



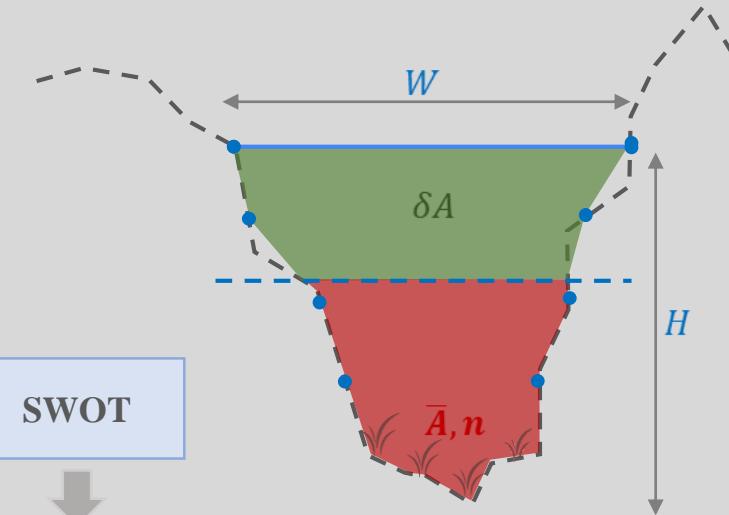
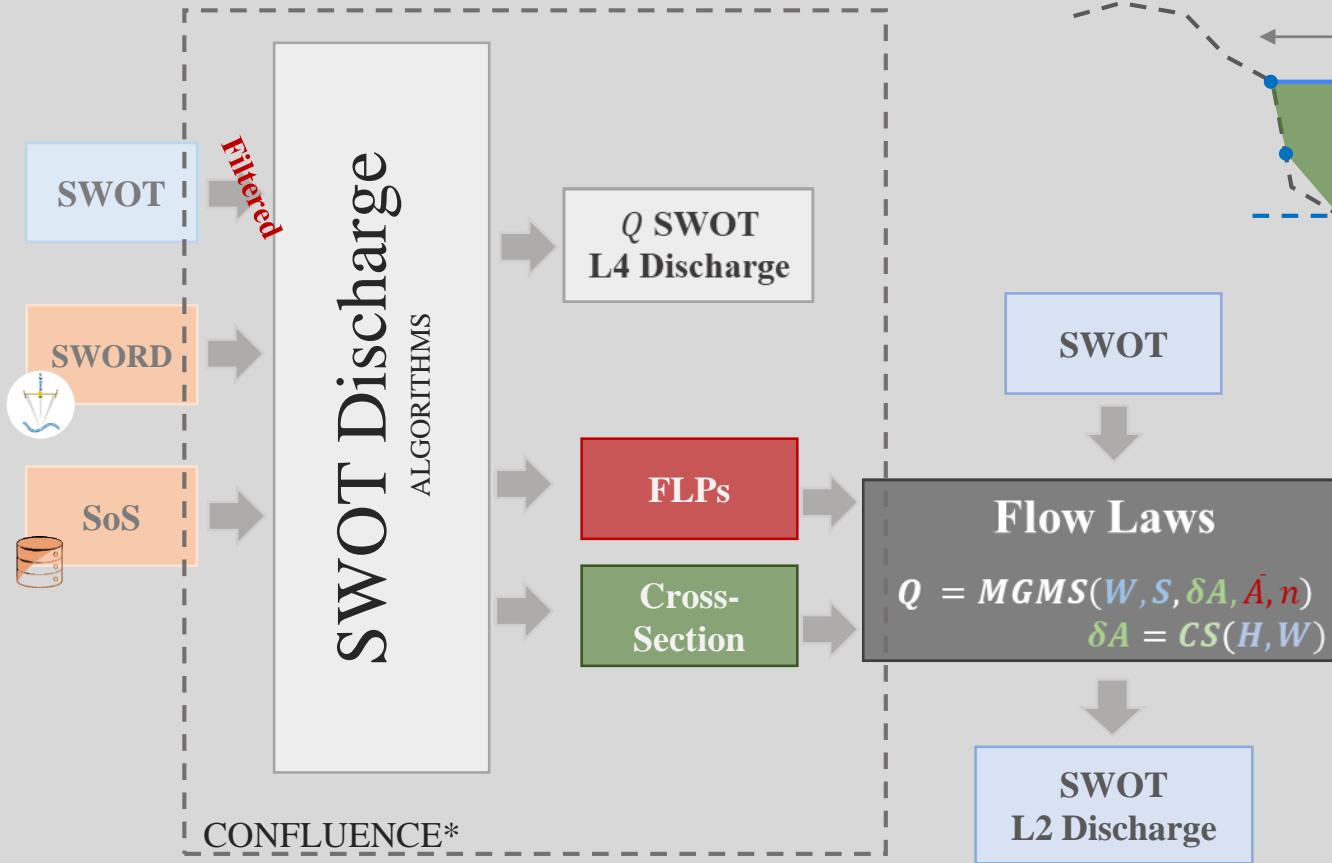
# SWOT River Bathymetry Products

