



## The use of innovative technology for fluvial monitoring

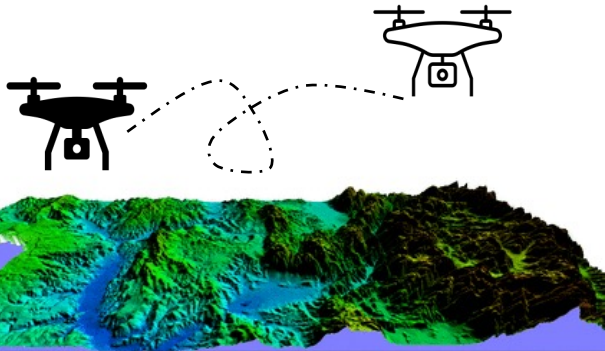
Gravel Bed Rivers 9 - 13 January 2023

**Alonso Pizarro**, Hernán Alcayaga, Matías García, Matthew Perks, Robert Ljubicic, Salvatore Manfreda, Silvano Dal Sasso

[www.alonsopizarro.cl](http://www.alonsopizarro.cl)

[alonso.pizarro@mail.udp.cl](mailto:alonso.pizarro@mail.udp.cl)

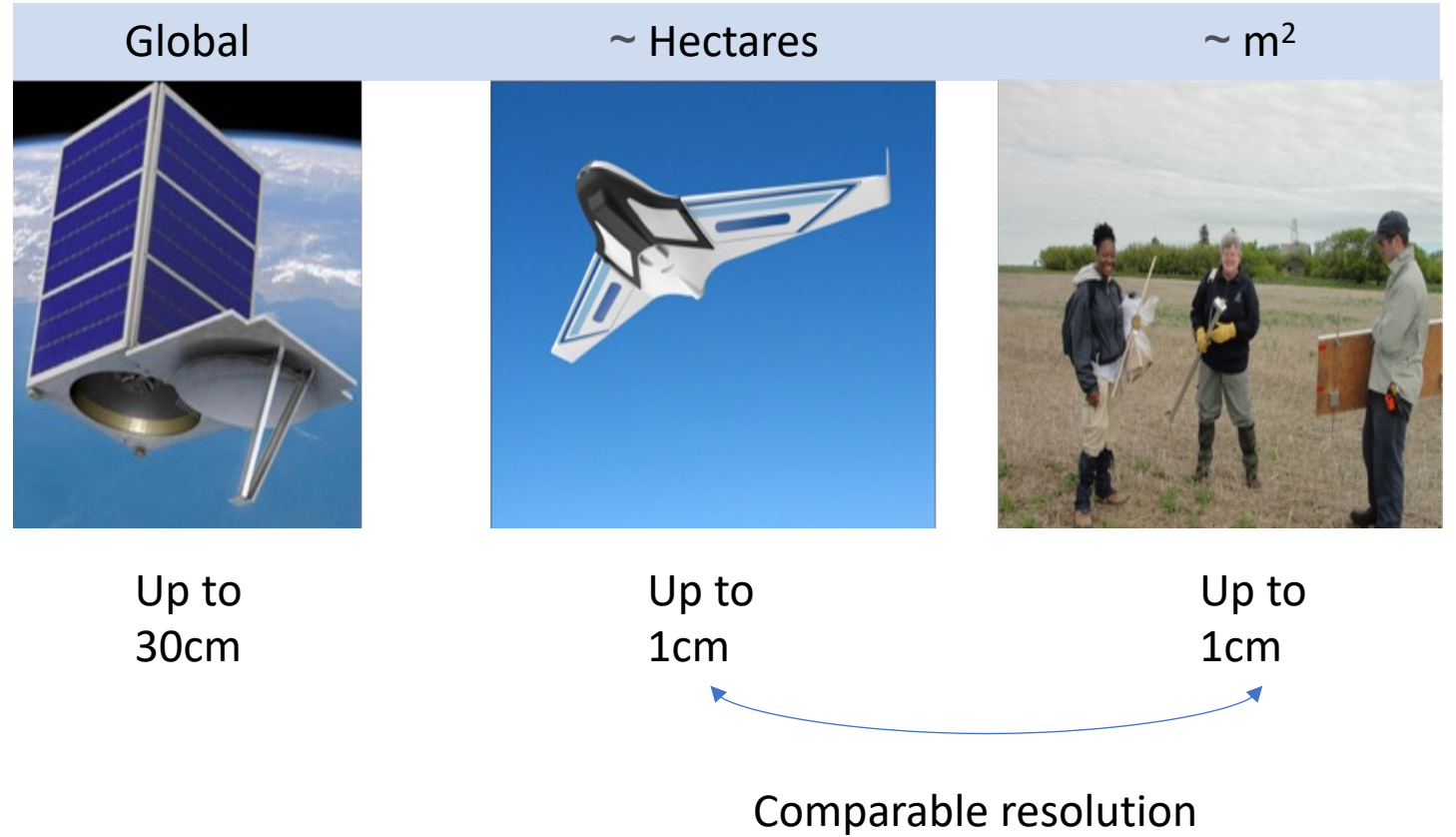
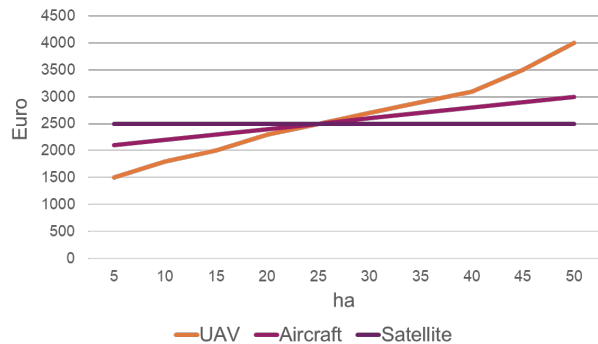
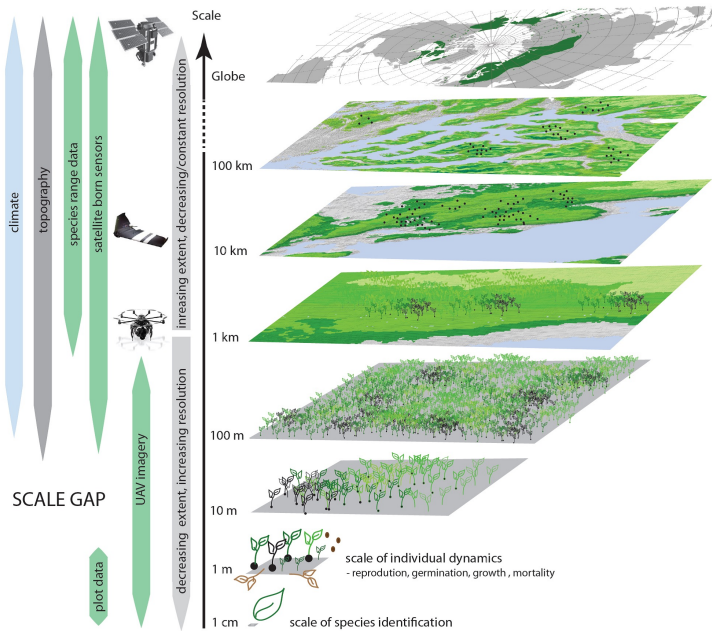
@AlonsoPizarroV



# Environmental monitoring at a glance

2

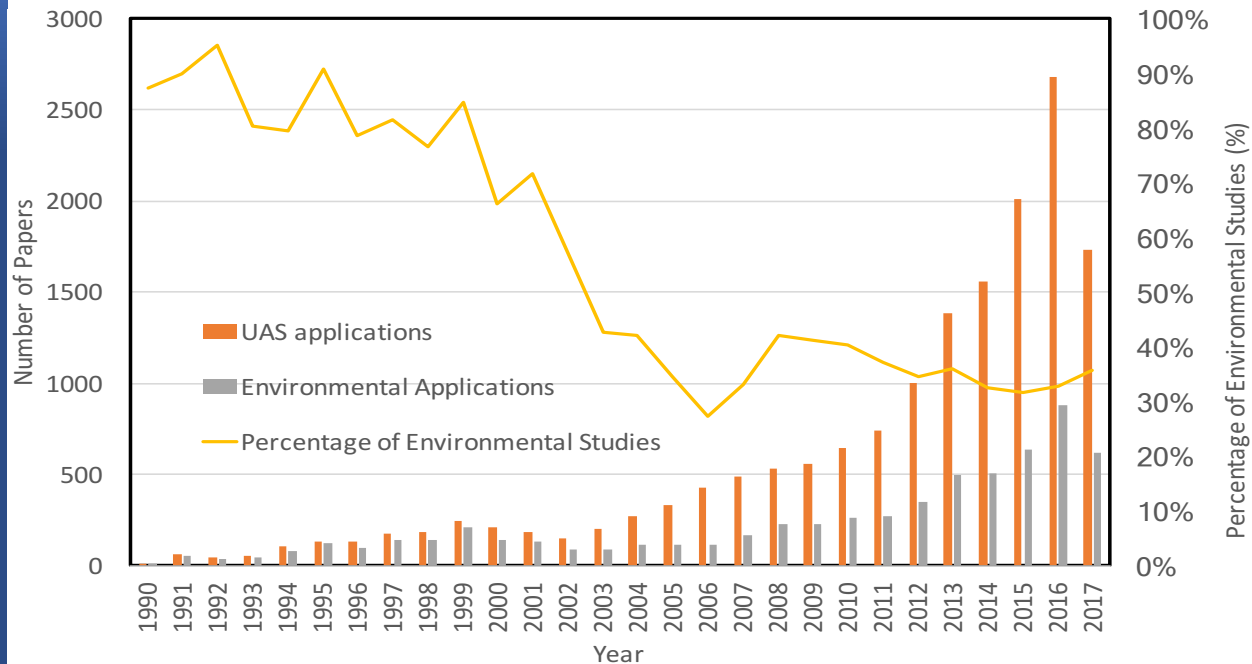
Different sensing alternatives. Which one should I use?



