EspaceDEV OBSERVATION SPATIALE, MODÈLES & SCIENCE IMPLIQUÉE

The CASCADES project: Automated processing chains for the monitoring of inland surface water using satellite images Renaud Hostache, Thibault Catry, Chhenglang Heng, Christophe Révillion, Vincent Herbreteau, Vannak Ann

Institut de Recherche pour le Développement F R A N C E









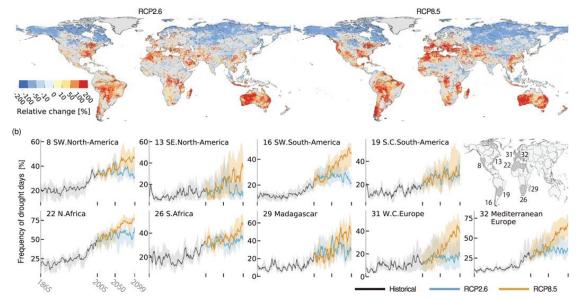
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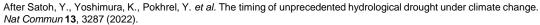




Context

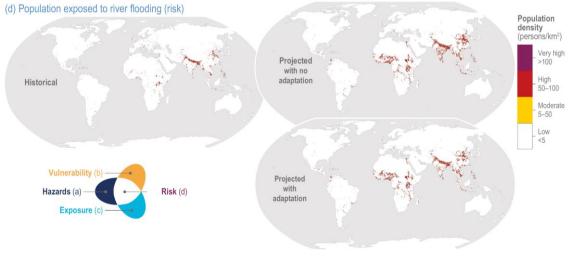
Water is one of the most impacted ressources by climate change with more and more intense extreme events and by increasing societal demand (Kreibich et al., Naure, 2022). This increases the vulnerability of social-ecological systems and further augments tensions between potentially competing uses (usage vs environmental safeguarding).





Context

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Source : IPCC

It is of primary importance to develop tools to evaluate and monitor the availability of water resources in a continuous and regular manner, as well as follow its trajectory over time to better understand and assess the impact of global changes on it.

CASCADES's objectives

1. Automatise from end to end Earth observation satellite image processing chains (Sentinel-1 et 2) to provide the following monitoring products:

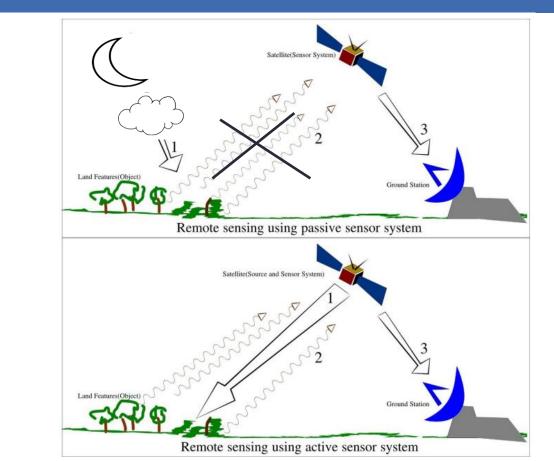
- Surface water maps
- Flood depth and elevation maps during large flood events, by combining water extent maps and CopDEM)

2. Co-develop a plateform in a participatory way with end-users to provide access to the maps.

The processing chains will be based on several algorithms developped and frequently utilized by project partners (e.g., Matgen et al. 2011, Chini et al. 2017, Alexandre et al., 2020, Frappart et al., 2021, Johary et al. 2023, Ayoub et al. Submitted).

 \rightarrow This will allow us to systematically process S-1 and S-2 images over selected areas and will therefore enable a continuous temporal monitoring.

