

ALL-IN-ONE SWOT LAKE APPROACH

VOLUMES, BATHYMETRY AND UNCERTAINTY ESTIMATION

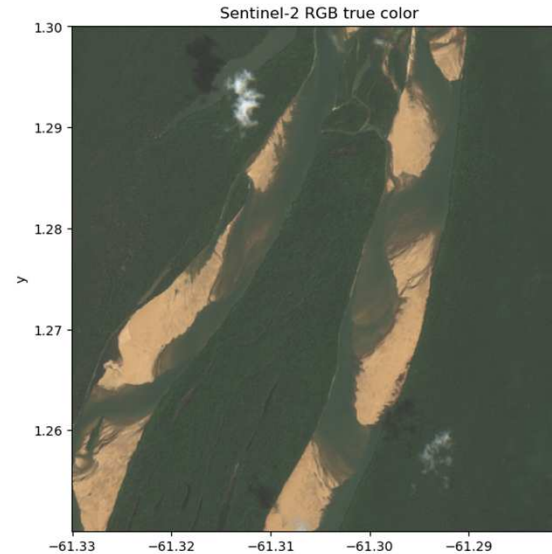
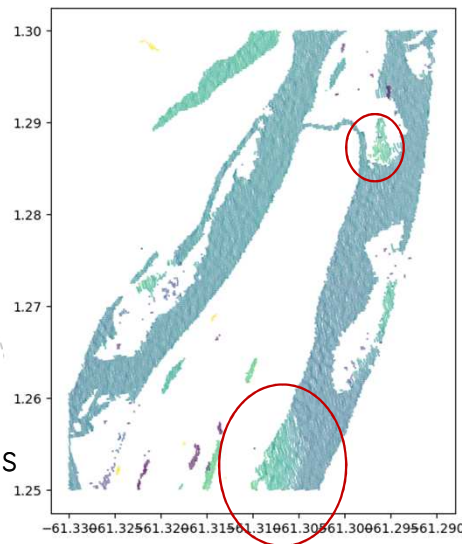
SWOT SCIENCE TEAM MEETING

14-17 OCTOBER 2025

Santiago PENA LUQUE (CNES)
Dawa DERKSEN
Nicolas GASNIER
Laetitia LALLA (CS-Soprasteria)
Benjamin TARDY

CONTEXT - PIXEL CLOUD OBSERVATIONS

- Evidence
 - SWOT PIXC shows over estimation of WATER class in Sandbanks
 - Water AREA/WIDTH estimation might be Wrong
- Opportunity
 - Actually, flat and smooth sandbanks may offer good geolocation/wse (high coherence and SNR) –(E.Salameh et al, 2024)



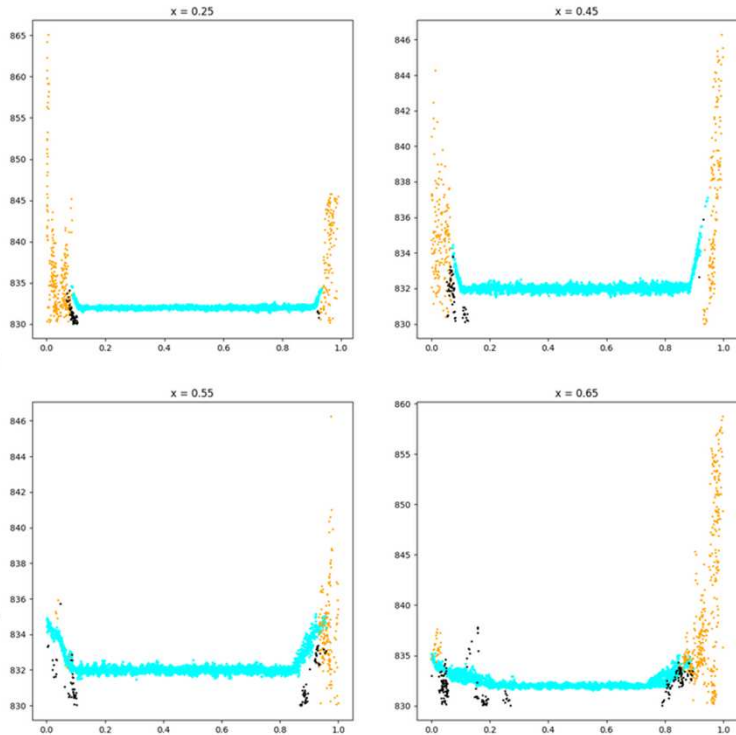
Rio Branco
Sentinel2

- Questions
 - How could we use WSE from sandbanks?
 - How to reclassify water areas ?
 - Do we take Land pixels as well?
 - How to automate it?