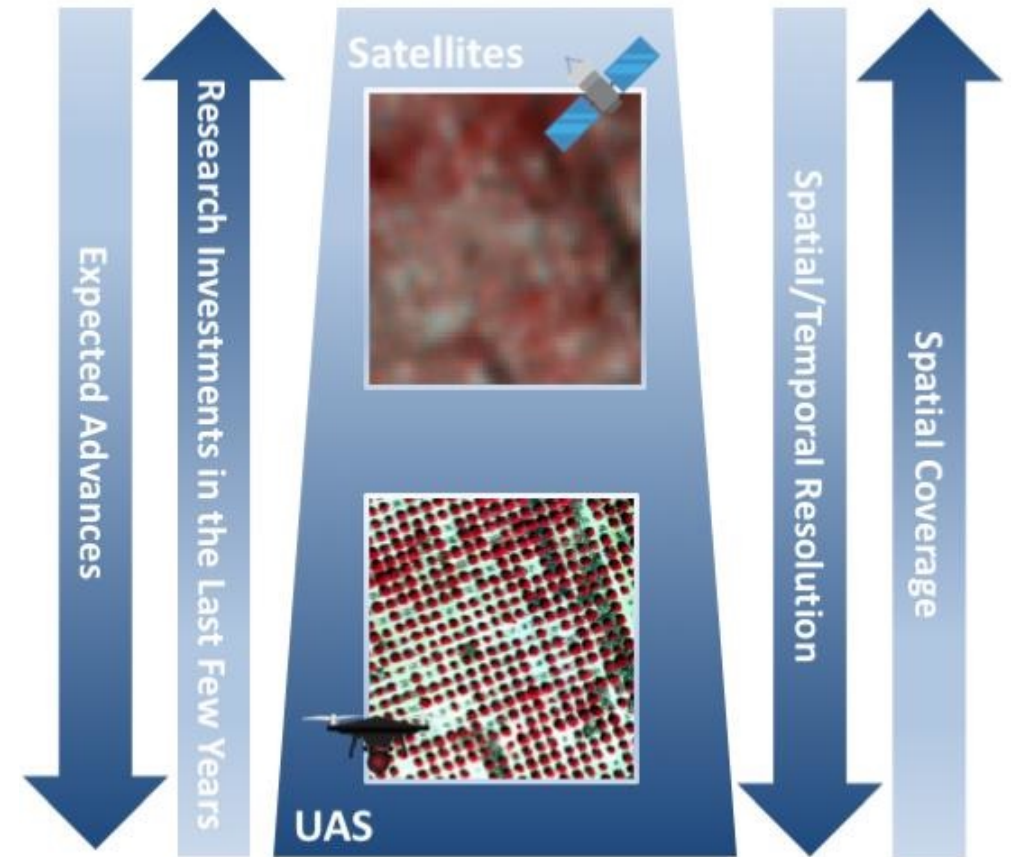
# Environmental monitoring at a glance

3

## Scientific contributions: UAS platform



- ✓ Decline of environmental studies
- ✓ Increase of UASs applications
- ✓ High spatial resolution





# Environmental monitoring at a glance

4

## Which term to use for drones

- Drone
- Unmanned Aerial Vehicles (UAVs)
- Unmanned Aircraft Systems (UAS)
- Unoccupied Aircraft Systems
- Uncrewed Aircraft Systems
- Remotely Piloted Aircraft Systems (RPAS)
- ...

Multirotor <2 Kg



Fixed – wing <2 Kg (micro)



Multirotor 4-6 Kg





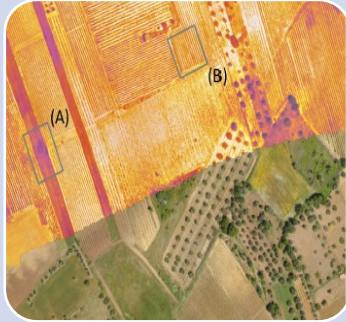
# UAS applications for Environmental monitoring

5

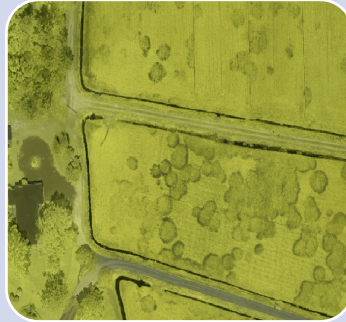
Some (of many!) applications...



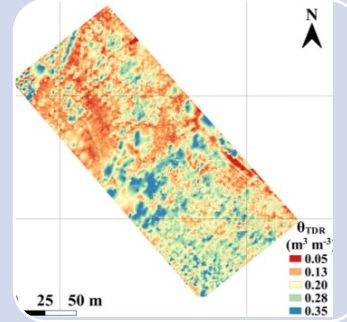
DSMs



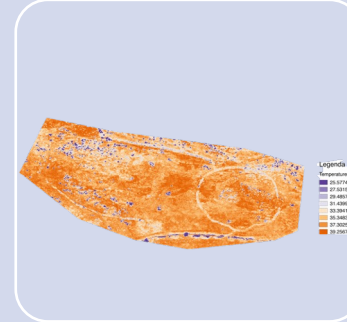
Crop water stress detection



Vegetation status



Soil water content



Archaeological surveys



fluvial monitoring



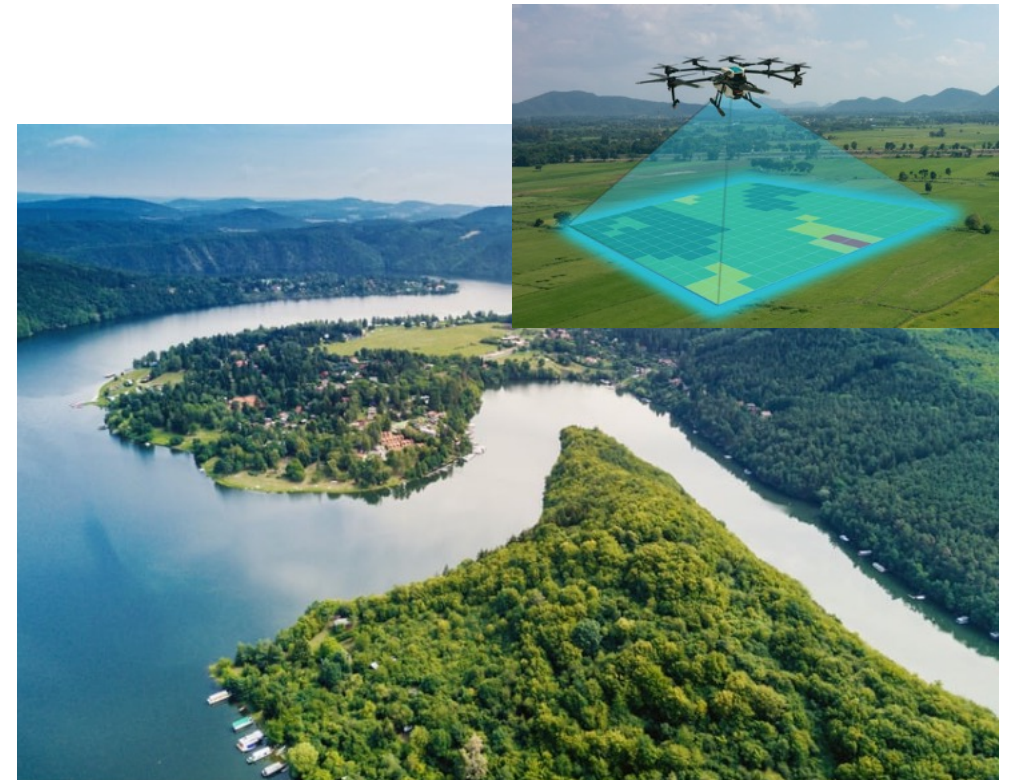
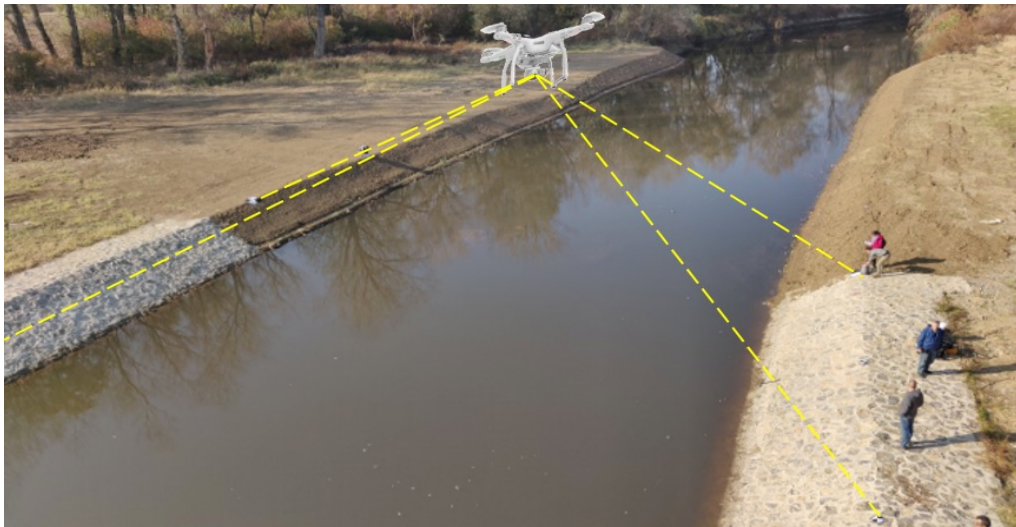
# Outline

6

1. Seeding quantification for image velocimetry applications

2. Footage stabilisation for image velocimetry applications

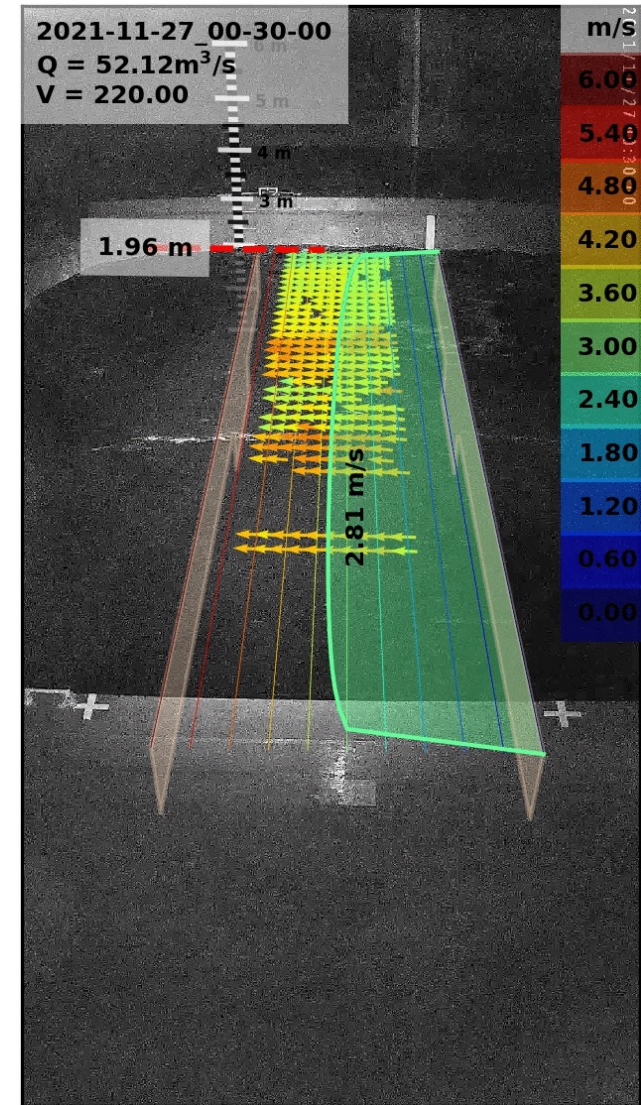
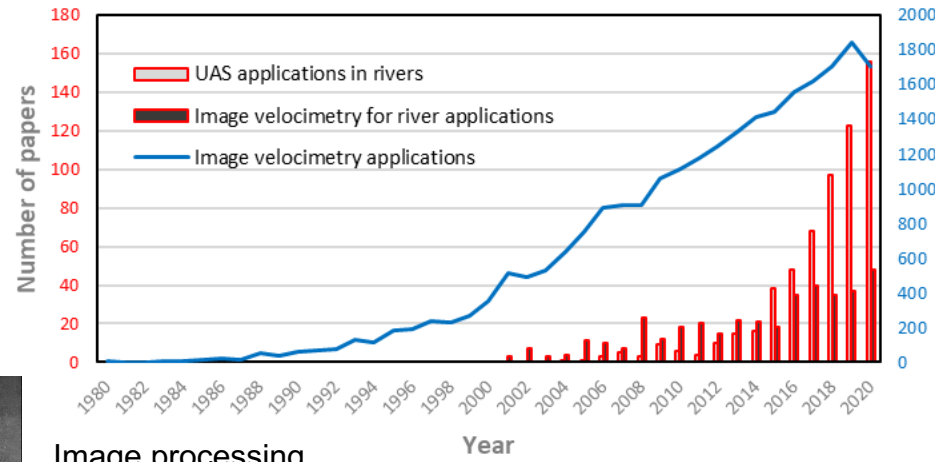
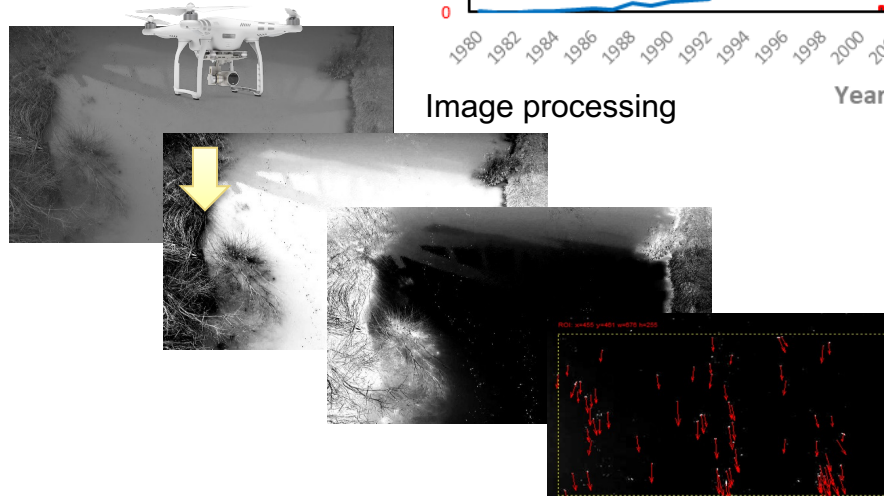
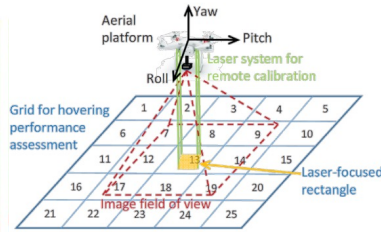
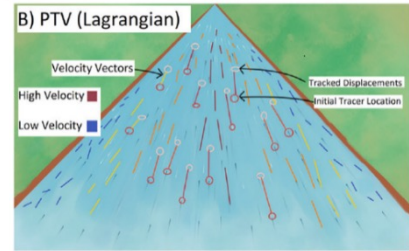
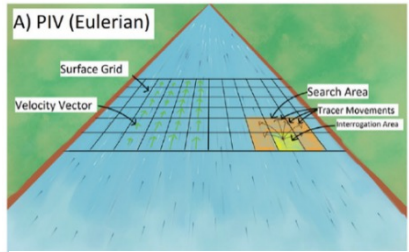
3. Water extent segmentation