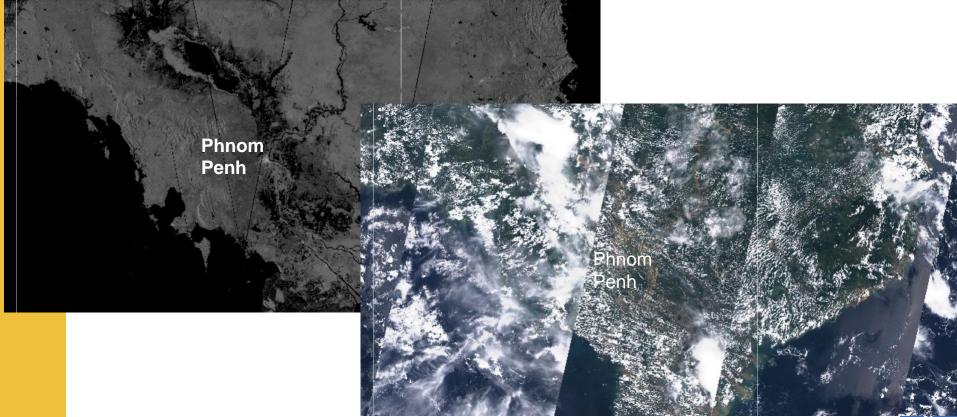
Active and passive sensors for flood monitoring



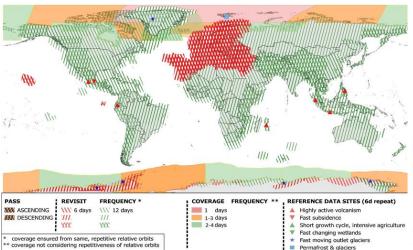
Optical remote sensing

RADAR remote sensing

Active and passive sensors: S-1 and S-2



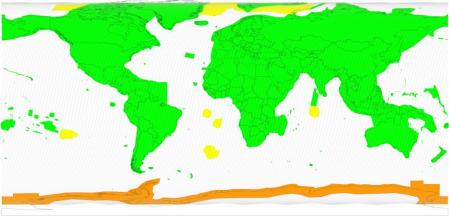
Sentinel -1 revisit time



Sentinel-1 constellation observation, revisit and coverage scenario (April 2021), after Mullissa et al (2021) Sentinel-2 Constellation Observation Scenario: Revisit Frequency

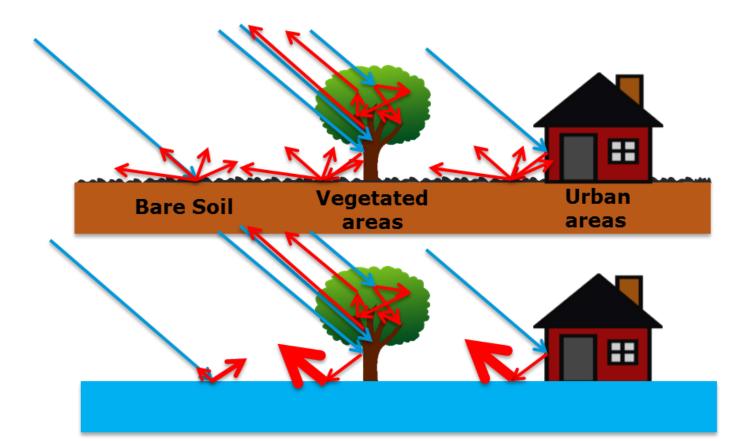


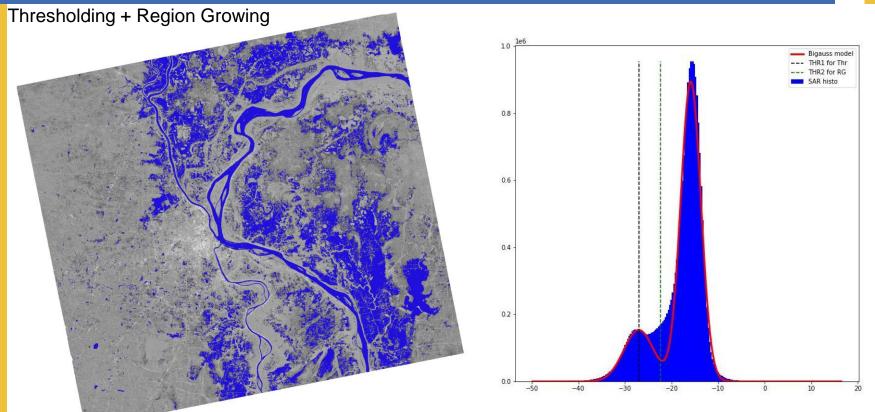
Validity start: June 2022



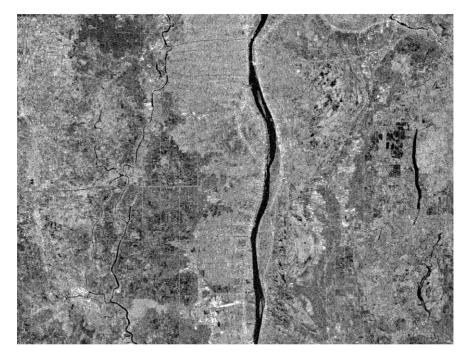
5 days 10 days 10 days access from alternated tracks