**Hind Oubanas and Jeffrey Neal**

P.-O. Malaterre, M. Durand, P. Bates, S. Chuter, C. Gleason, I. Rezende

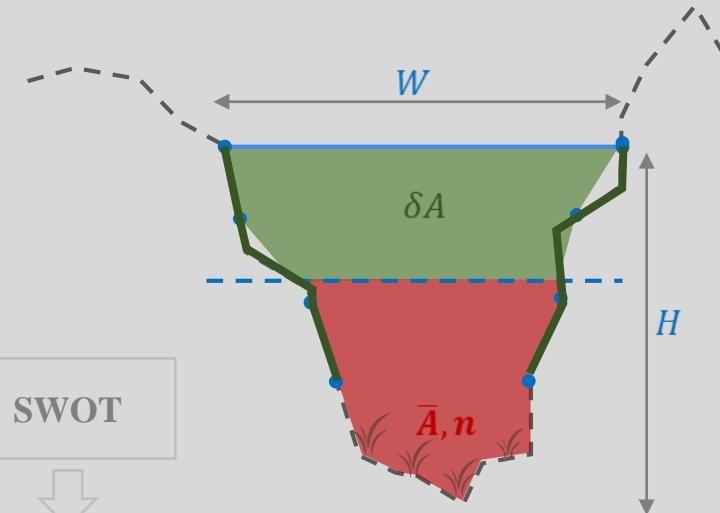
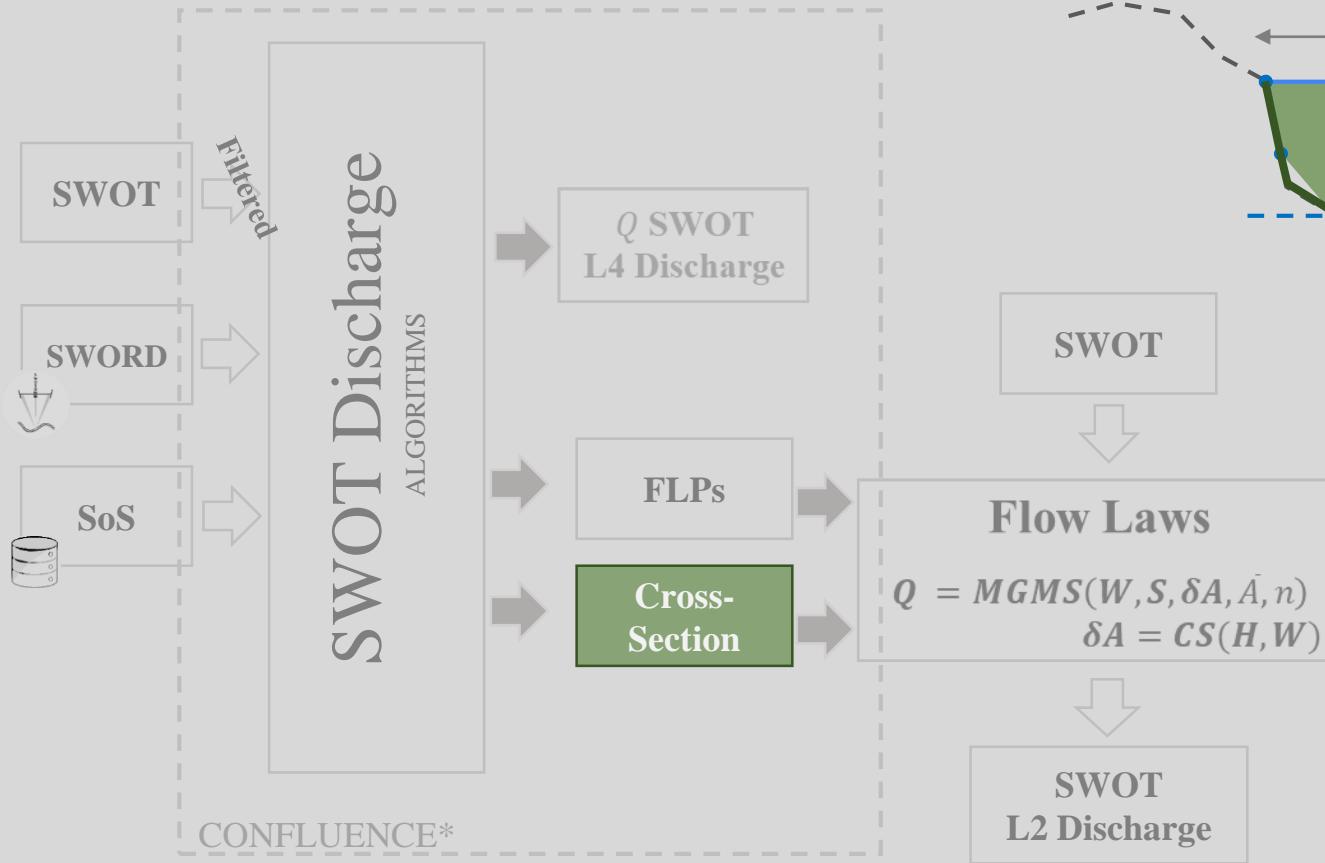
**SWOT**

# SWOT Discharge Estimation Framework



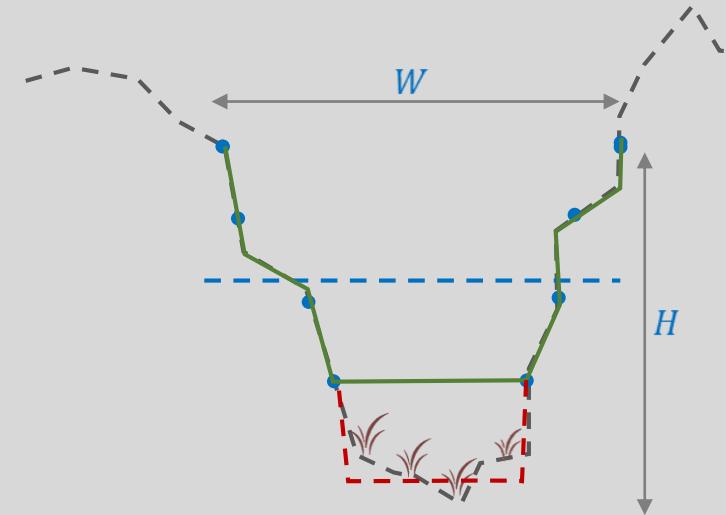
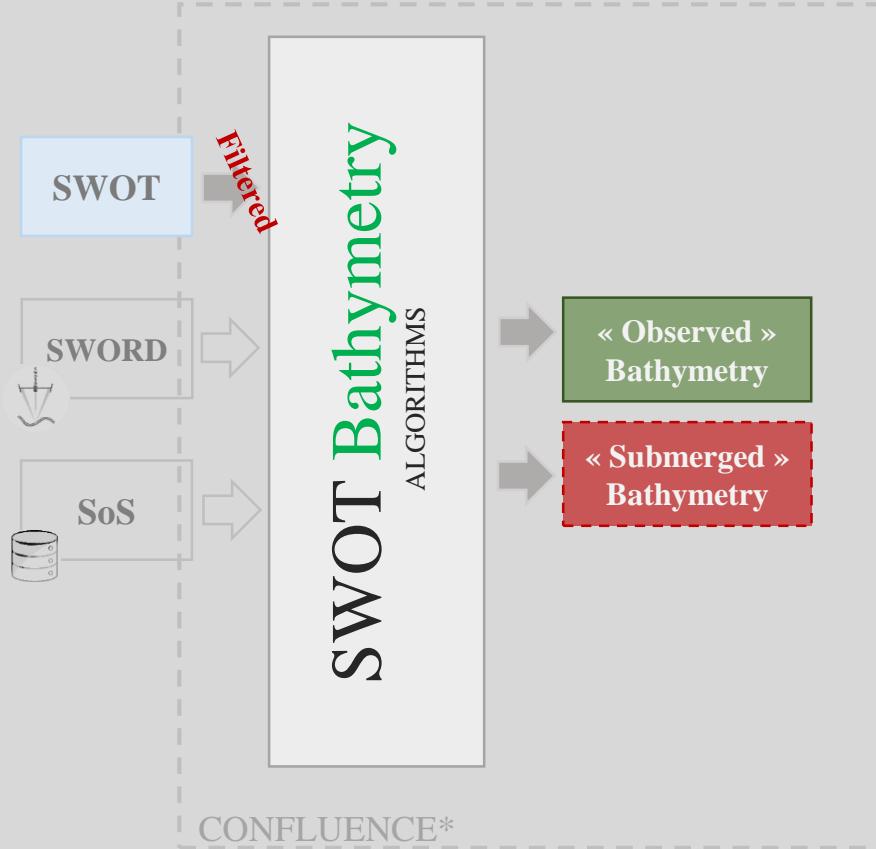
\* Add a few more layers