# Streamflow monitoring

7



✓ **Scope:** Streamflow monitoring: velocities, discharge, and water levels.

✓ **Versatility:** smartphones, fixed cameras, UASs.

<http://www.photrack.ch/dischargekeeper.html>

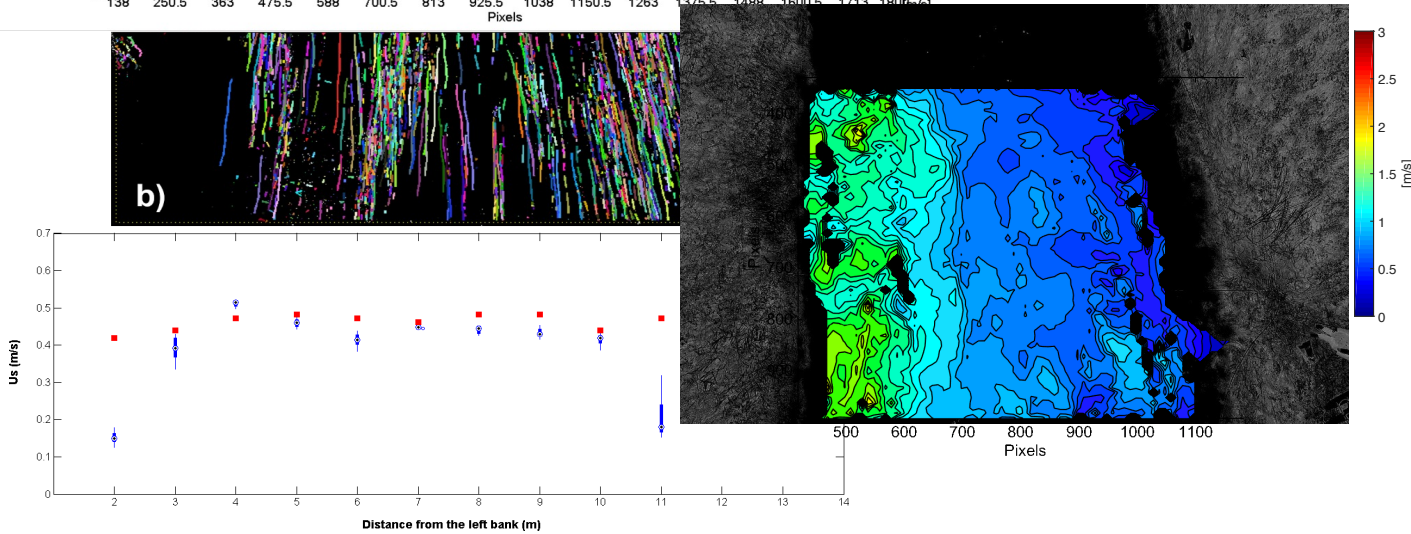
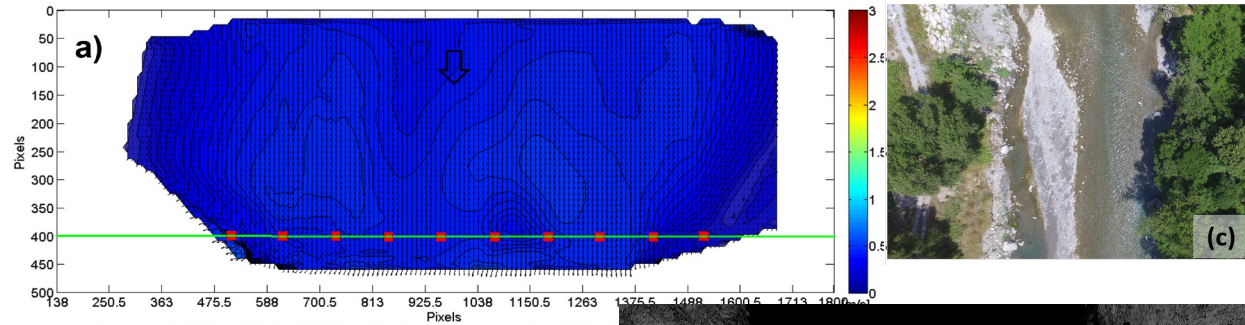




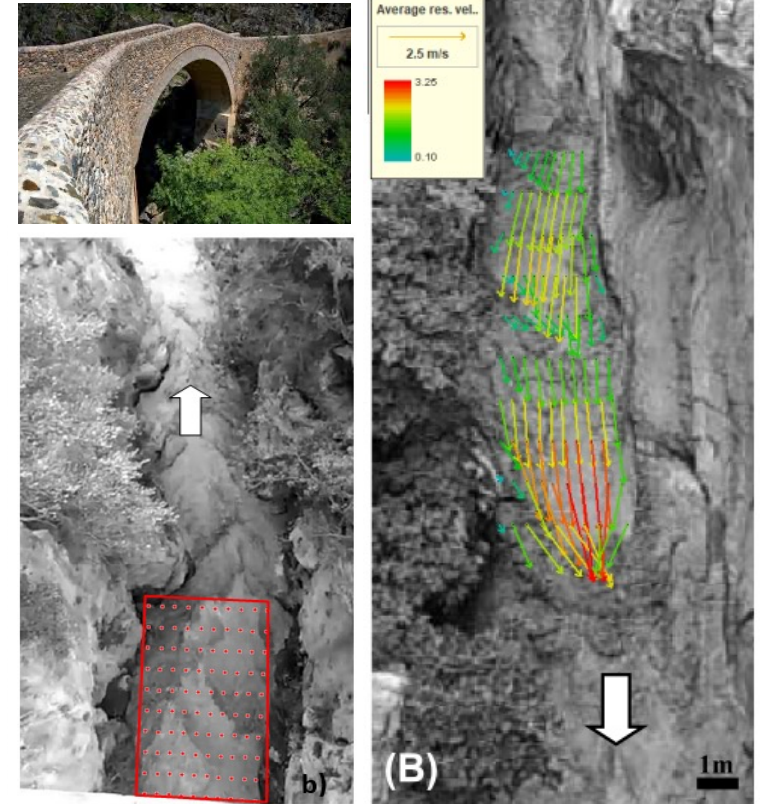
# Streamflow monitoring

8

## Field experience & Analysis

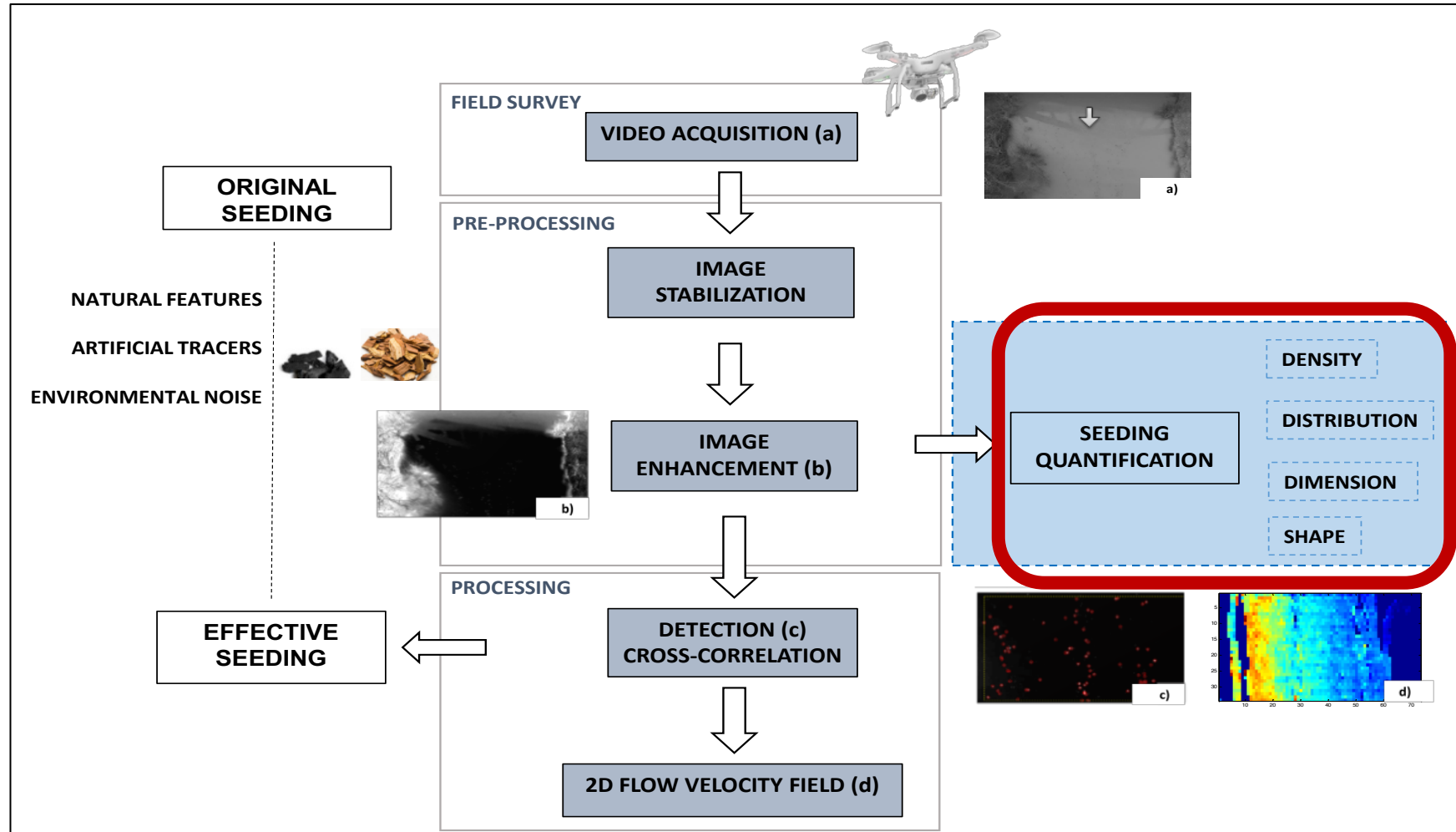


Torrente Raganello (Civita, Italy)