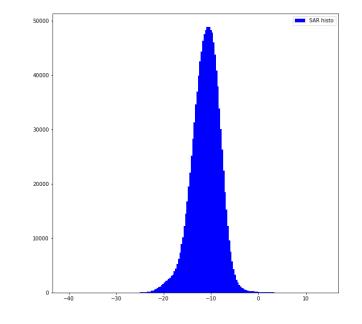
Detection of water on SAR images



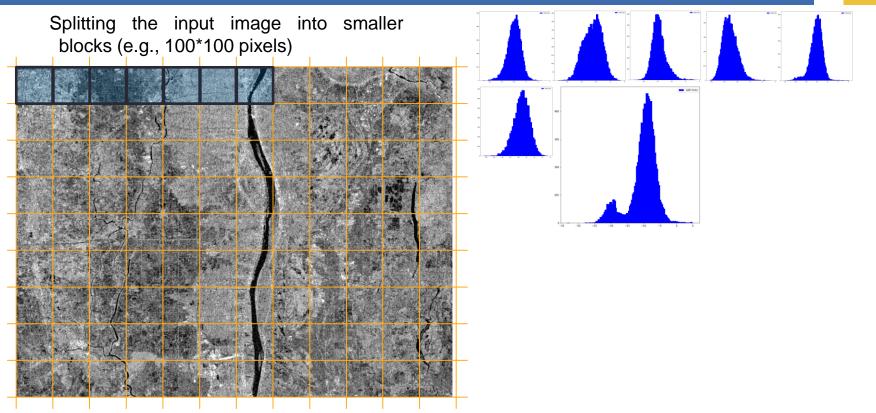


P. Matgen, R. Hostache, G. Schumann, L. Pfister, L. Hoffmann & H.H.G. Savenije. Towards an automated SAR-based flood monitoring system: Lessons learned from two case studies. PCE, 36(7-8):241-252, 2011, https://doi.org/10.1016/j.pce.2010.12.009.



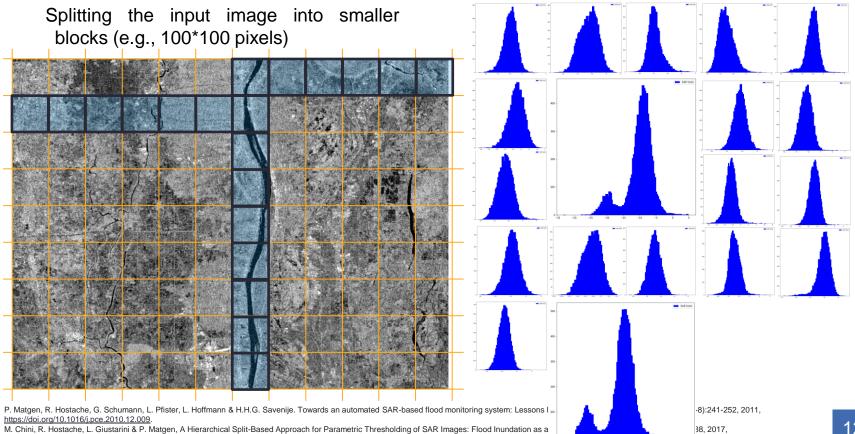


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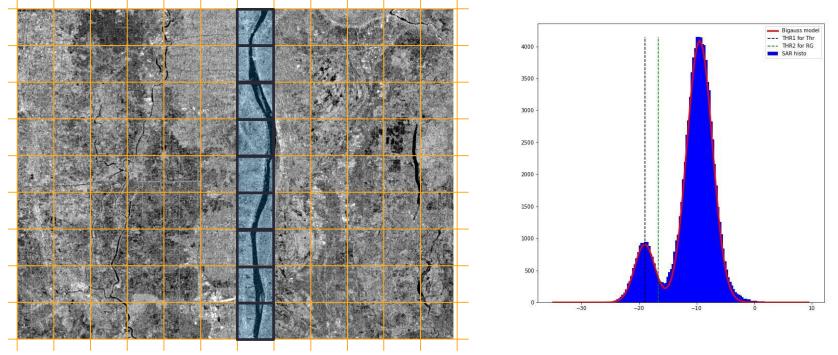