# SWOT Discharge Estimation Framework

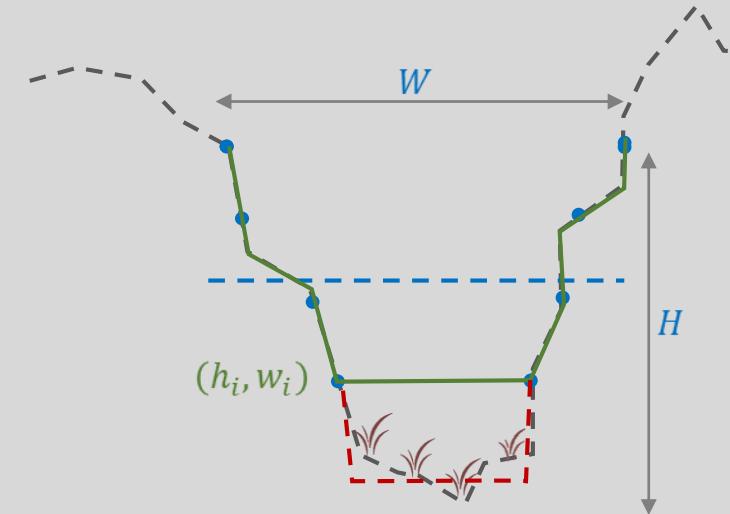
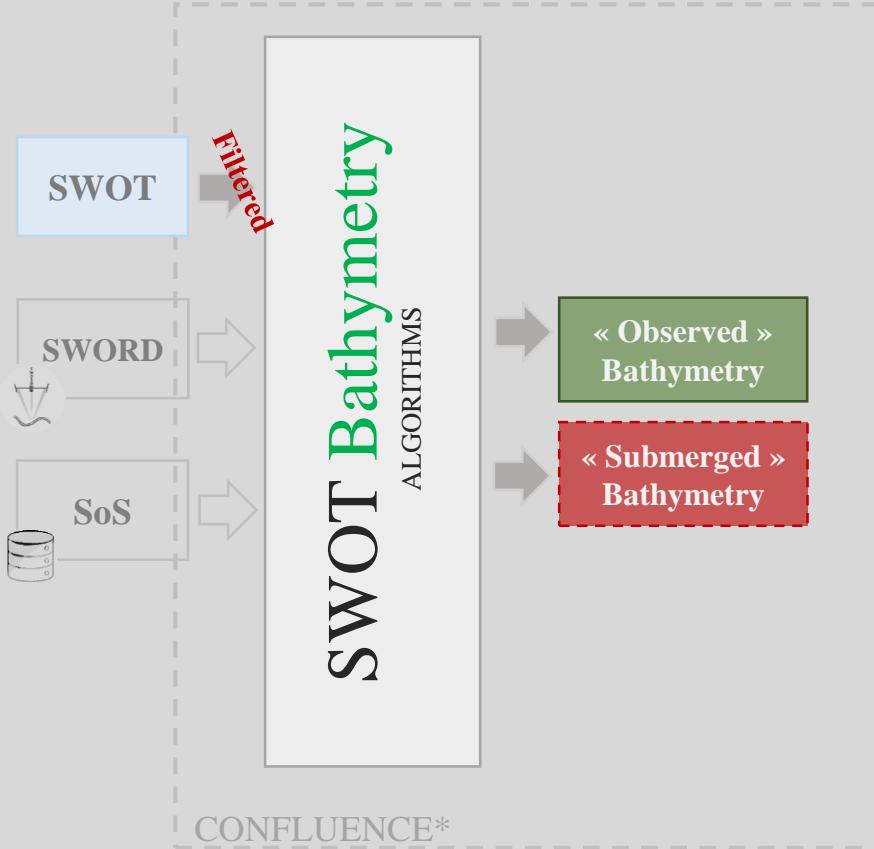


\* Add a few more layers

# SWOT Bathymetry Estimation Framework



# SWOT Bathymetry Estimation Framework



## Outputs/Format :

- Optimized nodes/reaches cross-section profils.
- Approximated Hypsometry curves.
- Cross-section variability with the reach.
- Other?

## Different approaches already exist ...

*We hope to get your feedback on the usefulness  
of such product, and recommendations on  
product format*

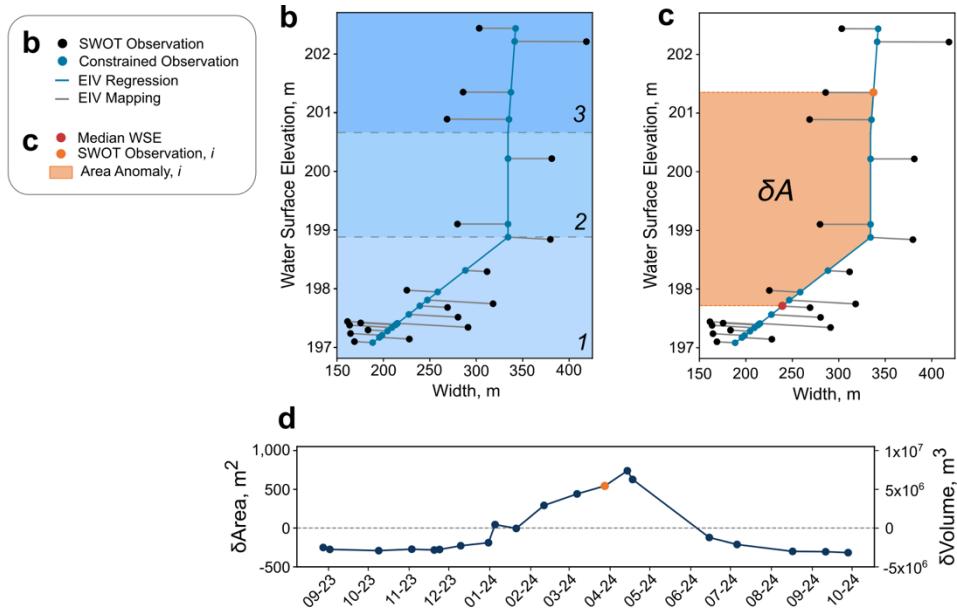


# Estimating reach-scale bathymetry

The observed bathymetry can be estimated by computing the WSE-width relationship (river hypsometry, Durand et al. 2024)