# Streamflow monitoring

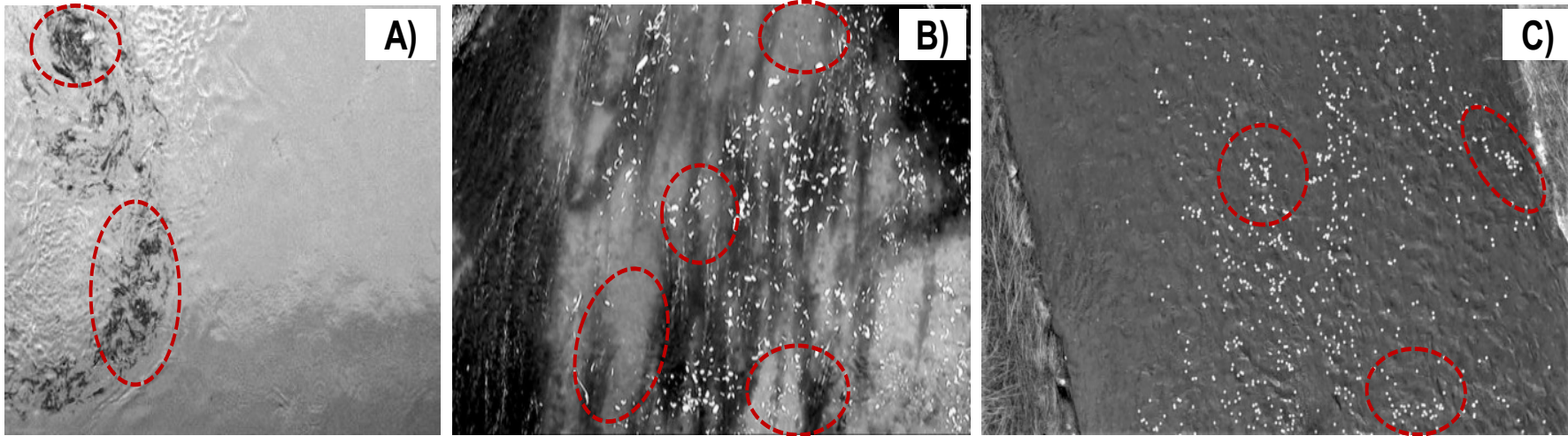
## 9 Workflow for image-velocimetry analysis



# Seeding Quantification

10

Image-velocimetry techniques are widely used, but How are their **accuracies at field conditions?**



Examples of moving and aggregated structures on the water surface: **A)** Natural seeding; **B)** and **C)** Artificial seeding at low/intermediate flow conditions.

- a) Field conditions **much more complex** than the Lab
- b) Image-velocimetry detects and tracks **features**
- c) Should we take a look at **features/tracers dynamics** to **optimise results?**



**Numerical  
approach**



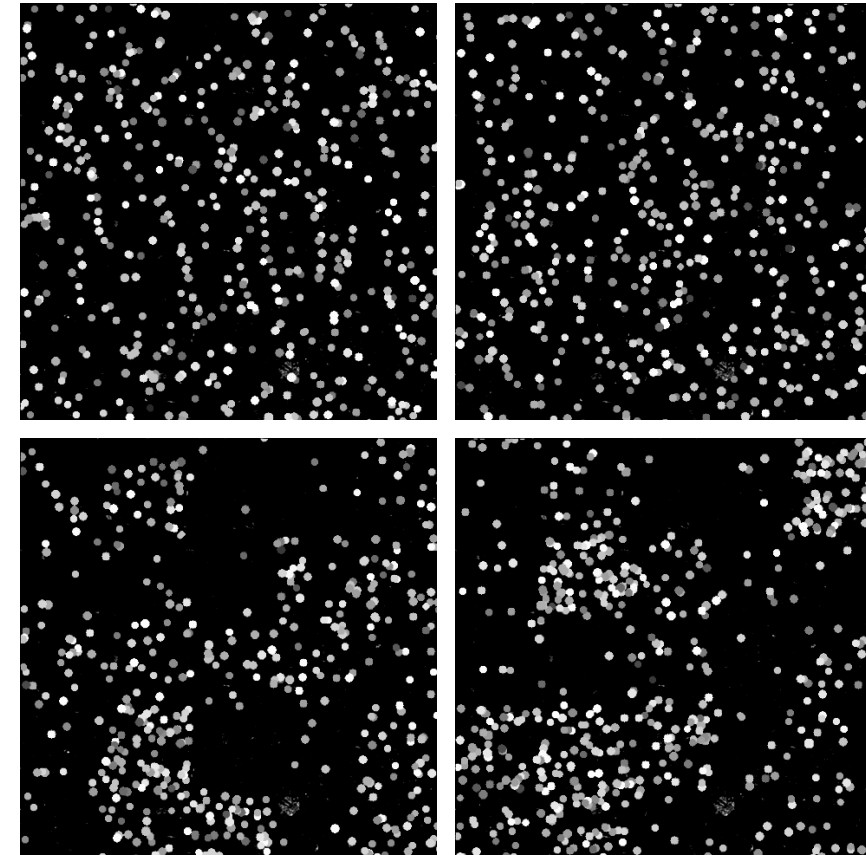
# Seeding Quantification

11

## Numerical framework: Synthetic generation

- Synthetic tracers were randomly distributed in space with a unidirectional and constant velocity (15 px/frame).
- They consist of uniform circular shapes with diameter  $D_{xp} \approx 10$  px and uniform white colour.
- Both diameters and colours were altered with white noise in order to consider more realistic configurations.
- Their spatial distribution was controlled by a Generalised Poisson Distribution (GPD) with a theoretical seeding density  $\lambda$  and level of aggregation  $\nu$ .
- The quality of the results was determined by the magnitude of the errors that were computed as

$$\epsilon = 100 \times \frac{(u_c - u_R)}{u_R}$$

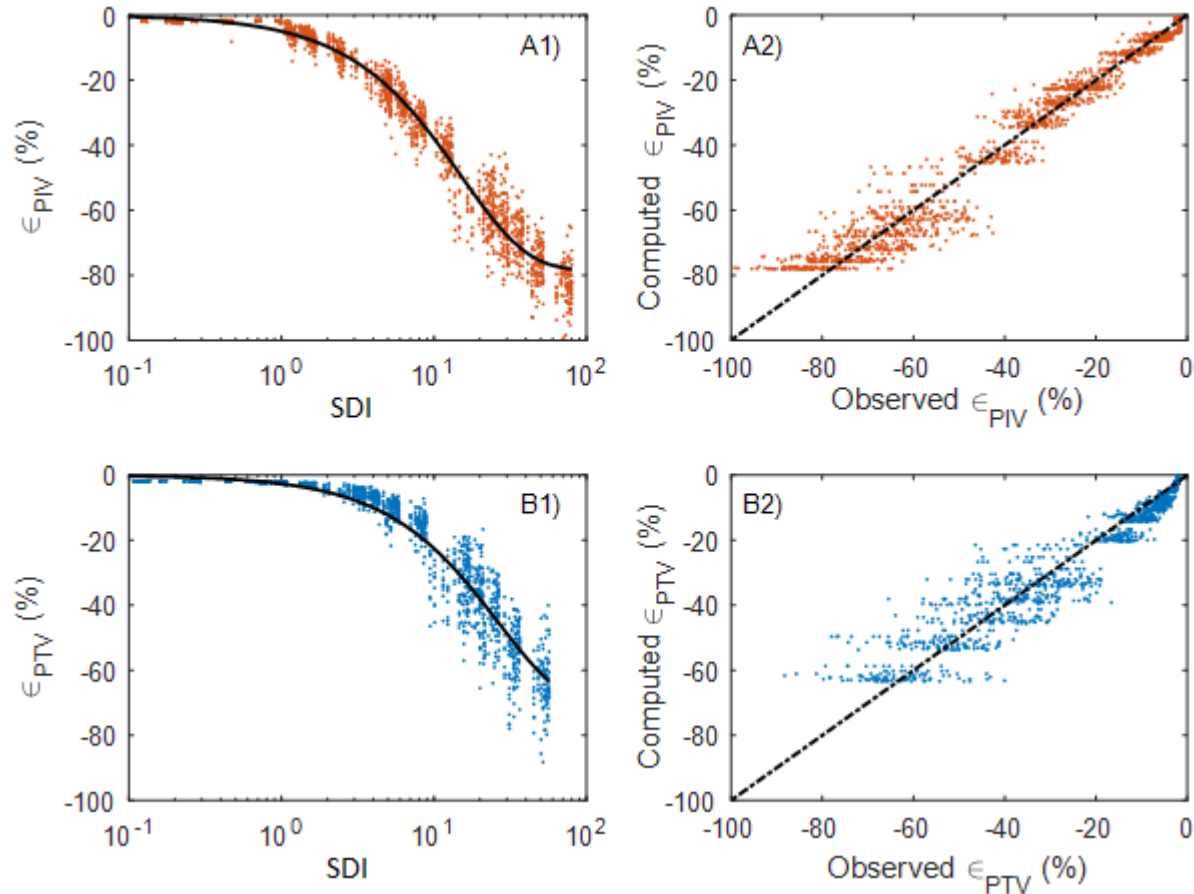


Numerical simulations of synthetically generated particles that present different aggregation levels: **33,600** images generated in total