IMPLICIT NEURAL REPRESENTATION - LAKES

INPUT

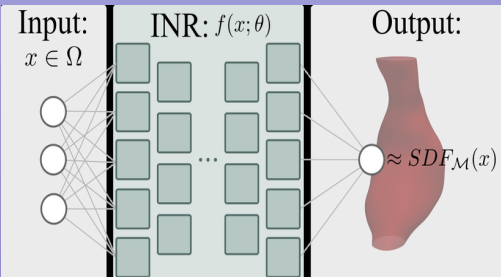
4D SWOT Point Cloud
 $\{x_i, y_i, z_i, t_i\}, i \in [1 \dots N]$



2 branches model:

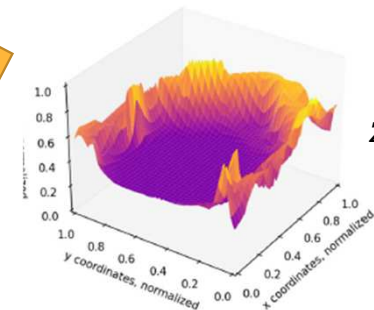
- Static : Ground/Bathy (x,y)
- Dynamic : Water level (t)

Implicit neural representation

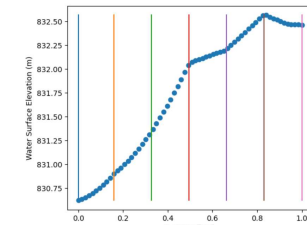


Example of 3D volumes (inspired by NERF)

Implicit representation of Ground DSM/Bathymetry



Water level timeseries



Max or softmax

Z_w

Z_g

IMPLICIT-LAKE : INVERSE PROBLEM


Uncertainty assesment

Forward rendering / simulation

$$\hat{z}(x, y, t) = \max(z_g, z_w)$$

Iterative process to fit the implicit model to the data

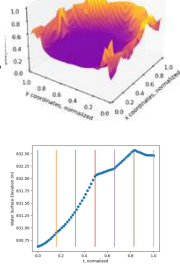
SWOT PIXC Data
 $\{x_i, y_i, z_i, t_i\}, i \in [1 \dots N]$



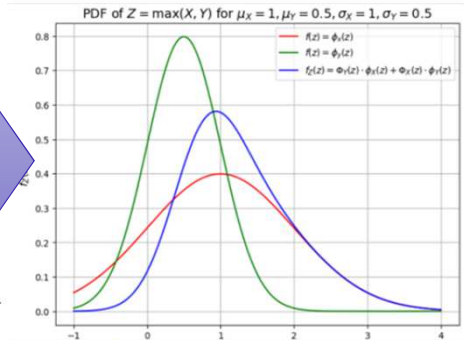
2-branch MLP parameters « θ »

$z_g(x, y)$

$z_w(t)$



Probabilistic distributions



Output variables

Water presence

Surface

Volume

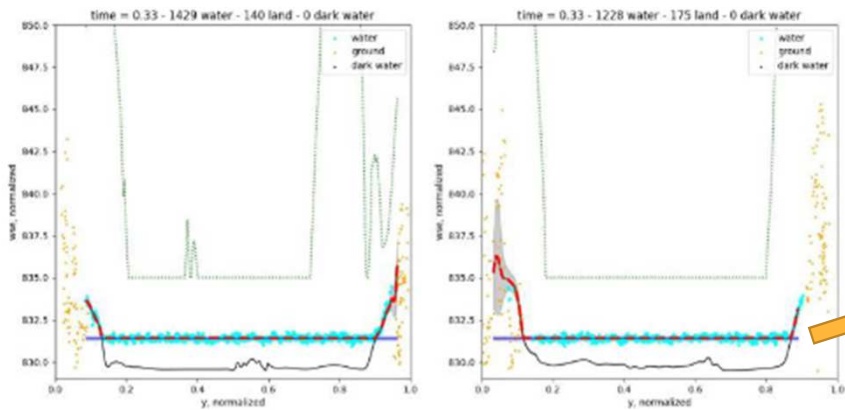
phase_noise_std

Cost function
 « Data fidelity »

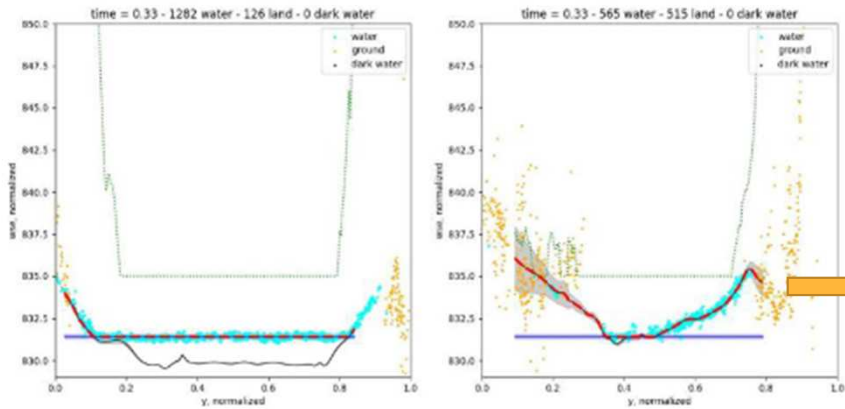
$$\nabla_{\theta} (\hat{z}(x_i, y_i, t_i) - z_i)^2$$

- Physical priors :
- The DEM/Bathymetry does not change through time
 - The water has the same elevation everywhere on the ROI : lake area
 - The observed elevation is the max of ground and water elevations