The hypsometric curve fits are already running in Confluence

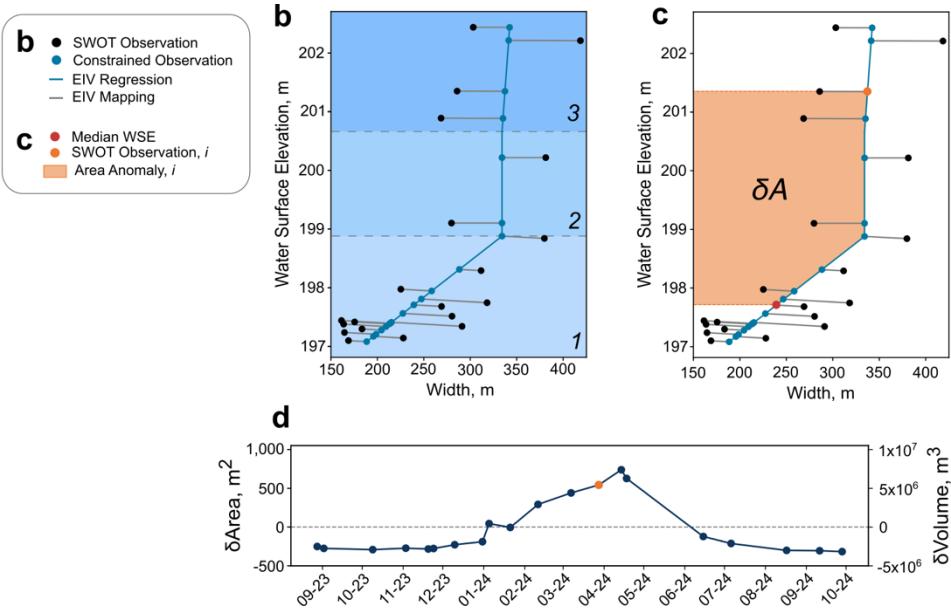
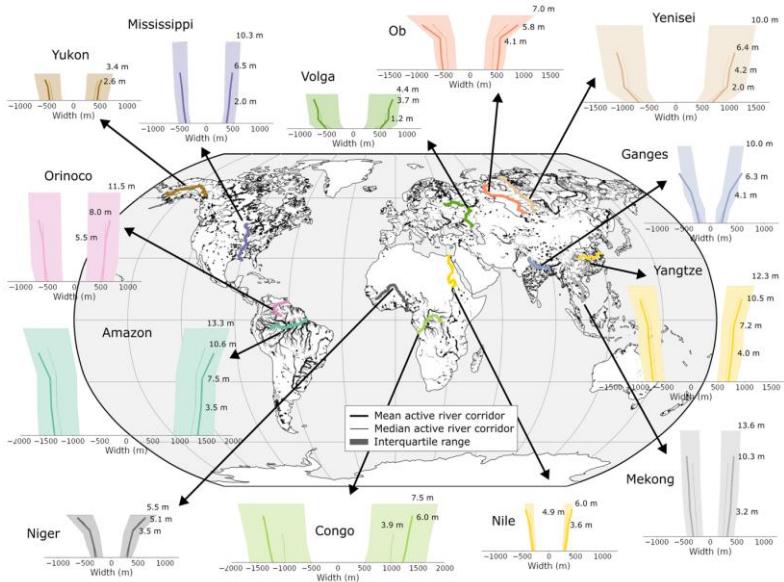
Caveat: they estimate the whole river width, so this would give you a symmetrical bathymetry on each side





THE OHIO STATE  
UNIVERSITY

# Estimating reach-scale bathymetry



Wade, J., Cerbelaud, A., David, C., Durand, M., Frasson, R., Pavelsky, T., Oubanas, H. "Wide-Swath Altimetry Maps Bank Shapes and Storage Changes in Global Rivers". In review.

Durand, M., Dai, C., Moortgat, J., Yadav, B., Frasson, R. P. de M., Li, Z., et al. (2024). Using river hypsometry to improve remote sensing of river discharge. *Remote Sensing of Environment*, 315, 114455. <https://doi.org/10.1016/j.rse.2024.114455>

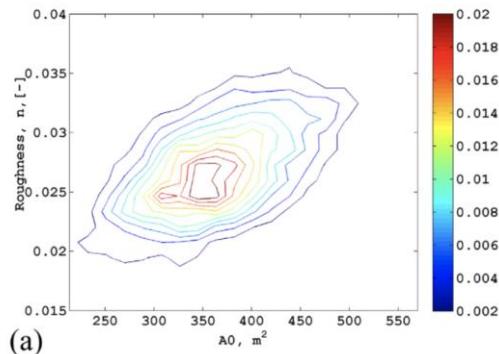


Pre-launch studies (e.g. Yoon et al. 2016) confirmed that at least at reach-scale, multiple values of resistance (Manning's  $n$ , Strickler's  $K$ , etc) and submerged bathymetry yield are equally like (a).

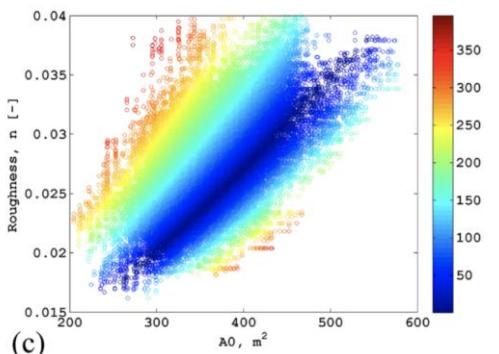
This does not affect discharge accuracy (b), and such a bathymetry product may still be valuable.