# Seeding Quantification

12

## Numerical framework: Analysis



**Fig.** Numerical simulations of synthetically generated particles that present different aggregation levels.

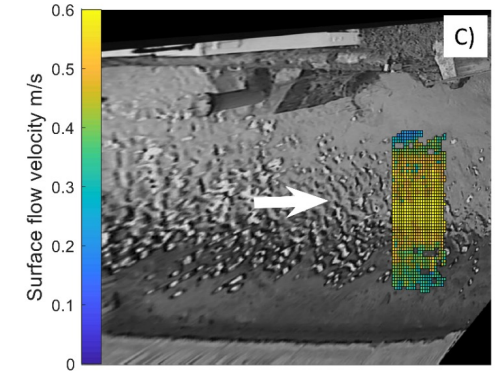
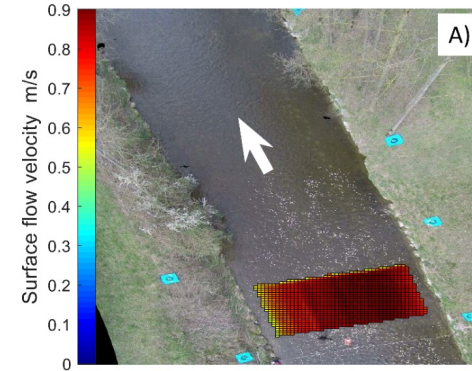
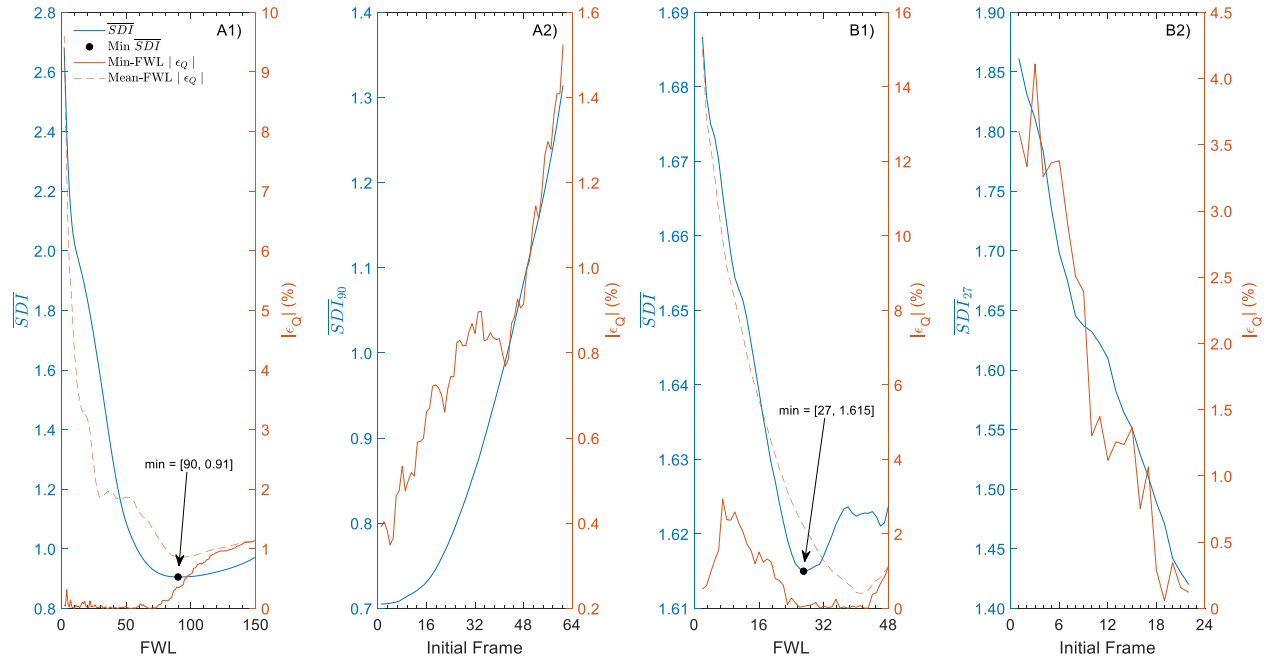
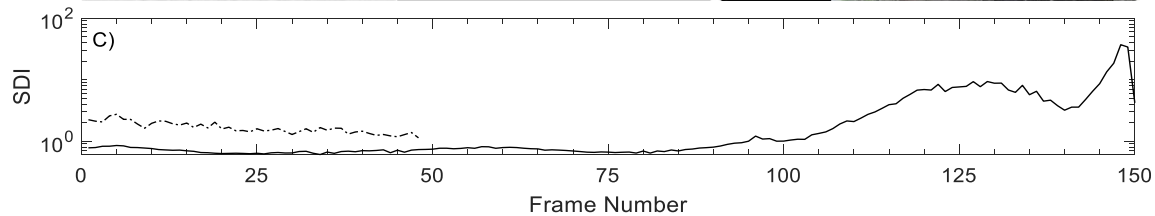
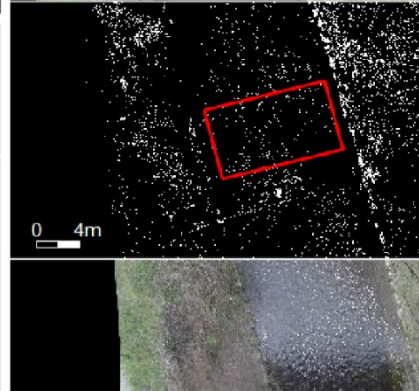
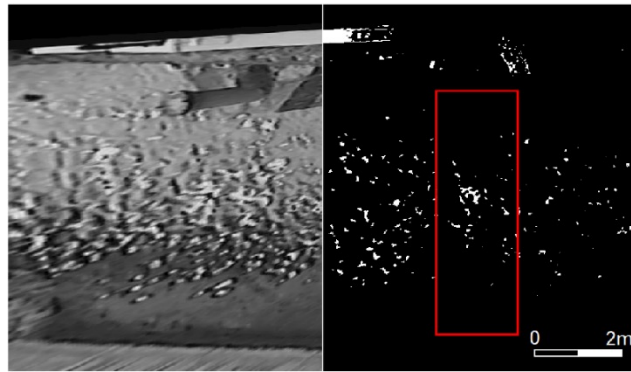
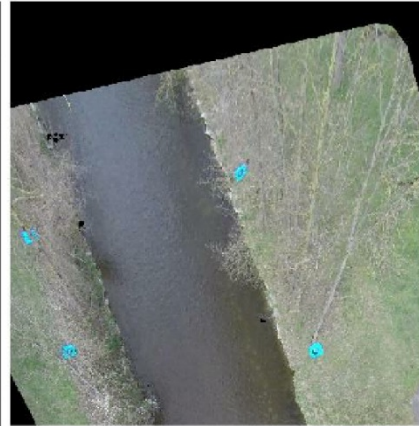
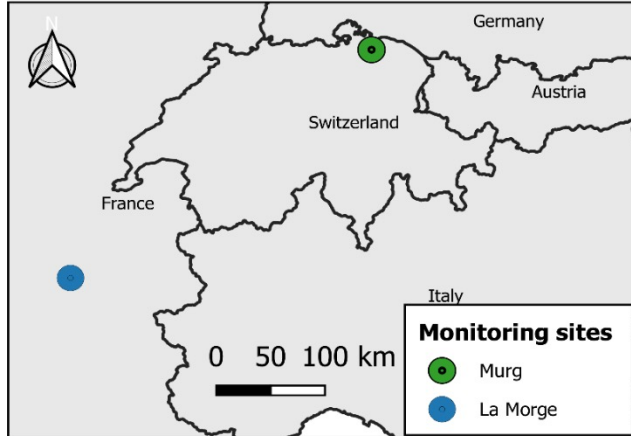
$$SDI = D^{*0.1} / \left( \frac{\rho}{\rho_c D^* 1} \right)$$

- ✓  $\rho$  := Seeding density
- ✓  $D^*$  := Dispersion Index
- ✓ Clearly visible relationship between Errors and SDI.
- ✓ The lower SDI, the lower the Errors.

# SDI at field conditions

13

## SDI applications: 1.0

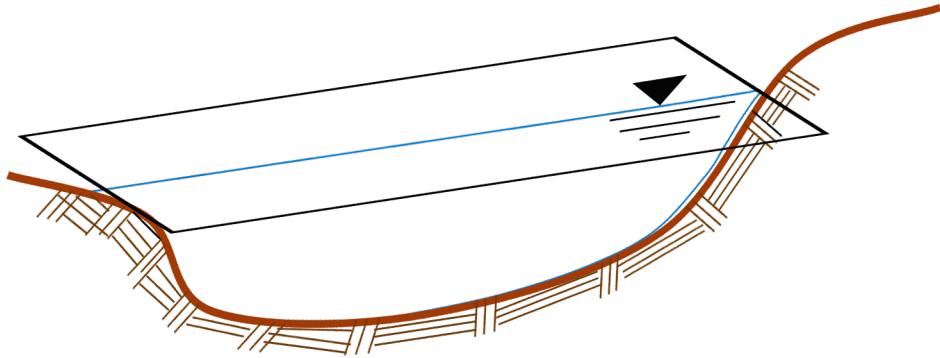




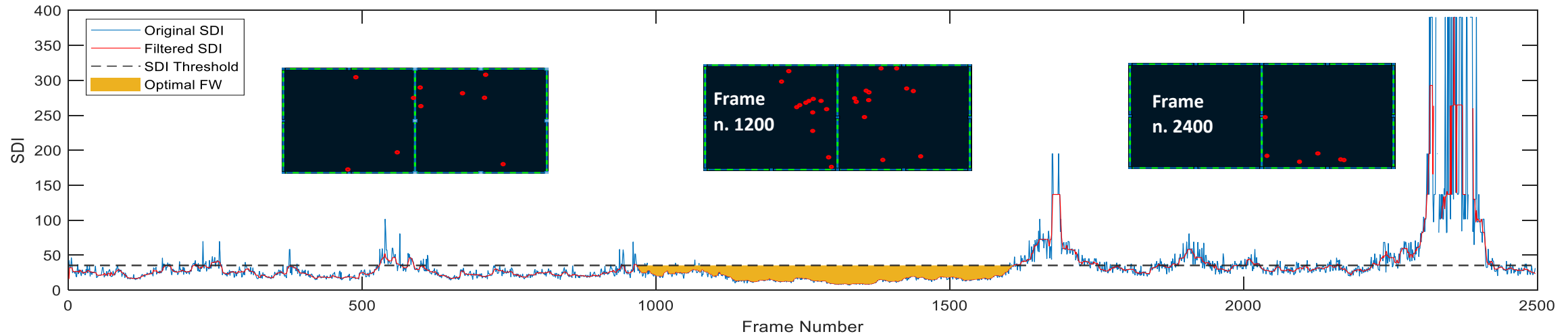
# SDI at field conditions

14

SDI applications: 2.0



- ✓ Optimal FW approach based on SDI.
- ✓ Different spatial scales.
- ✓ Errors reduction of  $\sim 20-40\%$



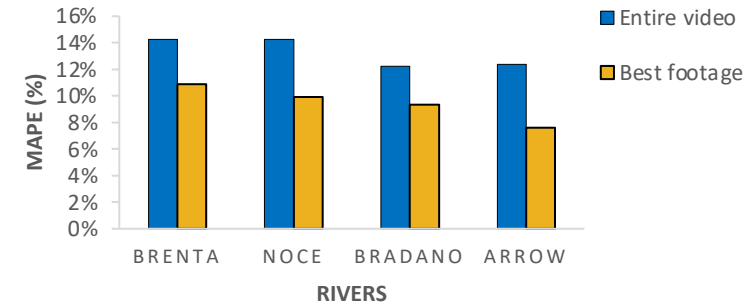
# SDI at field conditions

15

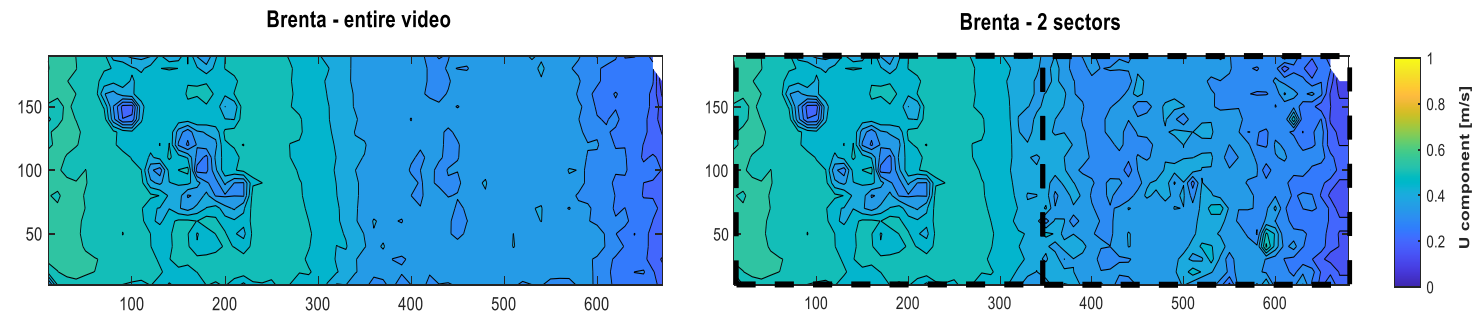
## SDI applications: 2.0

- The SDI-based method improved LSPIV performances with a reduction of image velocimetry errors at sector and sub-sector scales