RESULTING DISTRIBUTIONS

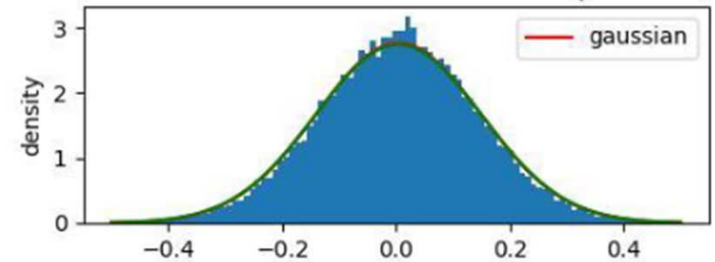


**Water
Distribution
 $F(t)$**



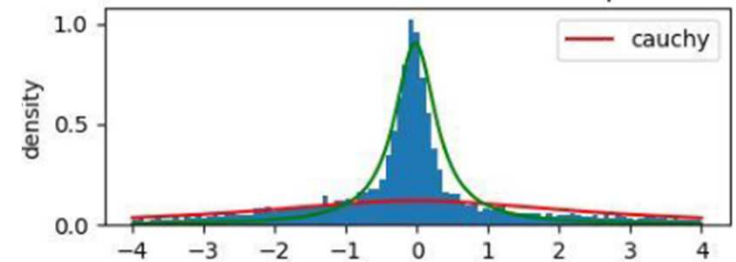
**Ground
distribution
 $F(x,y)$**

WATER ERROR = GT swot wse - model prediction h



mean(wse-h) -> MAD=0.45cm
 mean(|wse-h|) -> AAD=11.23cm
 Red : Distrib with std = STD des erreurs =14.43cm
 Green : Distrib with std = Average Sigma predicted =14.55cm

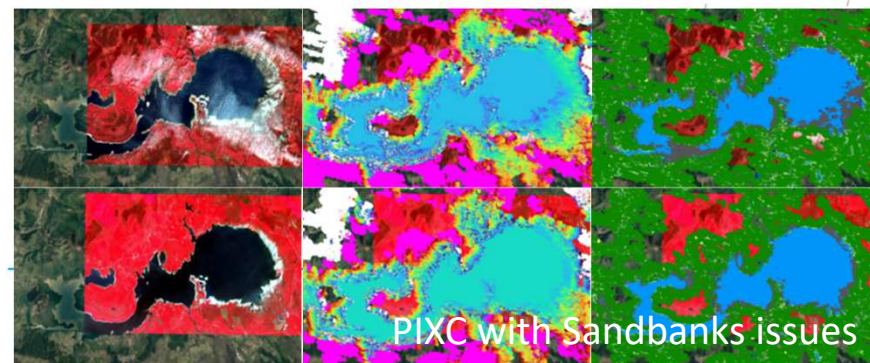
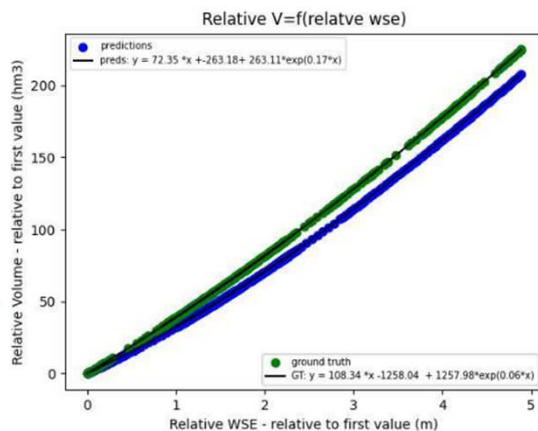
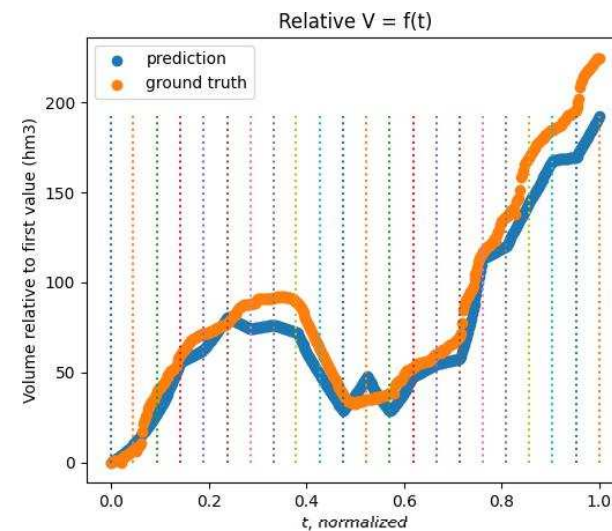
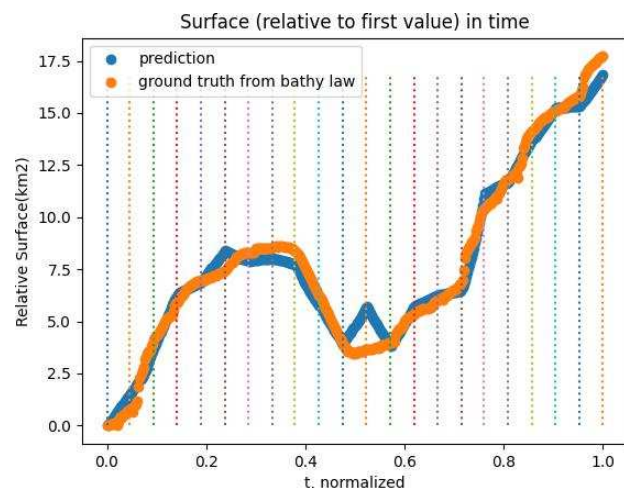
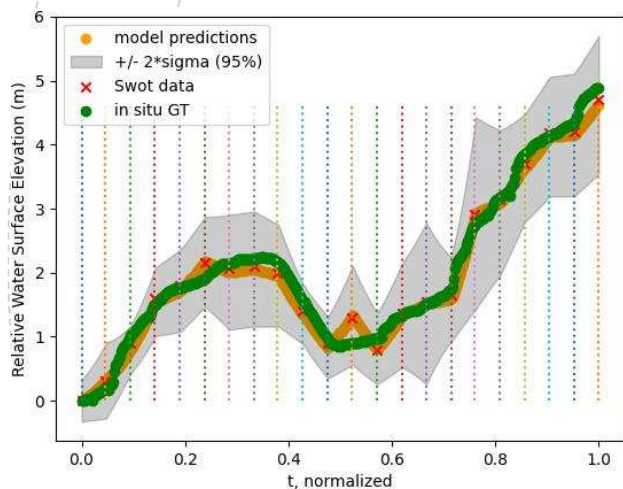
GROUND ERROR : GT swot wse - model prediction h