Equifinality may have less impact on node-estimates of bathymetry

## Caveat: Equifinality at reach-scale



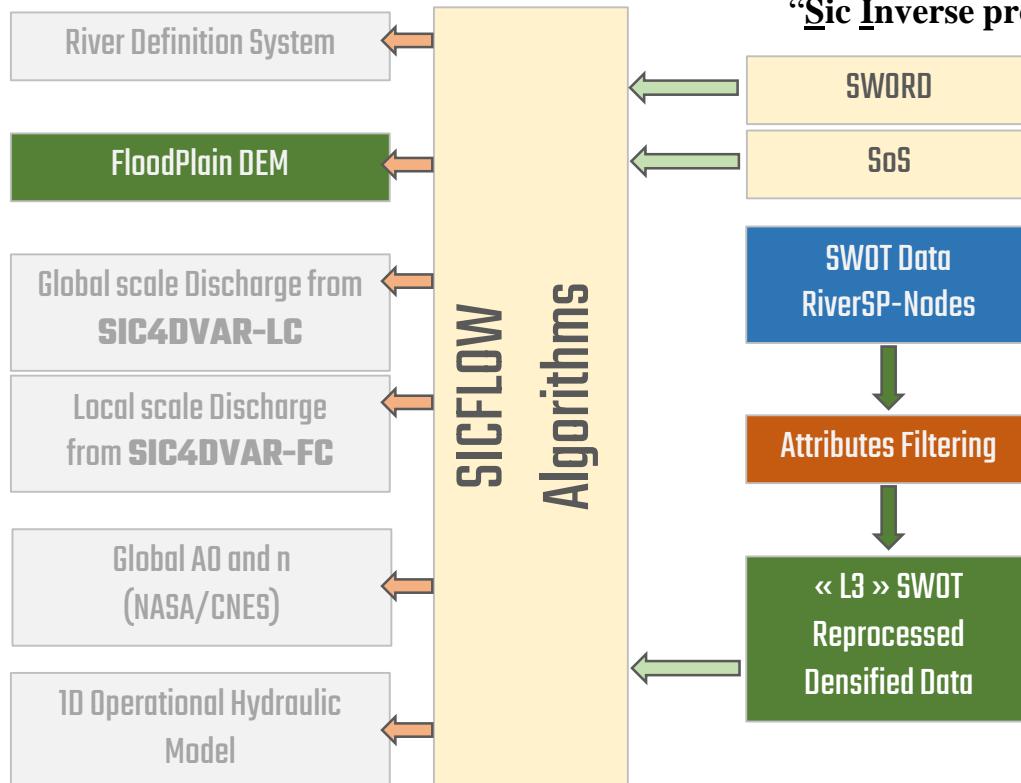
(a)



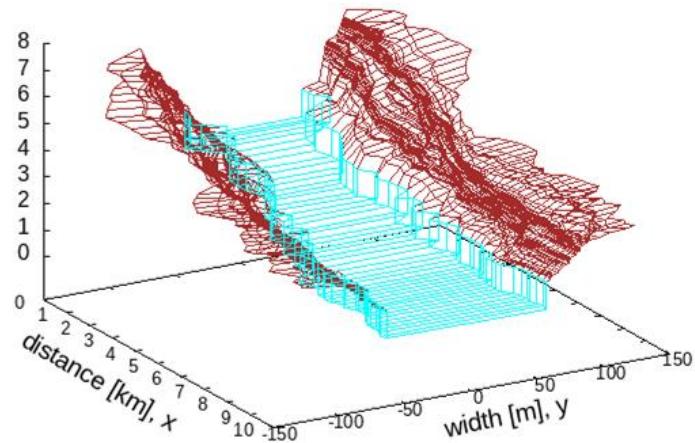
(c)

# SICFLOW

“Sic Inverse problem Capabilities for river FLOW dynamics”



Bathymetry: case br4ri3sr2160260125



Estimating node-scale bathymetry

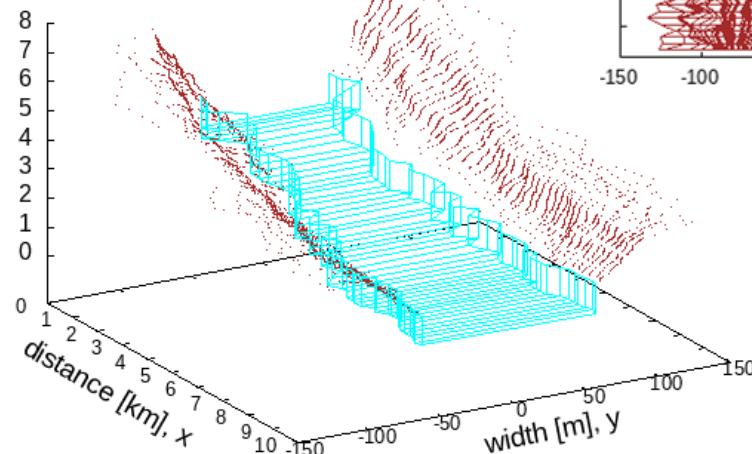
# SICFLOW

## Estimation Approach :

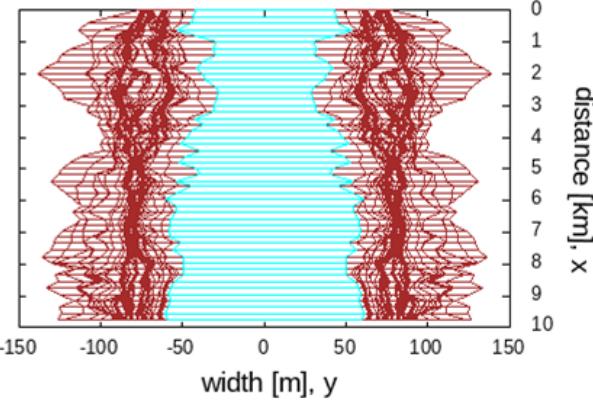
- **Observed Bathymetry** : Double sweep for smoothed parametric curve  $w(h)$  enforcing continuous increase condition.
- **Submerged Bathymetry** : **SIC4DVAR** Discharge Estimation (combined estimation of bed level and friction).

## Output Format :

- Node Level Product
- **Observed Bathymetry** : Cross-section approximation with pairs of  $(h_i, w_i)$  with optimized number of points.
- **Submerged Bathymetry** : Bed-level and min-width for simplified rectangular shape.