River	Number of frames	
	Entire video	Optimal range
Brenta	2500	153
Noce	200	70
Bradano	2496	642
Arrow	799	282



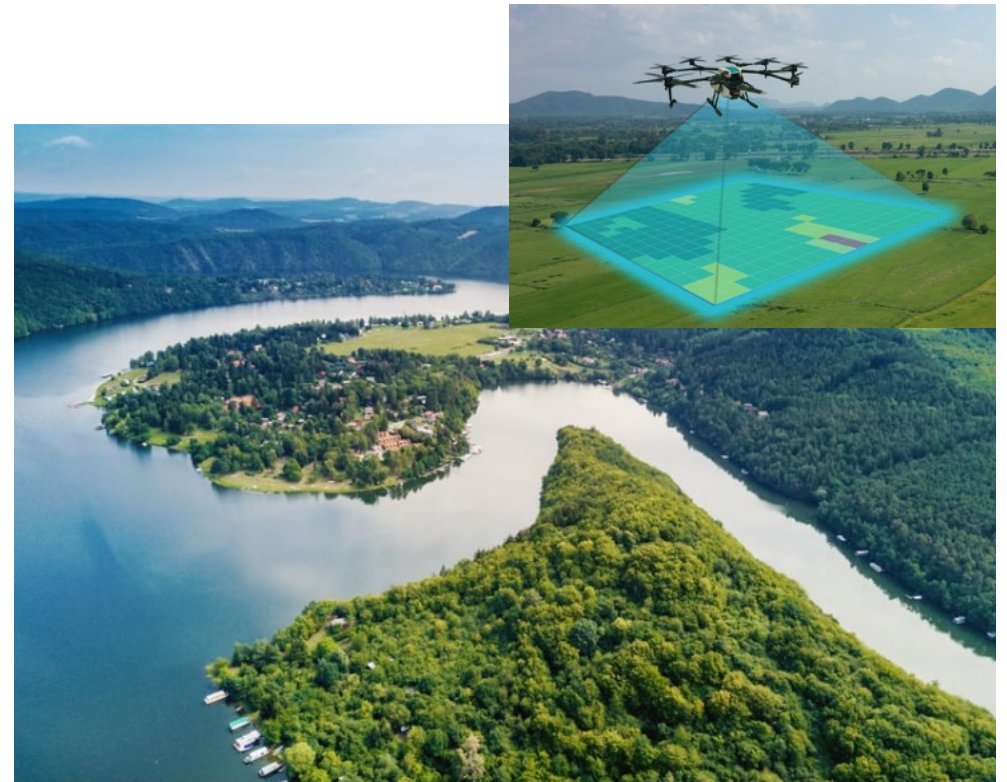
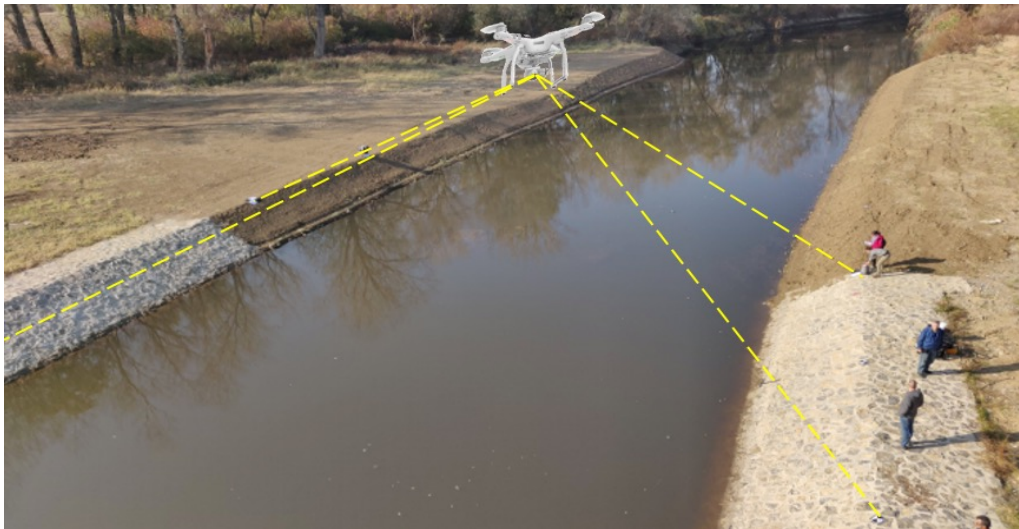
- In such cases, the average surface velocity maps contain details (e.g., velocity fluctuations and divergences) that are not visible and appreciable in the entire video configuration (standard approach).



# Outline

16

1. Seeding quantification for image velocimetry applications
2. Footage stabilisation for image velocimetry applications
3. Water extent segmentation