mean(wse-h) -> MAD=-2.41cm
 mean(|wse-h|) -> AAD=154.26cm
 Red : Distrib with std = STD des erreurs =263.07cm
 Green : Distrib with std = Average Sigma predicted =35.16cm

RESULTS - EBRO RESERVOIR (SPAIN)

Implicit Lake : Reservoir Stock Monitoring from Multitemporal SWOT Pixel Cloud
 (D.Derksen, N.Gasnier, S.Pena, L.Lalla) [on going]



Diapositive 6

PLS2

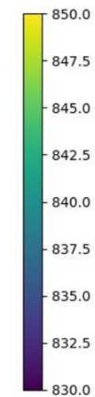
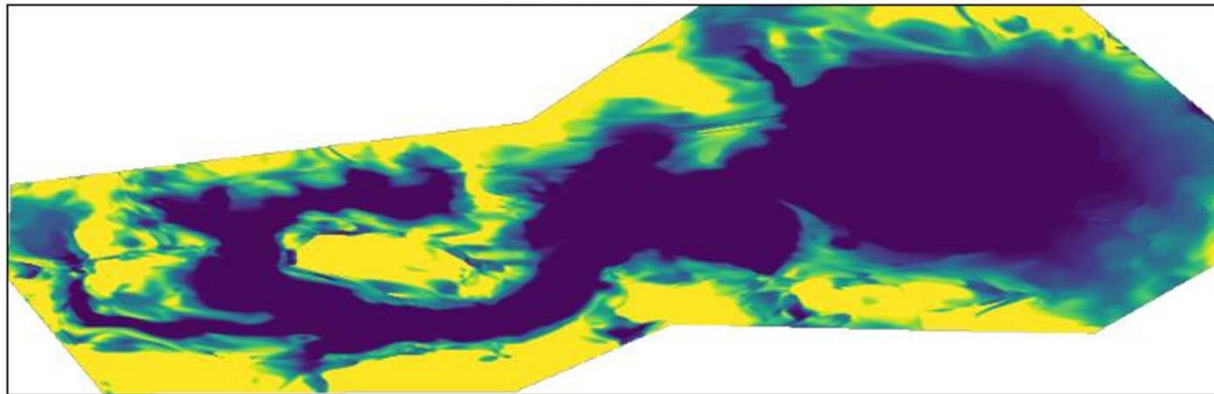
Pena Luque Santiago; 14/10/2025

RESULTS - EBRO RESERVOIR (SPAIN)

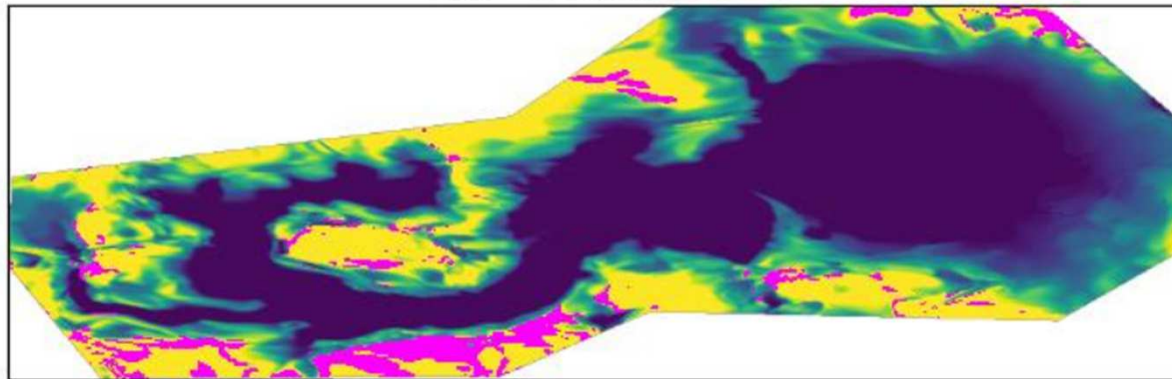
Spatialized & Probabilistic Results

- Ground DSM: Most probable solution
- Evaluation - Pléiades DSM on wet bathymetry
 - 0.7m RMSE