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https://doi.org/10.1109/TGRS.2017.2737664

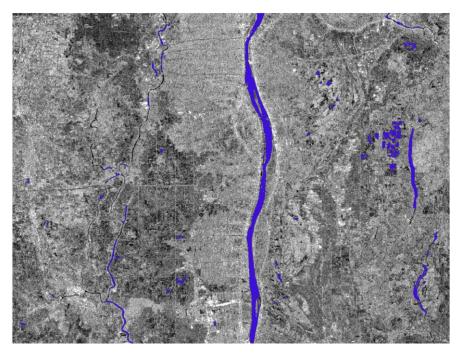


Splitting the input image into smaller blocks (e.g., 100*100 pixels)

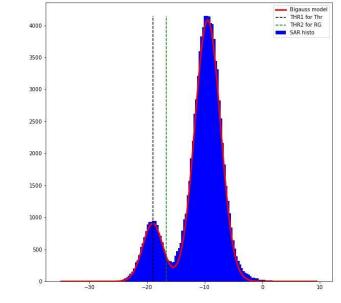


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Thresholding + Region Growing



The algorithm also integrates Change Detection on the same principle and will soon integrate VV and VH polarization

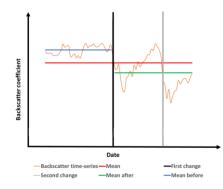


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Surface water mapping using S1: 3 algorithms

CuSum is an algorithm designed for time series processing (Manogaran and Lopez, 2018).

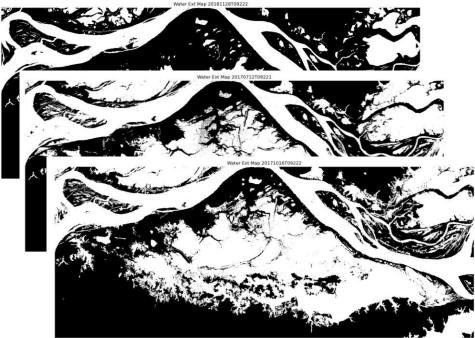
It uses a unique input parameter: Tc the critical threshold corresponding to the level of confidence of a change detection (bootstrap analysis). Cumsum is applied on VV and VH polarization.



Ygorra, B., Frappart, F., Wigneron, J. P., Moisy, C., Catry, T., Baup, F., ... & Riazanoff, S. (2021). Monitoring loss of tropical forest cover from Sentinel-1 time-series: A CuSum-based approach. International journal of applied earth observation and geoinformation, 103, 102532.

Ygorra, B., Frappart, F., Wigneron, J. P., Moisy, C., Catry, T., Pillot, B., ... & Riazanoff, S. (2023). ReCuSum: A polyvalent method to monitor tropical forest disturbances. *ISPRS Journal of Photogrammetry and Remote Sensing*, 203, 358-372.

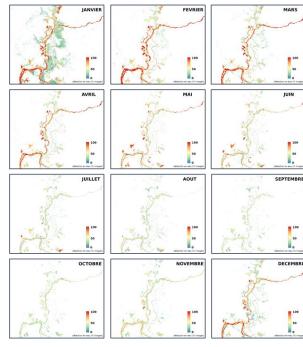
A third approach will be also applied and evaluated: This automated method is based on multotsu thresholding applied to VV and VH polarizations