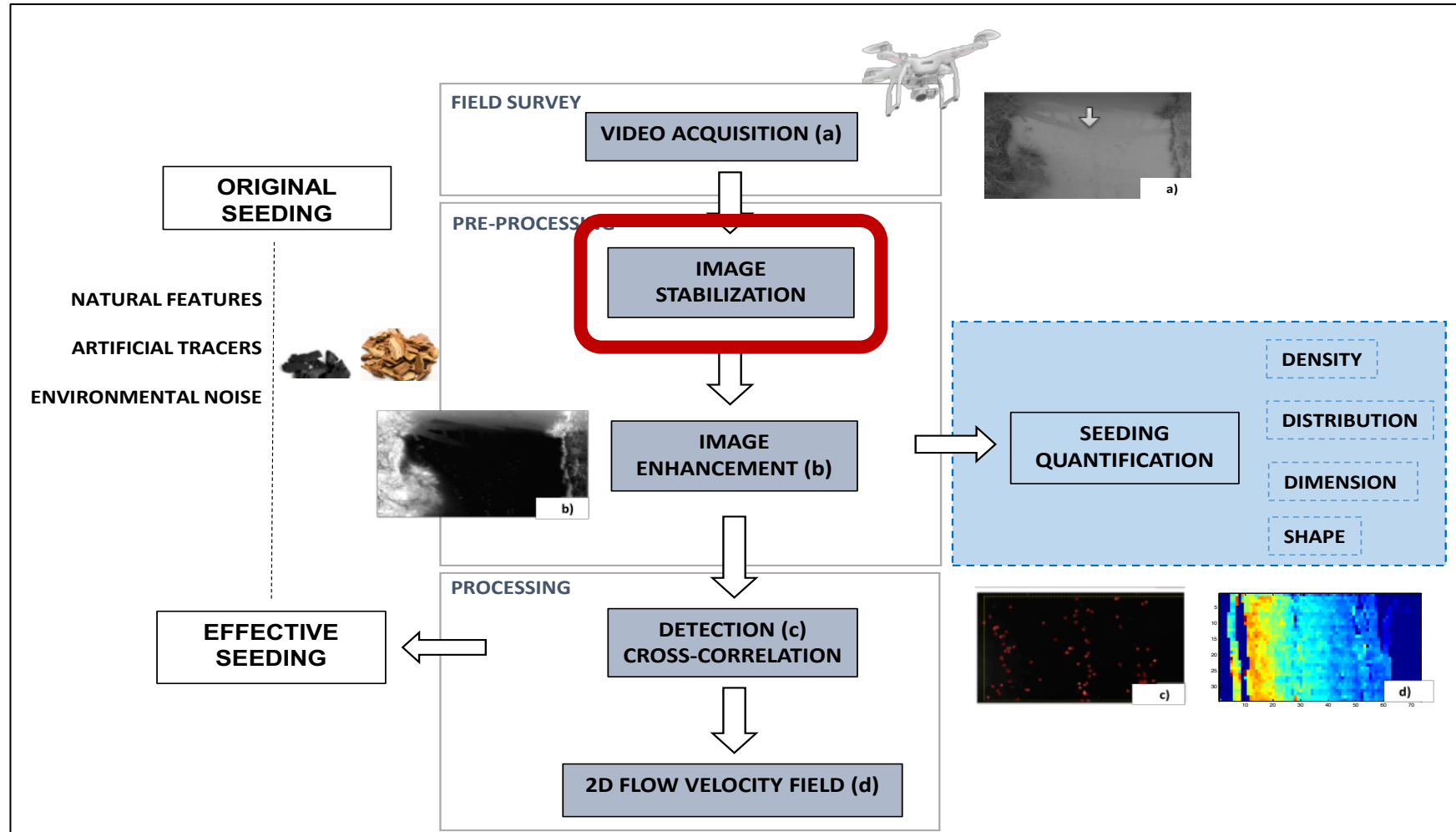




# Streamflow monitoring

17

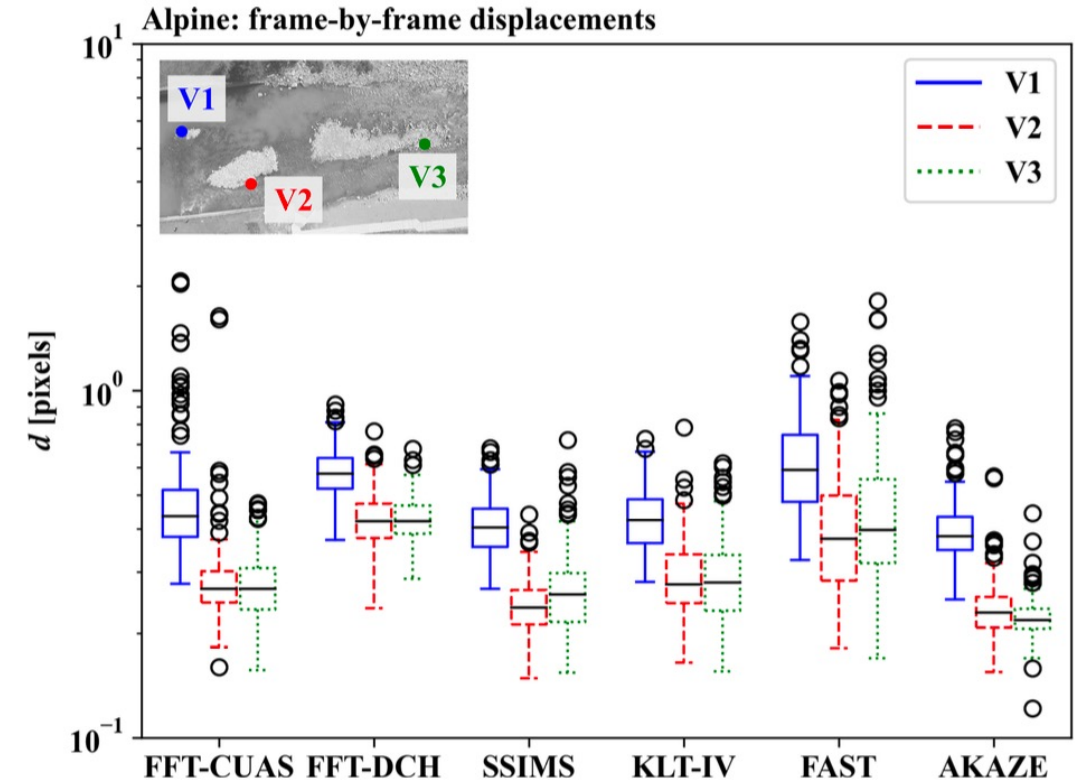
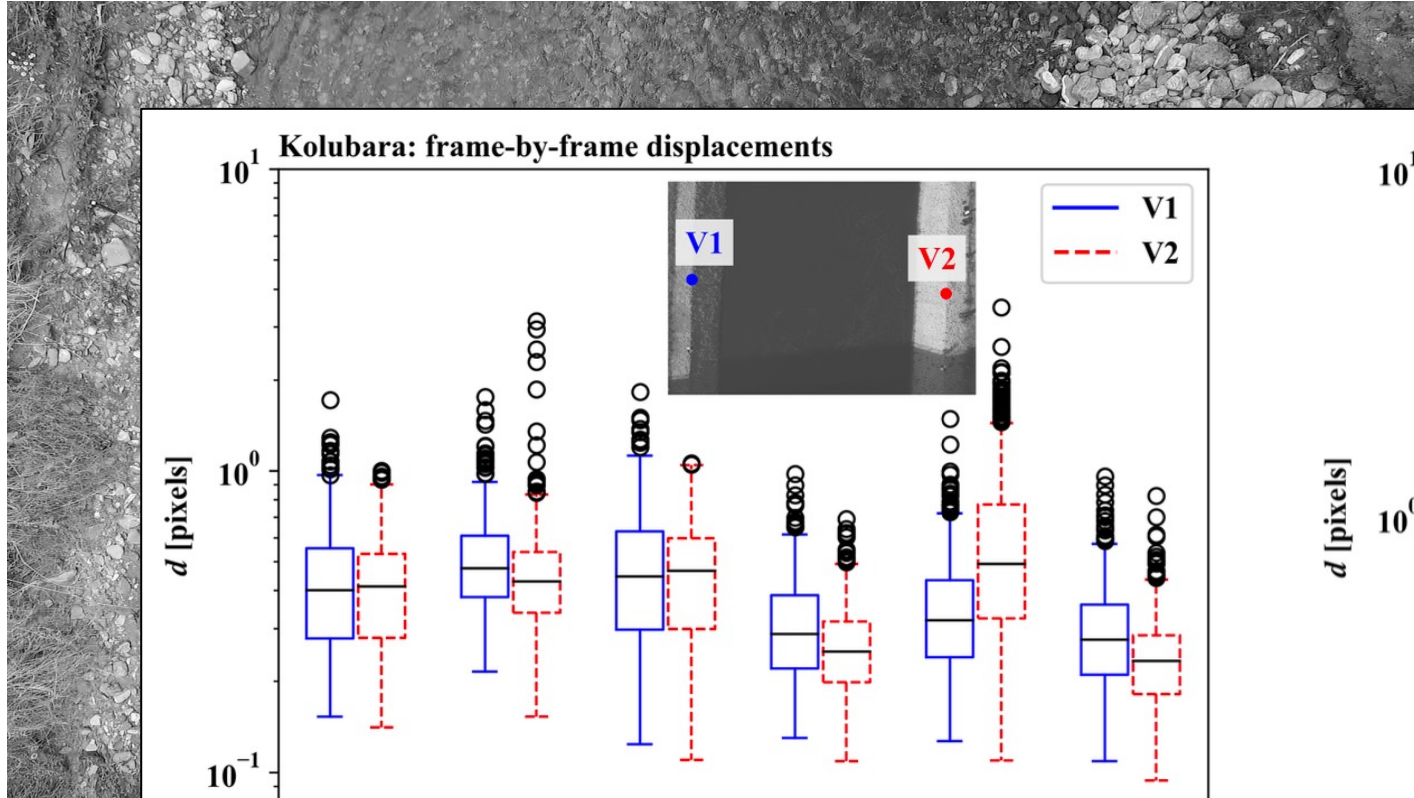
## Workflow for image-velocimetry analysis



# Streamflow monitoring

18

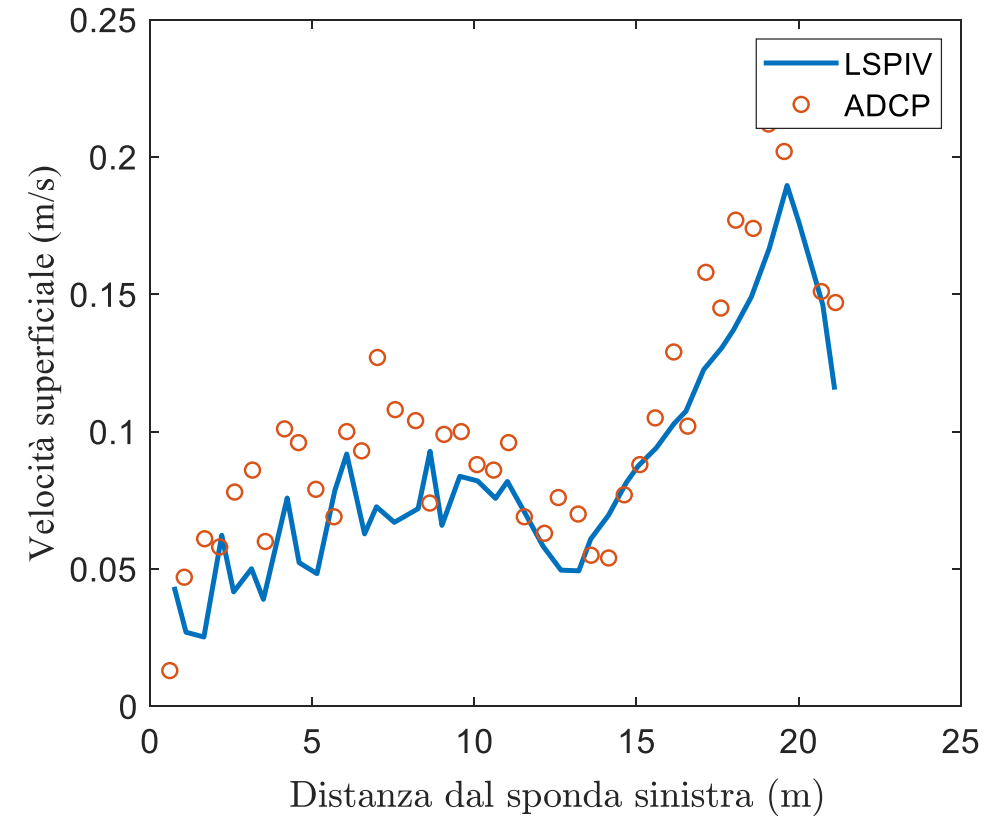
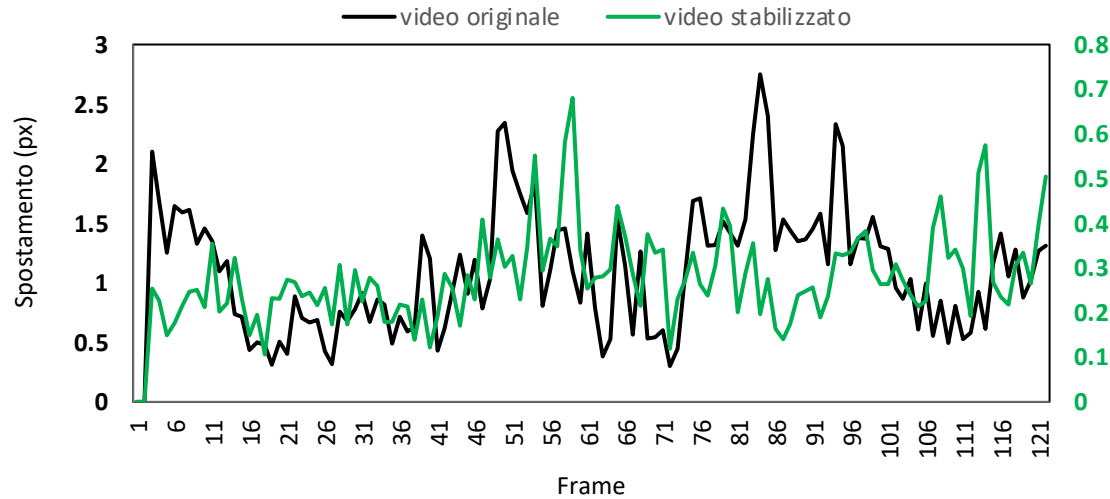
Stabilisation issues



# Stabilisation for image velocimetry

19

**VISION: Video Stabilisation** using automatic features selection for image velocimetry analysis in rivers



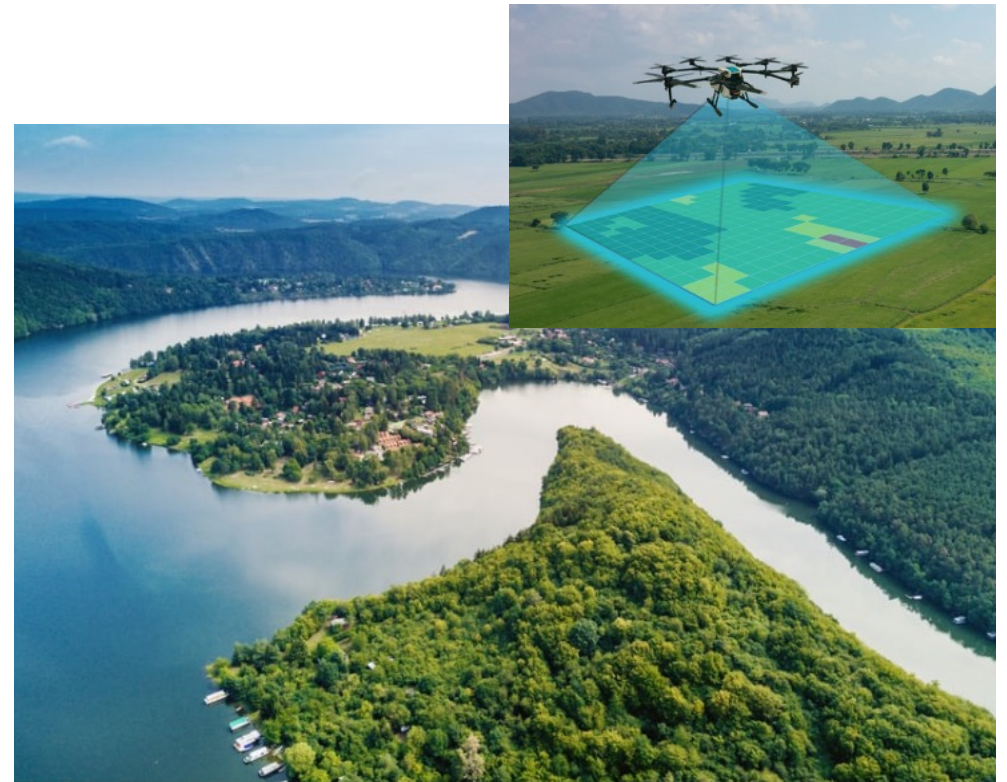
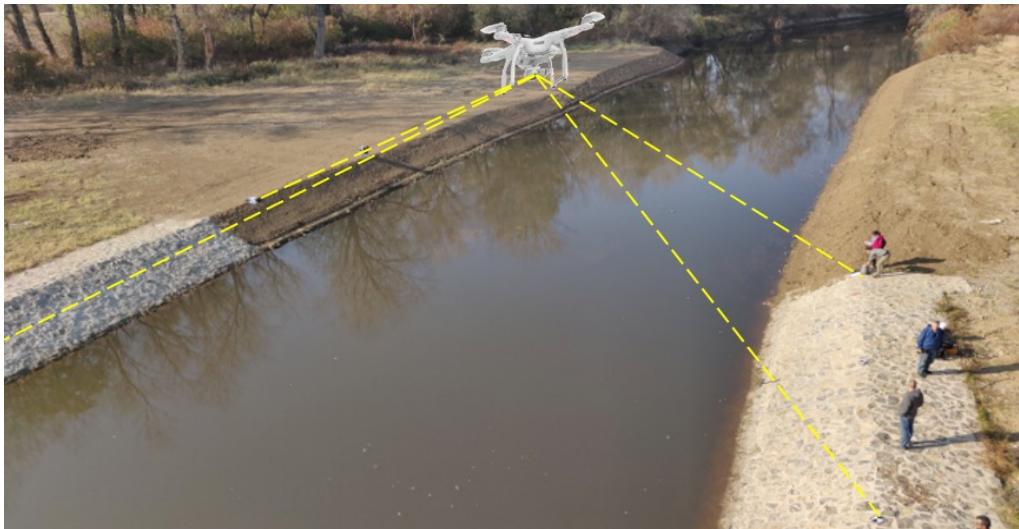
Sezione	Componente di Velocità	RMSE (m/s)		Riduzione errore (%)
		Video non stabilizzato	Video stabilizzato	
S1	v	0.07	0.06	8%
	u	0.02	0.02	-1%
	m	0.07	0.07	7%
S2	v	0.08	0.07	8%
	u	0.03	0.03	6%
	m	0.08	0.07	8%



# Outline

20

1. Seeding quantification for image velocimetry applications
2. Footage stabilisation for image velocimetry applications
3. Water extent segmentation





# Water extent segmentation

21

How do we (computers) see rivers?