## “Sic Inverse problem Capabilities for river FLOW dynamics”



Estimating node-scale bathymetry

# SICFLOW

## Estimation Approach :

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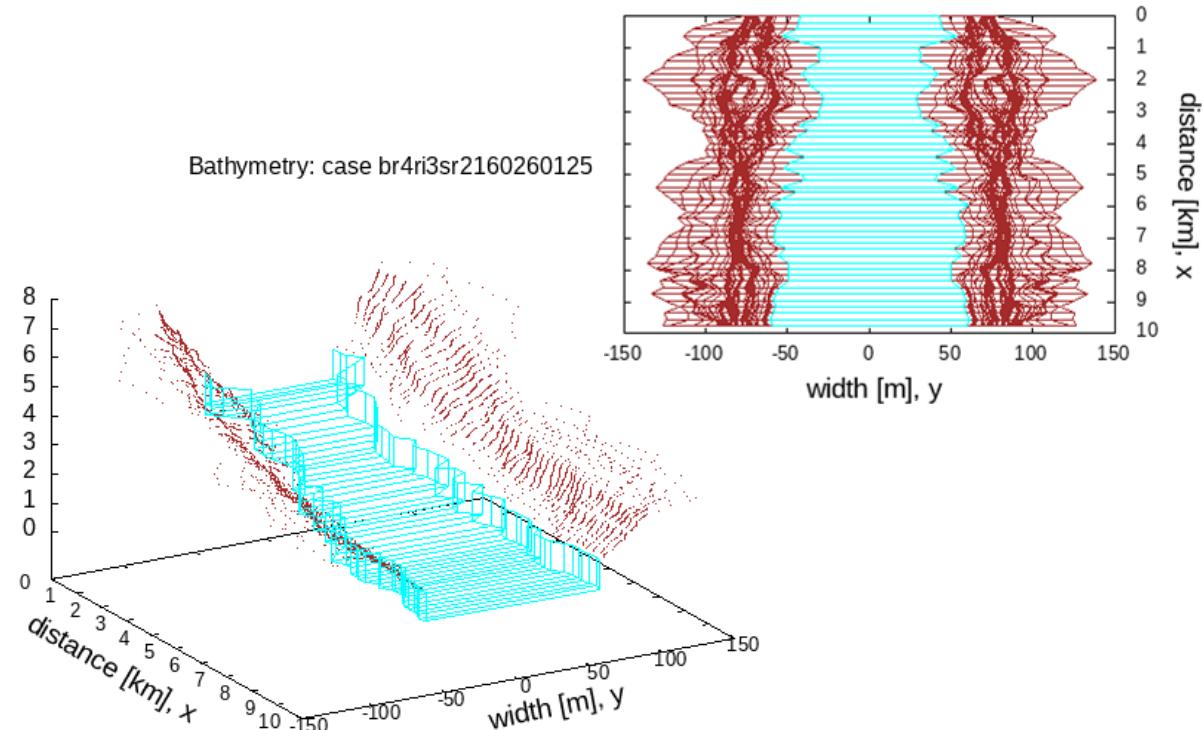
- Node Level Product
- **Observed Bathymetry** : Cross-section approximation with pairs of  $(h_i, w_i)$  with optimized number of points.
- **Submerged Bathymetry** : Bed-level and min-width for simplified rectangular shape.

## Availability :

Outputs **already available globally** in L4 SoS (SIC4DVAR's outputs)

**Validation** : Validation has been performed through discharge. Validation using Lidar/Optical data is On-going.

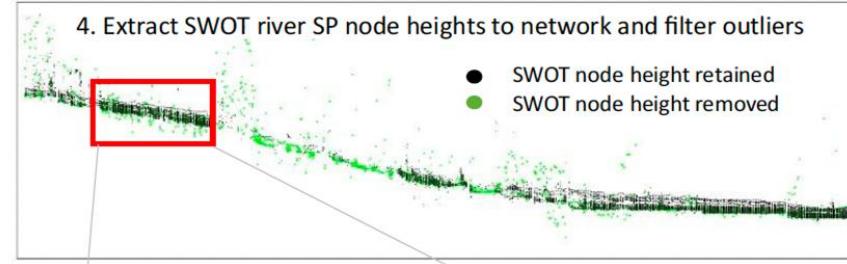
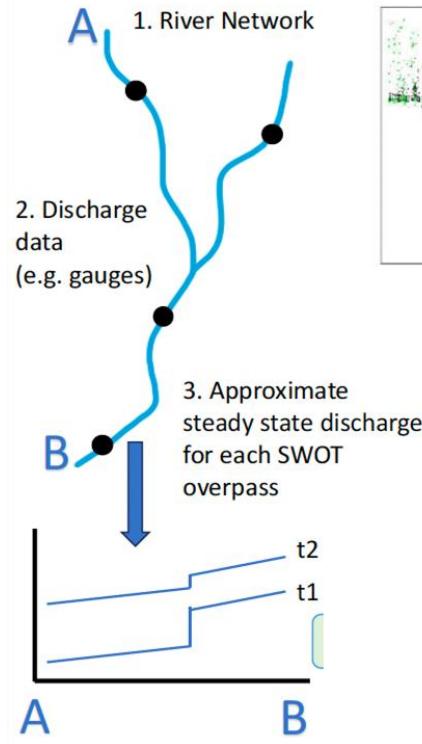
## “Sic Inverse problem Capabilities for river FLOW dynamics”



Estimating node-scale bathymetry



Inversion of 1D Gradually Varied Flow equations to solve for bed given a set of SWOT overpasses

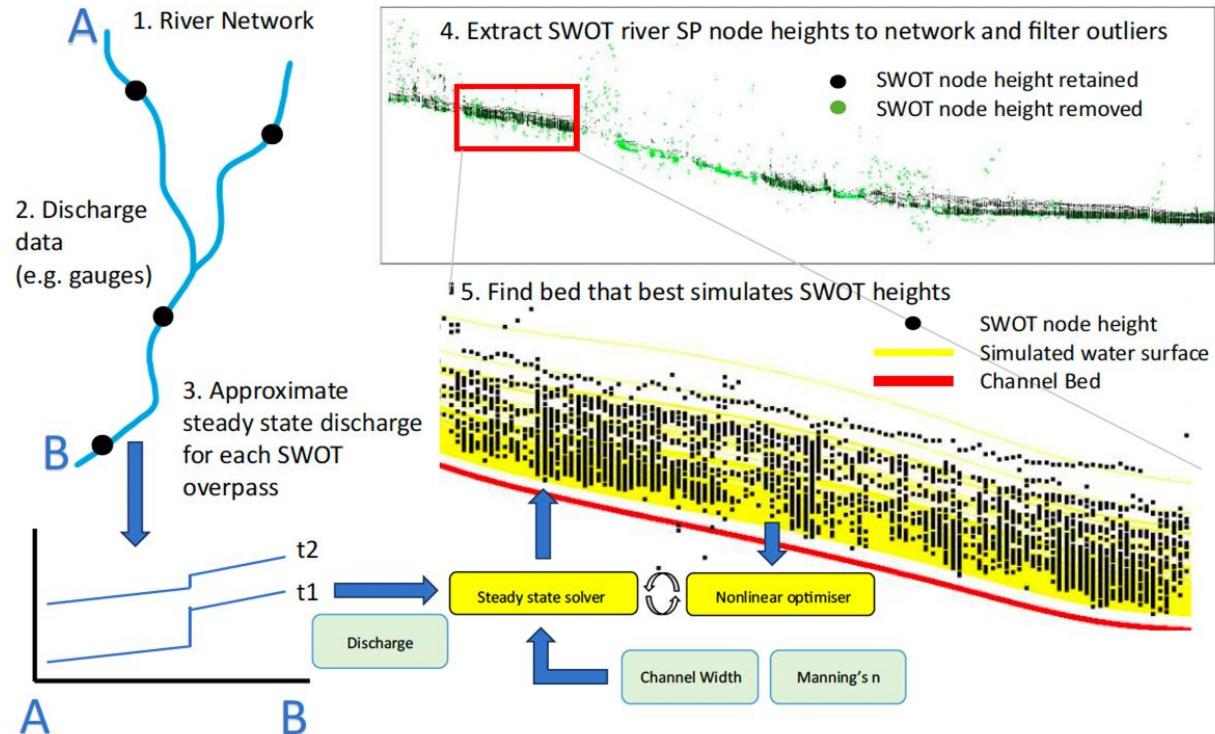




Inversion of 1D Gradually Varied Flow equations to solve for bed given a set of SWOT overpasses