Predicted z
vmin=830, vmax=850



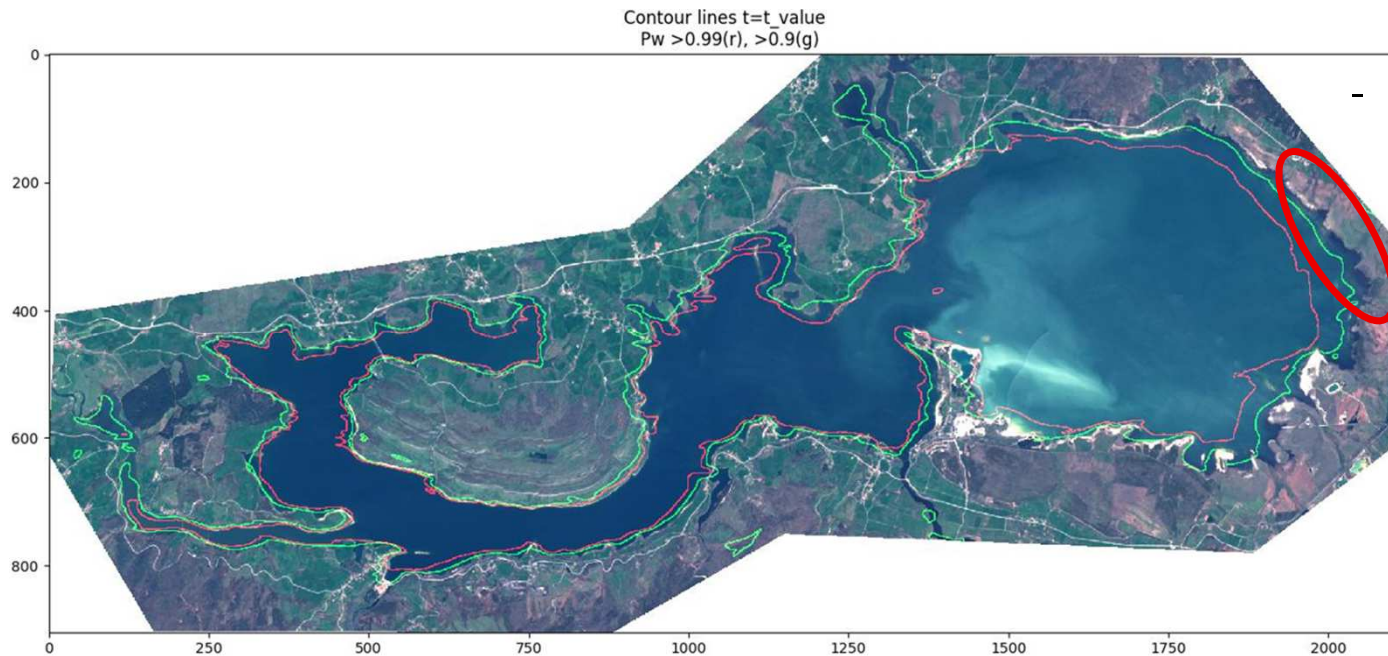
Predicted z
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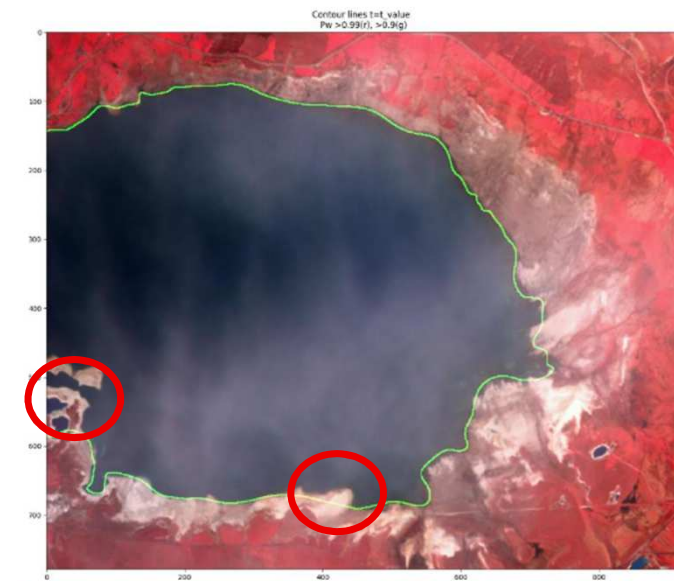
- Pink = High uncertainty (high predicted STD)

RESULTS - EBRO RESERVOIR (SPAIN)

