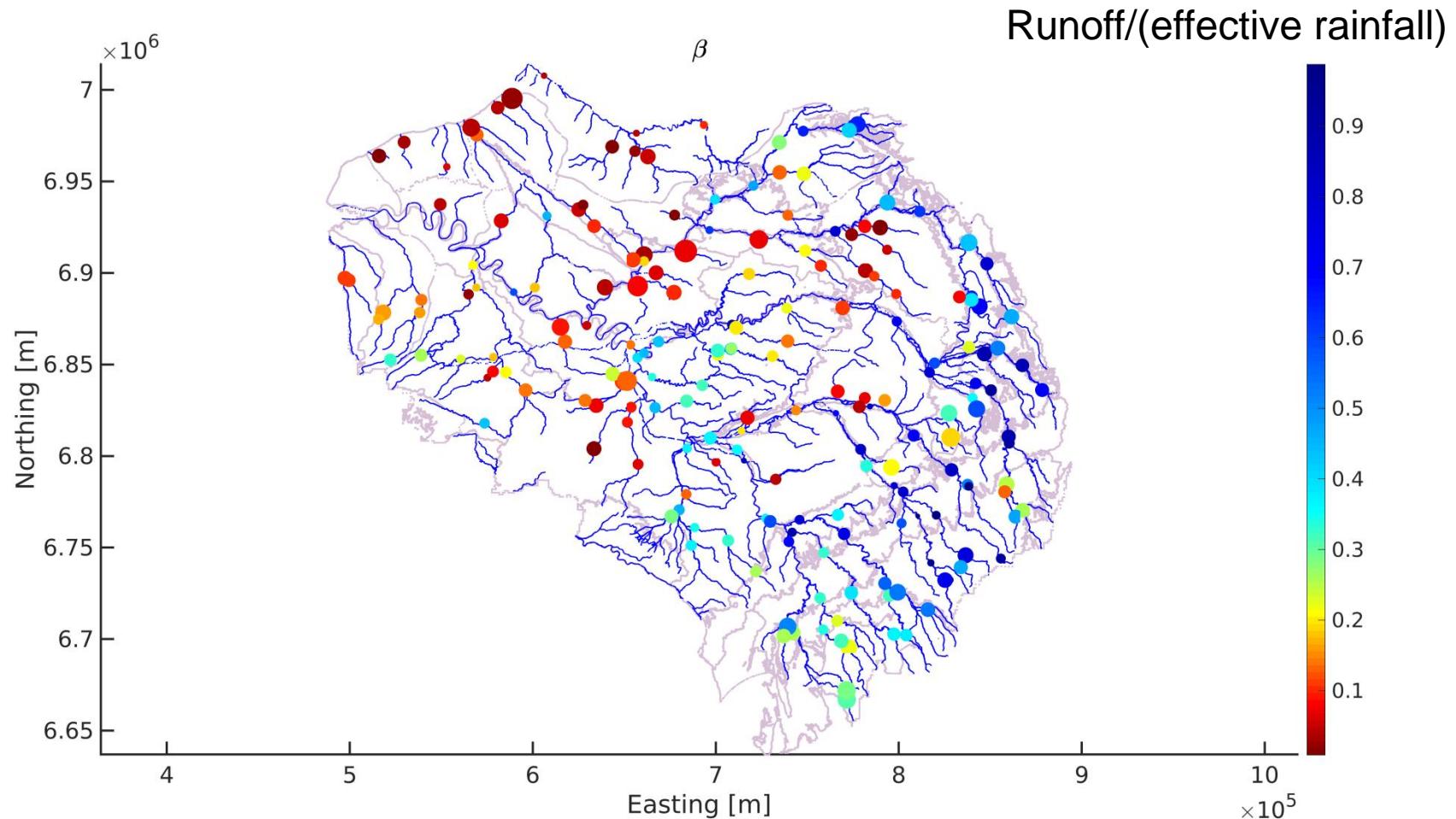
J. Schuite¹ , N. Flipo¹ , N. Massei² , A. Rivière¹ , and F. Baratelli¹ 

- HYMIT : HYdrological MInist Transfer function**
- Fast Estimate of the general physical properties and behavior of a catchment from commonly used data in hydrology: Q, Effective Rainfall**
- Subsurface fluxes are vertical in the unsaturated zone, horizontal in aquifers → **Adapt LSM structure ?**



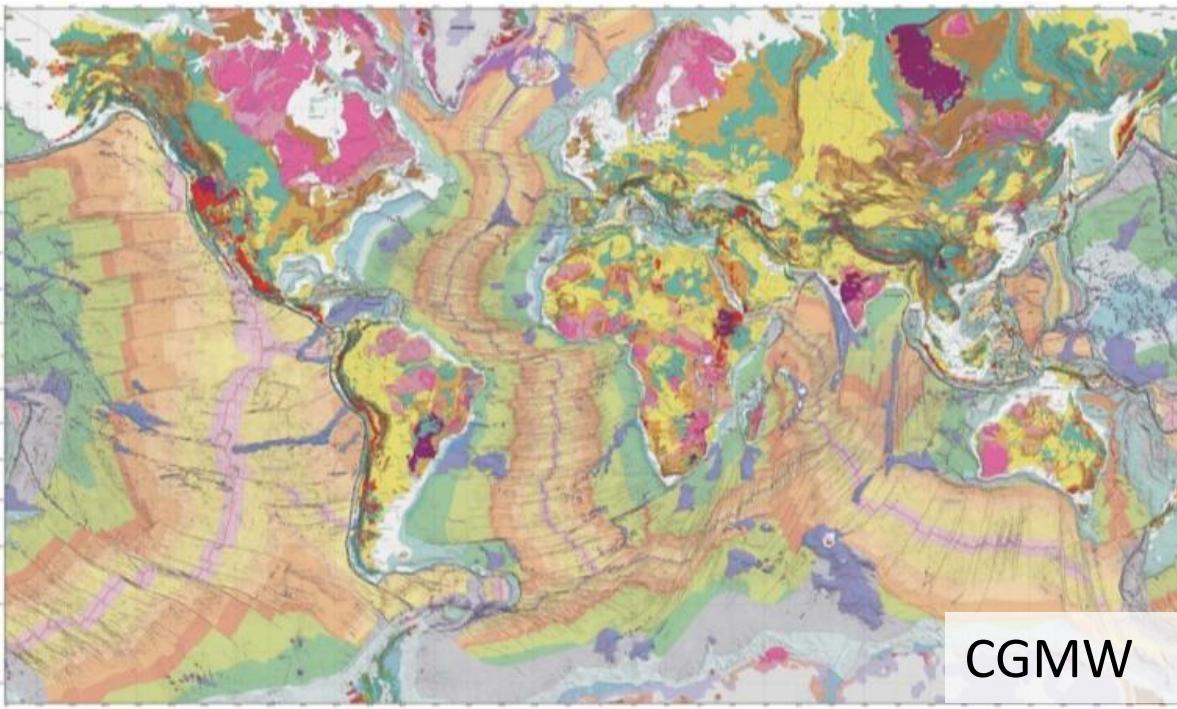
Runoff/infiltration partitioning on the Seine basin

- Calibrated conceptual surface model of CaWaQS (Flipo et al., 2012, WRR; Pryet et al, 2015, WRM; Baratelli et al., 2016 JH) + HYMIT



Perspectives

- Test more geological settings and larger basins



- Use TF approaches to guide calibration procedures of distributed hydrological model given a proper model structure
- Refine hydrograph separation methods based on spectral analysis

Thank you