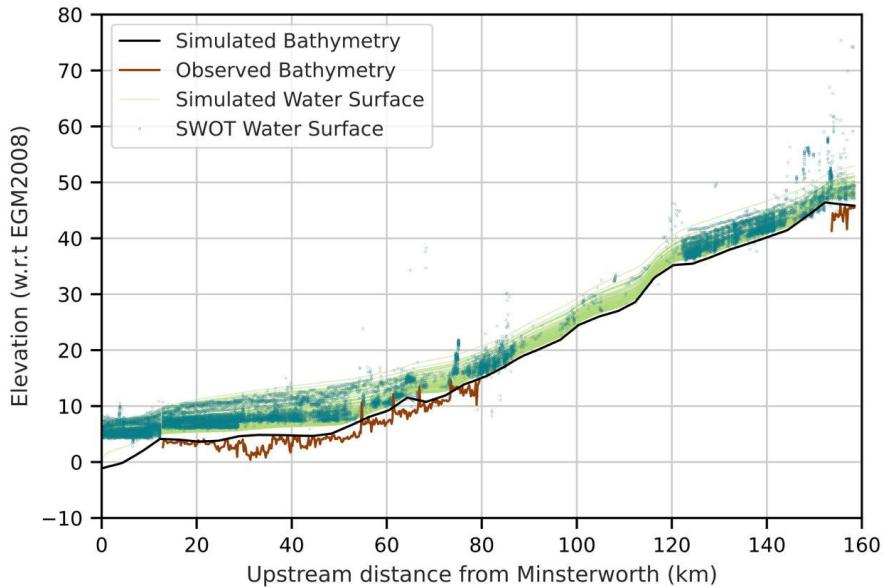
Two methods:

1. Non-linear optimiser (Neal et al., in review)
2. Physics-Informed Neural Network (PINN) (Rong et al., about to be submitted)

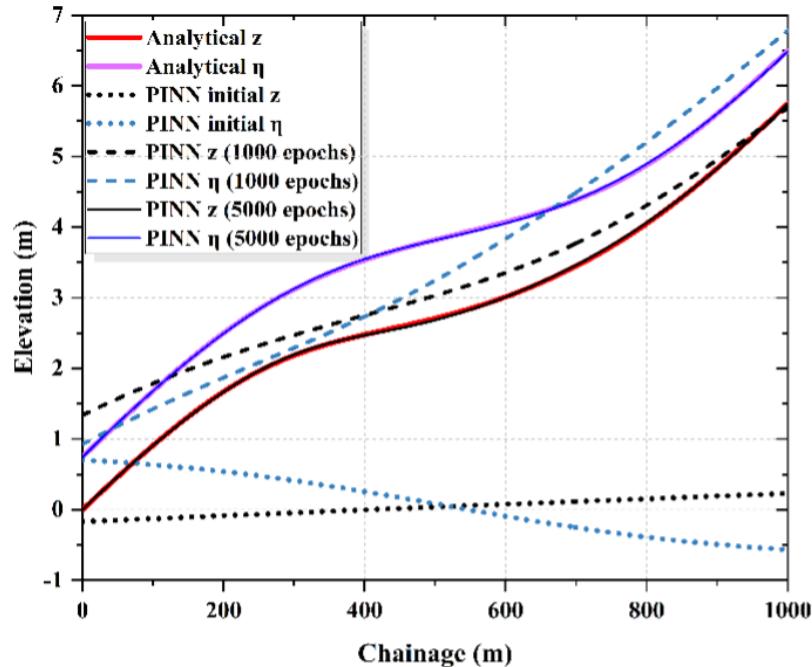




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(i) Non-linear optimiser  
PINN



(ii)

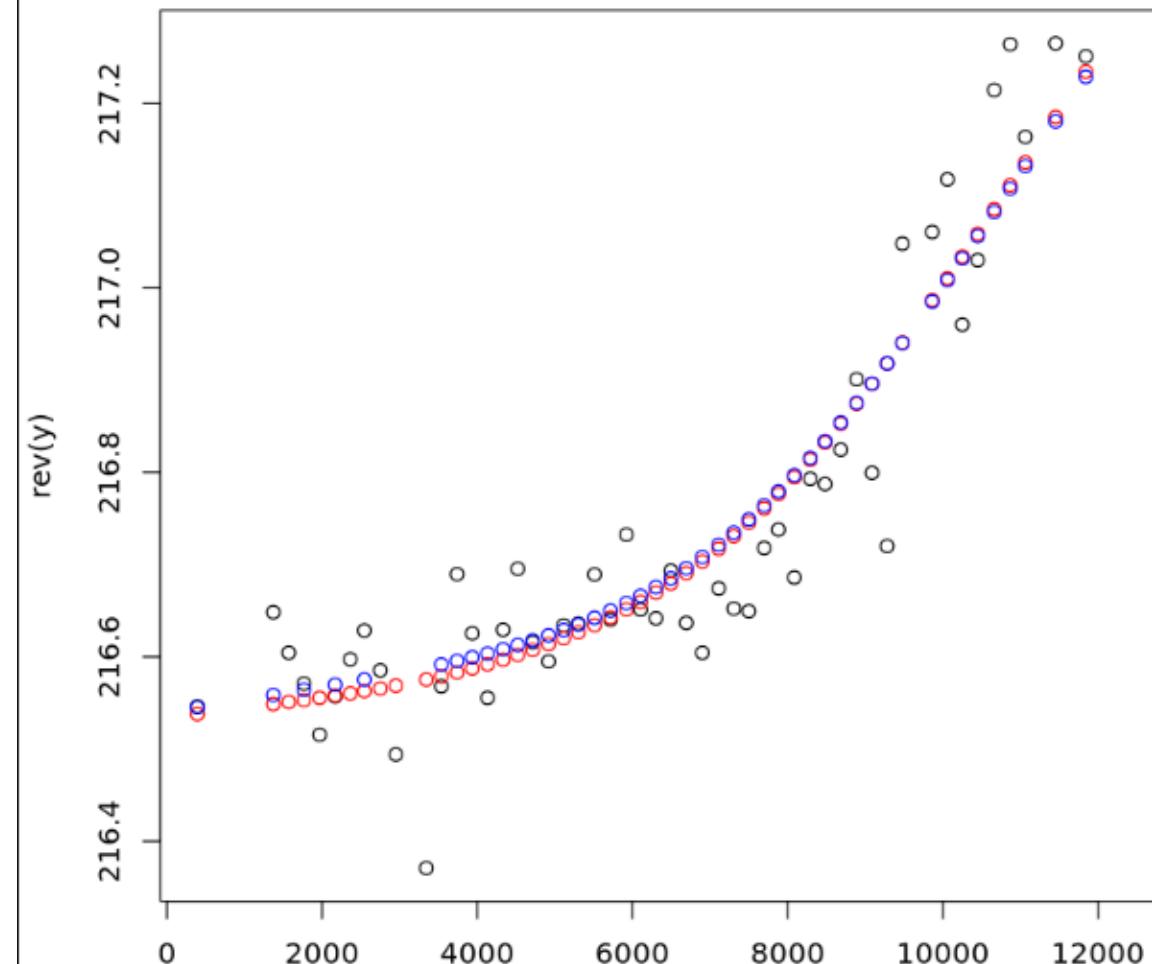
Co-solve for discharge and bathymetry:

Under GVF, and given a  $Q$  and channel, height is exactly specify. Solve for  $Q$  and channel geometry to match SWOT observations

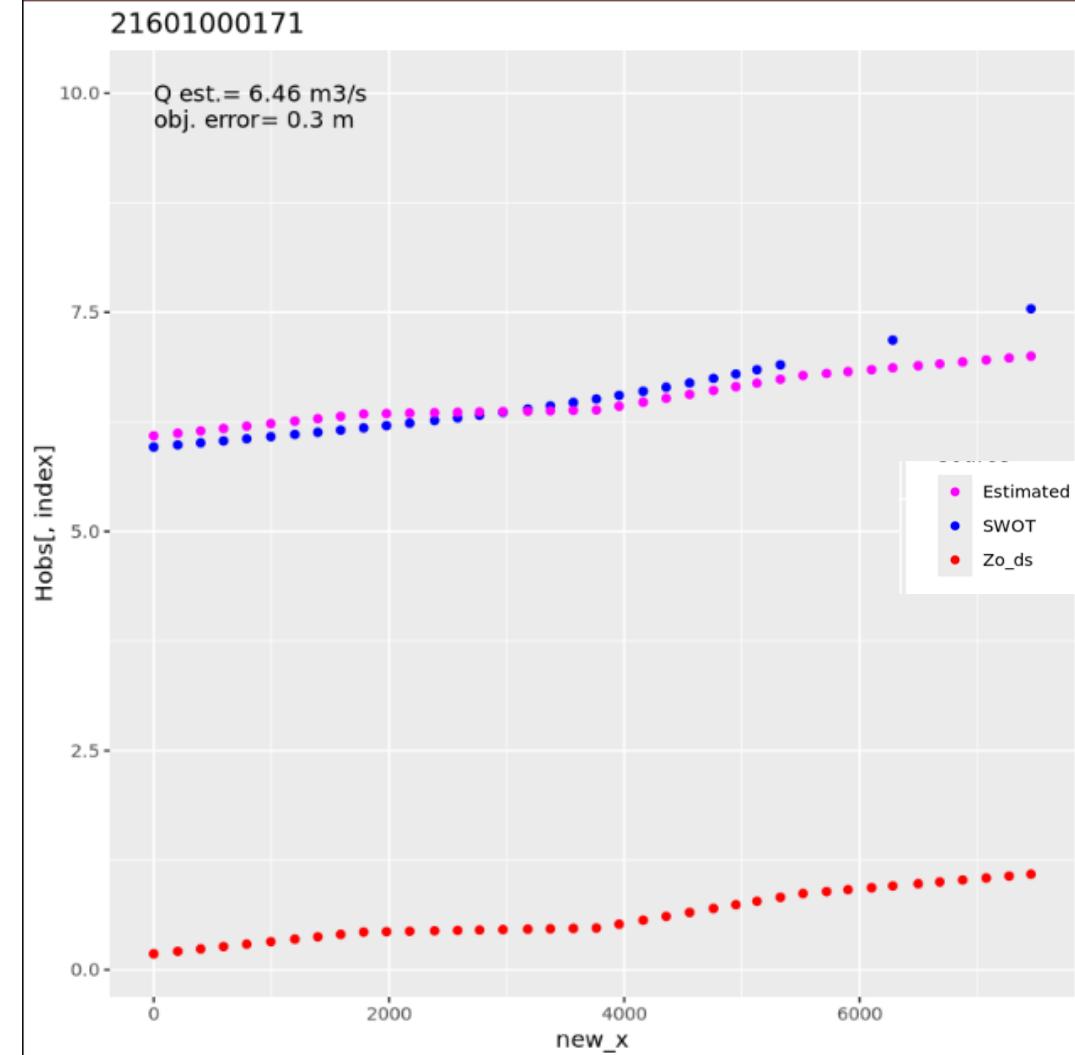
Black- filtered  
SWOT data

Red- first filter

Blue- 2nd filter



Given these filtered and smoothed water surfaces from SWOT, solve for a  $Q$  and parameter set (from the triplets) that best reconstructs the water surface under GVF



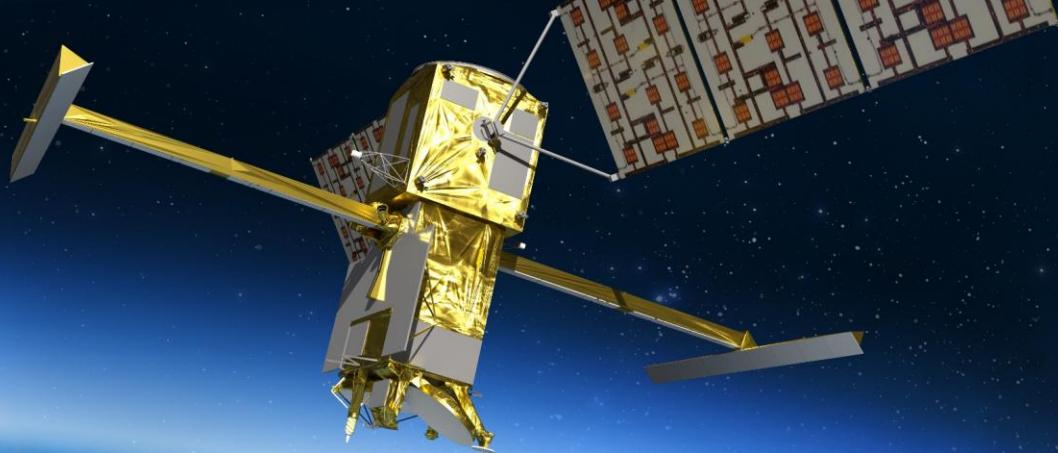
# Perspectives/Objectives

SWOT Bathymetry offline WG to standardize algorithms outputs and define relevant variables.

Bathymetry benchmark for inter-comparison (Gather Bathymetry data).

Define a Validation framework for 1D-simplified product.

Framework for global production (Confluence already has some of the algorithms).



Merci! Thank you

SWOT