Spatialized & Probabilistic Results



- Prone to area smoothing



- Smoothness of results

- Predicted Water Extent (SWOT) compared to simultaneous Sentinel 2

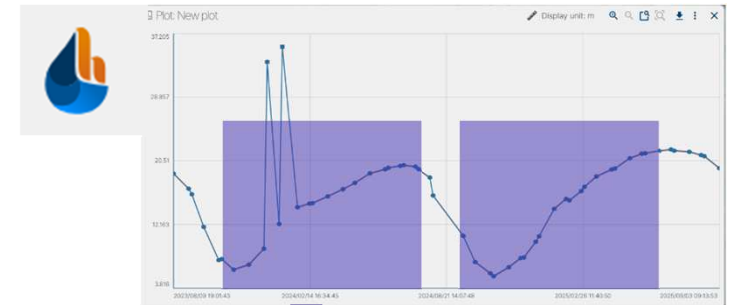
FLOOD PLAIN DEM / RIVERS

Amazon – Manaus



Manaos – Solimoes River

RiverSP – Drought periods 2023 & 24



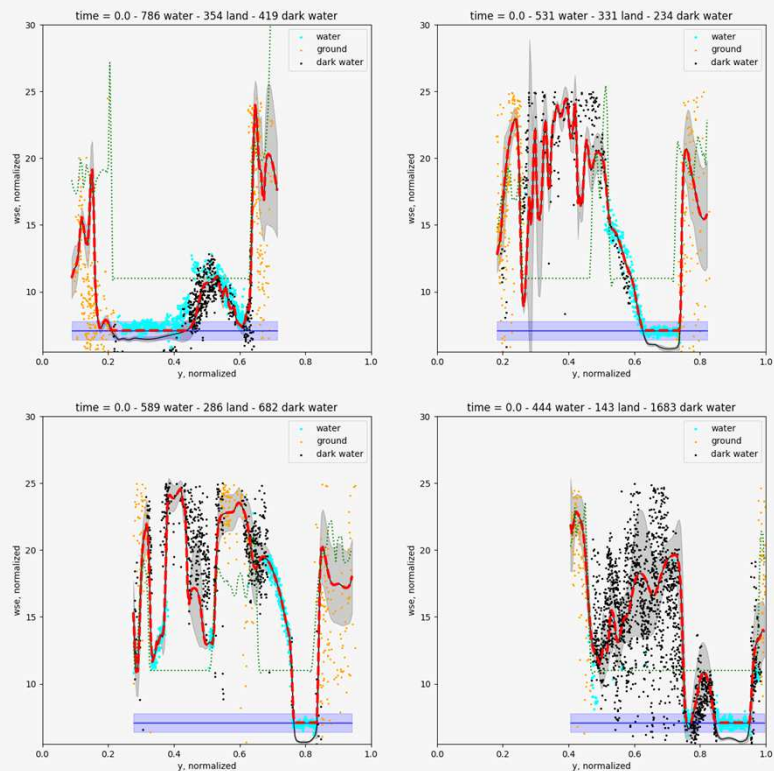
SWOT PIXC « Water » pixels - WSE

Hypothesis:
Similar WSE in ROI
<
SWOT RMSE (10,
15cm)

IMPLICIT RIVER

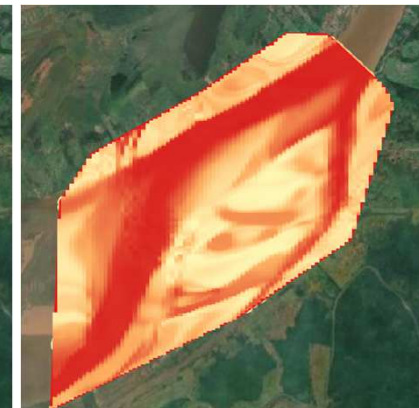
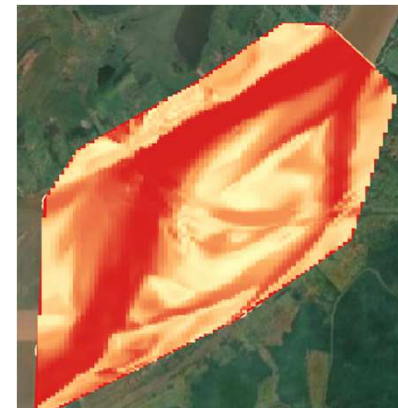
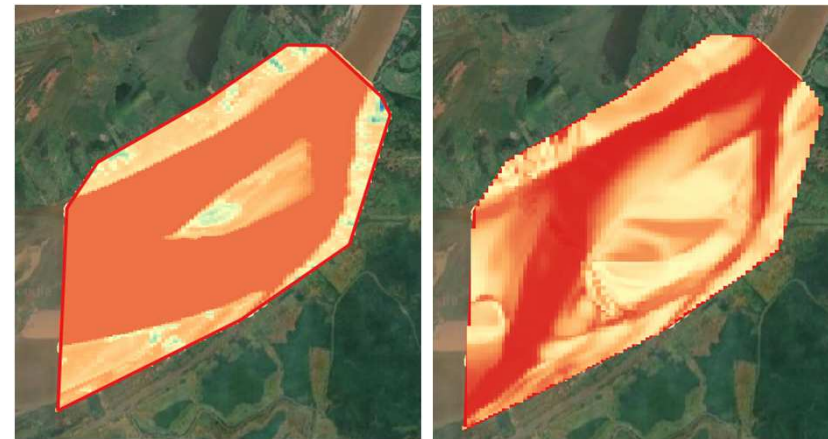
Estimation of
DEM+ Wet bathymetry