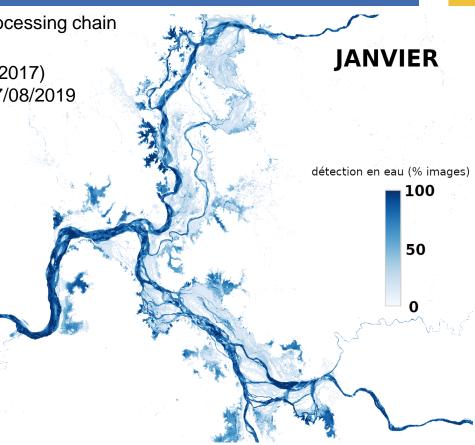


Master Thesis, Ana Carolina Pires Pereira

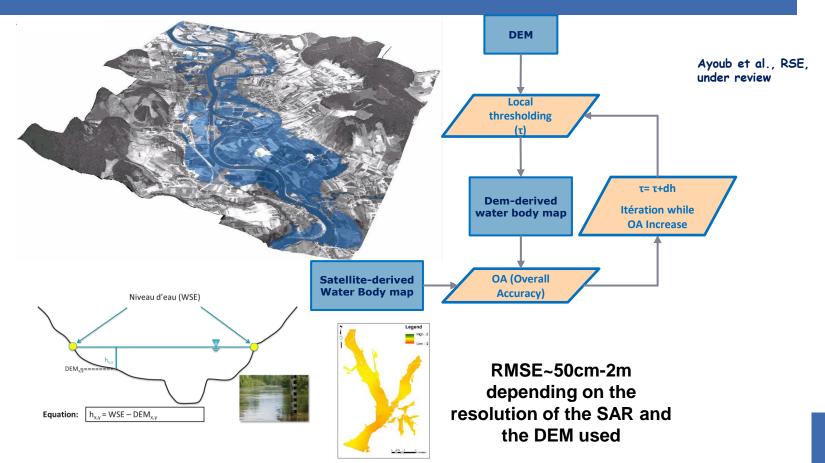
Analysis of Water Class using Sen2Change processing chain

- 225 images
- Reference taken from dry season (11/10/2017)
- Monthly composite from 22/10/2015 -> 17/08/2019

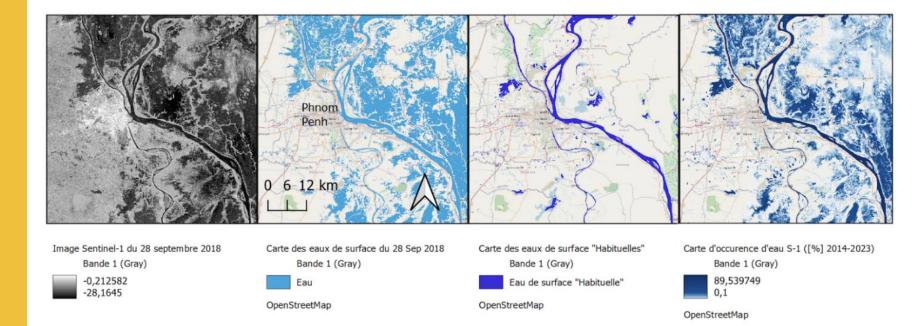




Water level estimation during flooding periods



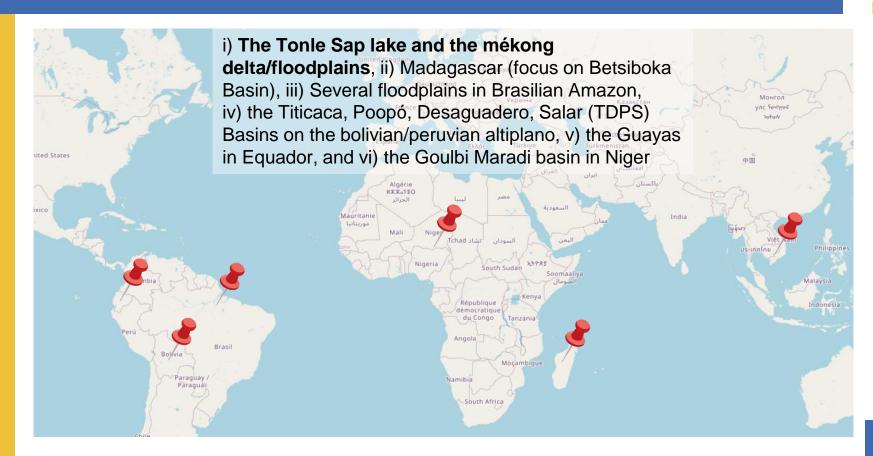
Various products



Output products will be made available via a platform codesigned with end

users

Several markedly different test case around the world





Questions?