Preliminary Results from [Implicit river](#) :
Water levels and wet bathymetry



Copernicus 30m

Estimation DEM - 149L



Estimation DEM - 160L

Estimation DEM - ALL

CONCLUSION

- Automatic approach with just SWOT PIXC (independent from PIXC Classification)
- Management of Probability Distribution functions to handle Uncertainty
- Limitations:
 - Only one Water body (WSE) to be processed in ROI
 - Hyperparameters sensitivity
 - Static DSM/Bathymetry hypothesis >> Pay attention to Bathymetry changes/ Seasons



Candidate for « FloodPlainDEM »
Estimation on SWOT Expertise Center
(HAWK) – Coordination with
D.Desroches/R.Fjortoft

Perspective:

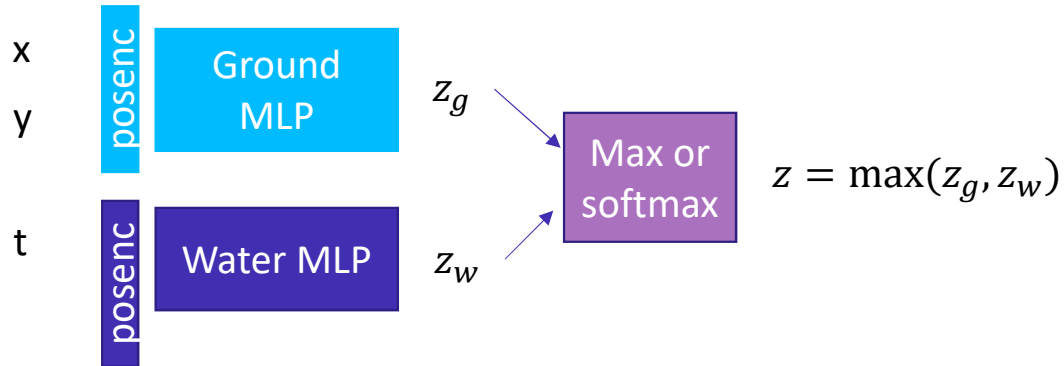
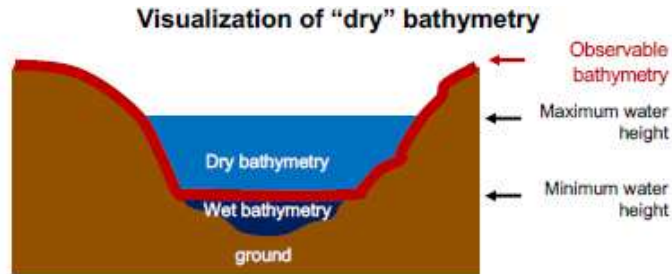
- Ready to be applied on Rivers and therefore, capabilities on Flood Plain DEM generation
- Better **Hyperparameters** management
- Analysis on Observation angles and complementarity of different orbits.
- Multiple water bodies to be segmented in ROI areas (nearby lakes, multiple water bodies reservoirs, etc)



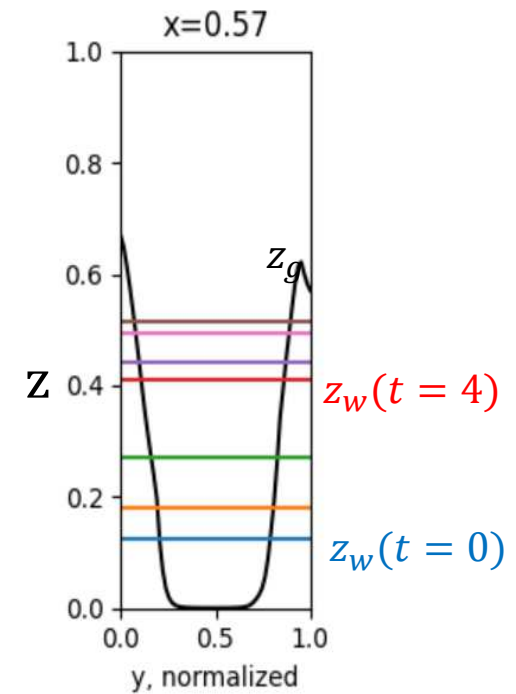
THANKS
FOR YOUR ATTENTION

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IMPLICIT NEURAL REPRESENTATION



$$\tilde{z} = [z_g, z_w] \times \text{softmax}(\lambda, [z_g, z_w])^T$$



IMPLICIT LAKE - PROBABILITY FUNCTIONS

Implicit model

- Digital Elevation Model

$$f : (x, y) \rightarrow z_g$$

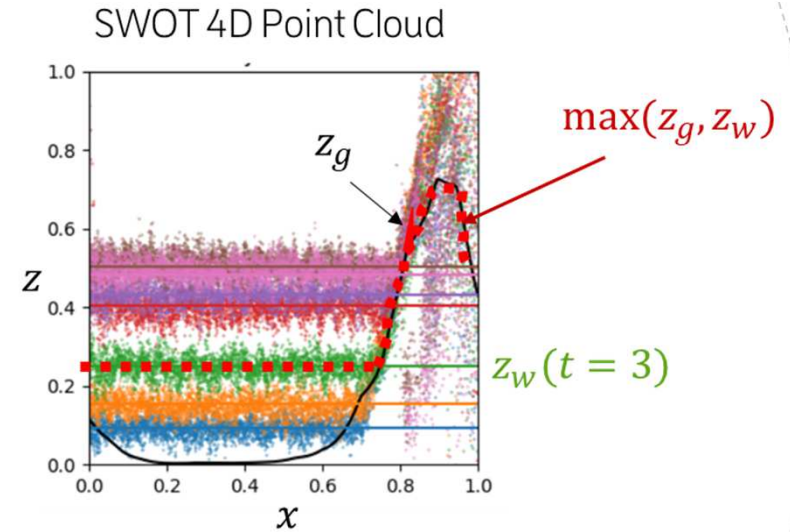
- Water elevation

$$f : (t) \rightarrow z_w$$

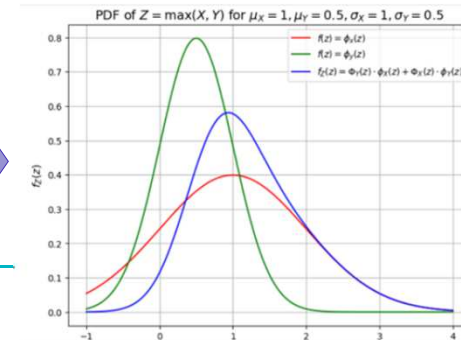
- Observed elevation

$$f : (x, y, t) \rightarrow z = \max(z_g, z_w)$$

Results :



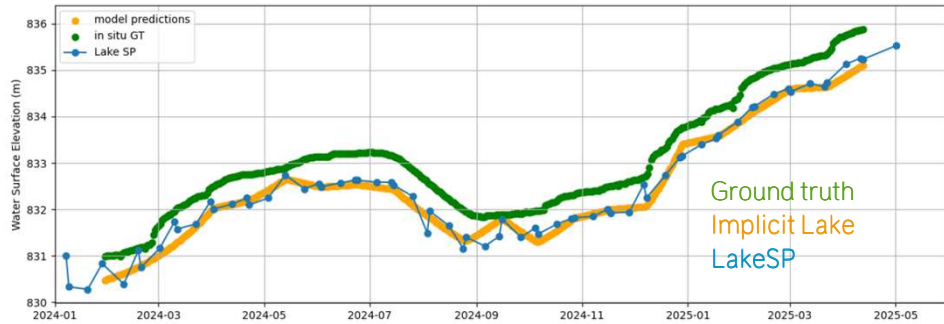
Probabilistic distributions



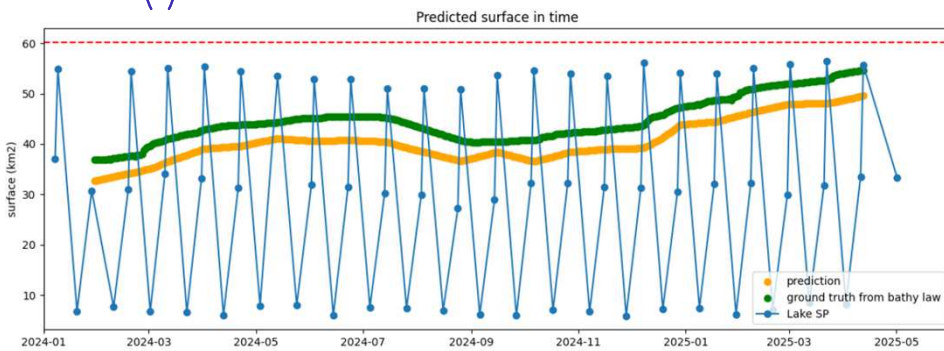
Uncertainty assesment

COMPARISON OF VOLUME ESTIMATIONS

WSE (t)

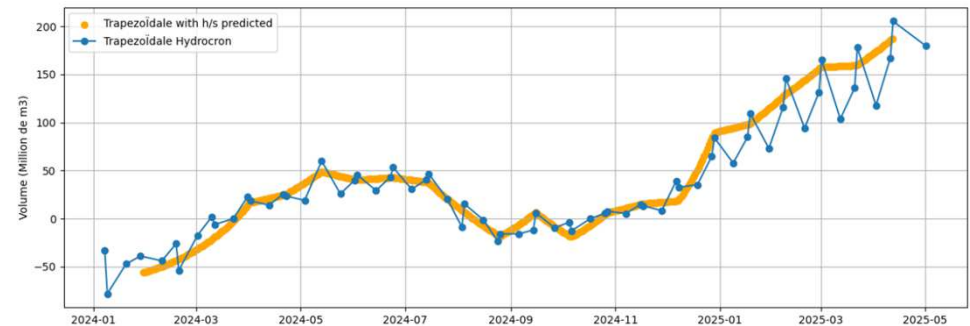


Area(t)



* Warning: Lake SP includes partial views

Volume change (t) – Trapezoidal approach



Improvement on Volume estimations compared to Lake SP estimations