**Humans:**

- **Colour**
- **Sound**
- **Water movement**
- **etc**



**Computers:**

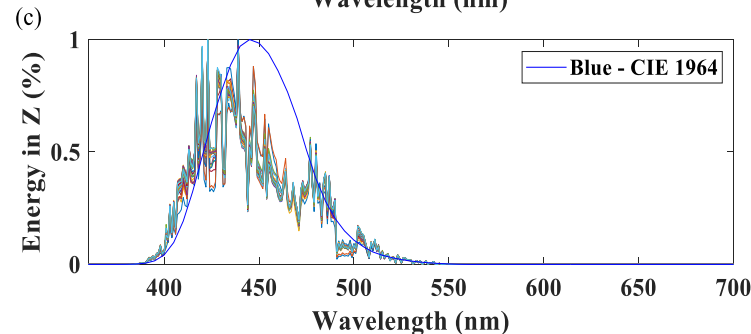
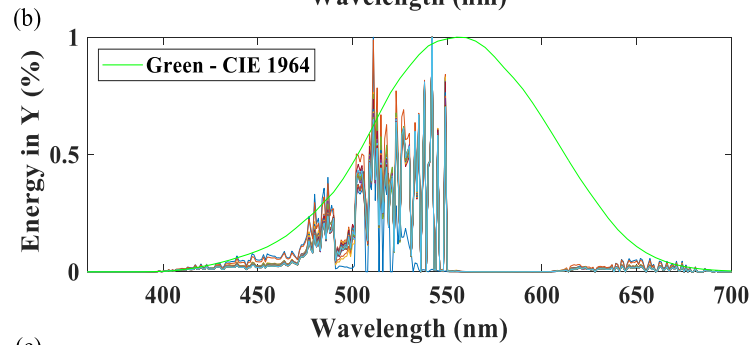
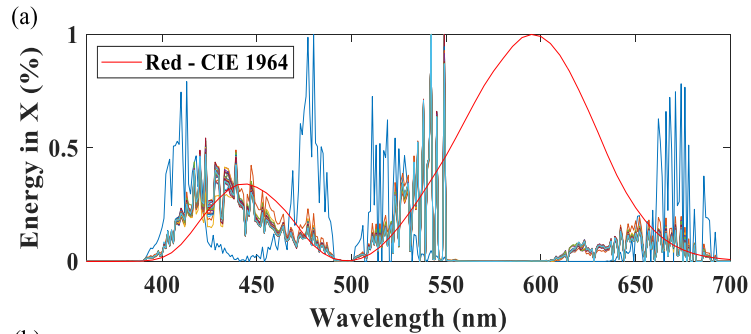
- **Ones & Zeros**



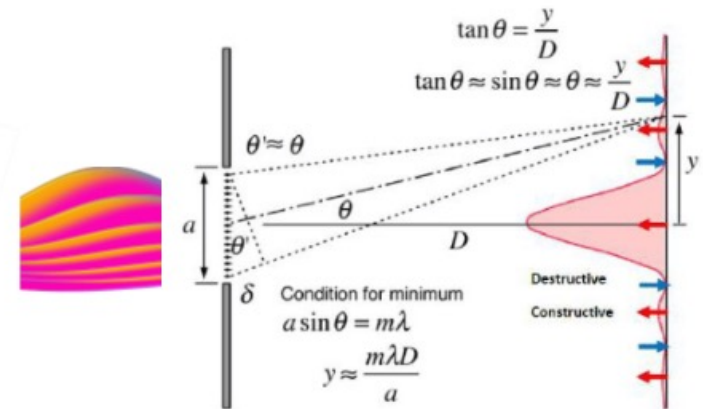
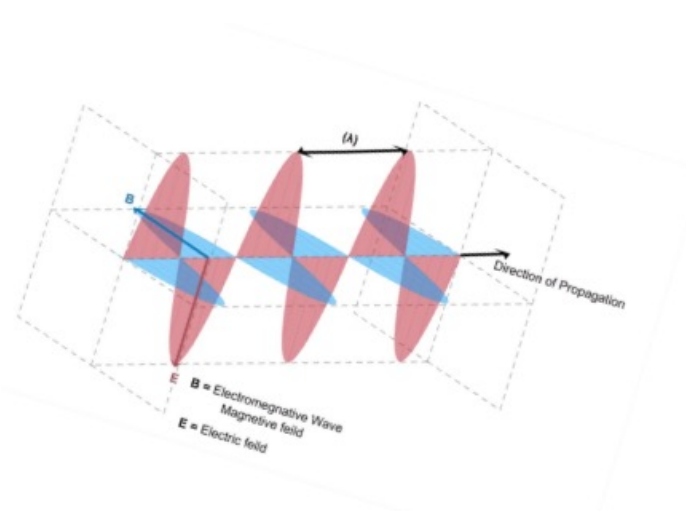
# Water extent segmentation

22

Water extent segmentation by single slit diffraction correction?



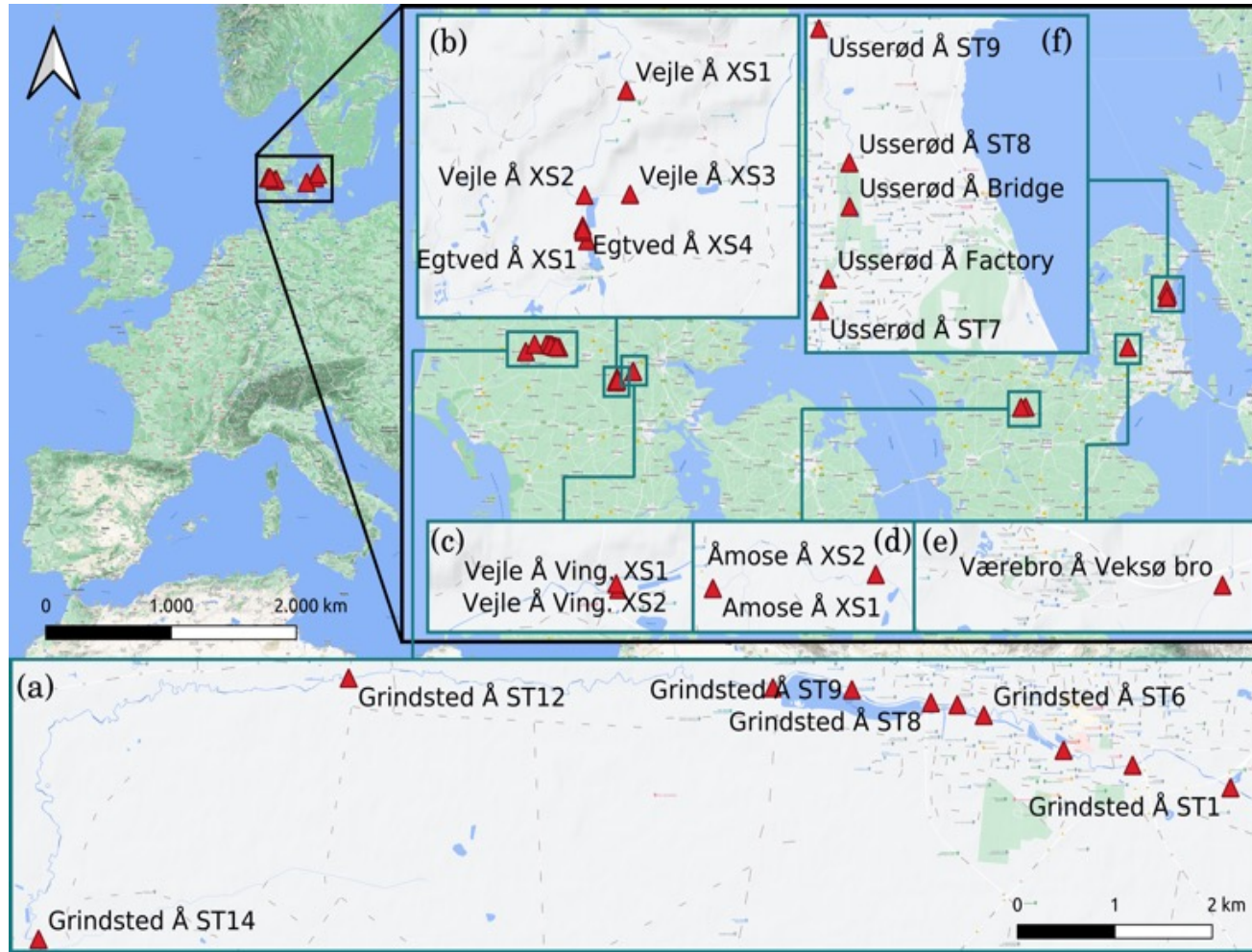
- Cameras are always affected by quantum interference
- CIE standard is a standardised way to represent colours
- What is acquired by the camera can be corrected to follow the CIE standard



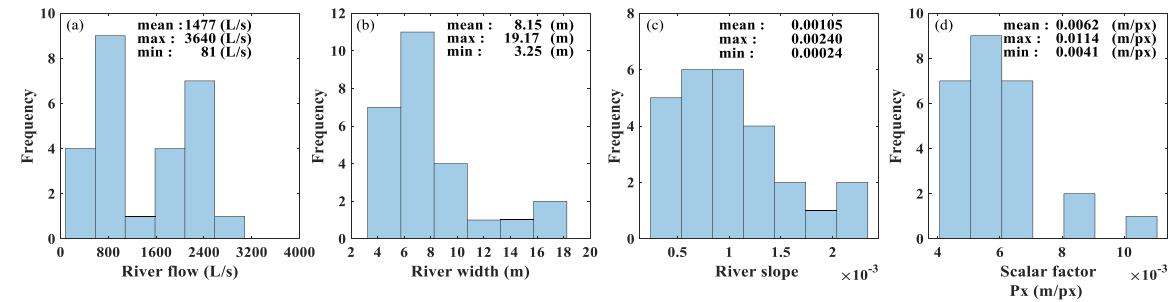
# Water extent segmentation

23

## Case studies to test the model



- 27 case studies located in Denmark
- Huge difference between them, e.g. Q ranges between 81 and 1477 L/s
- Riverbed in some cases has vegetation

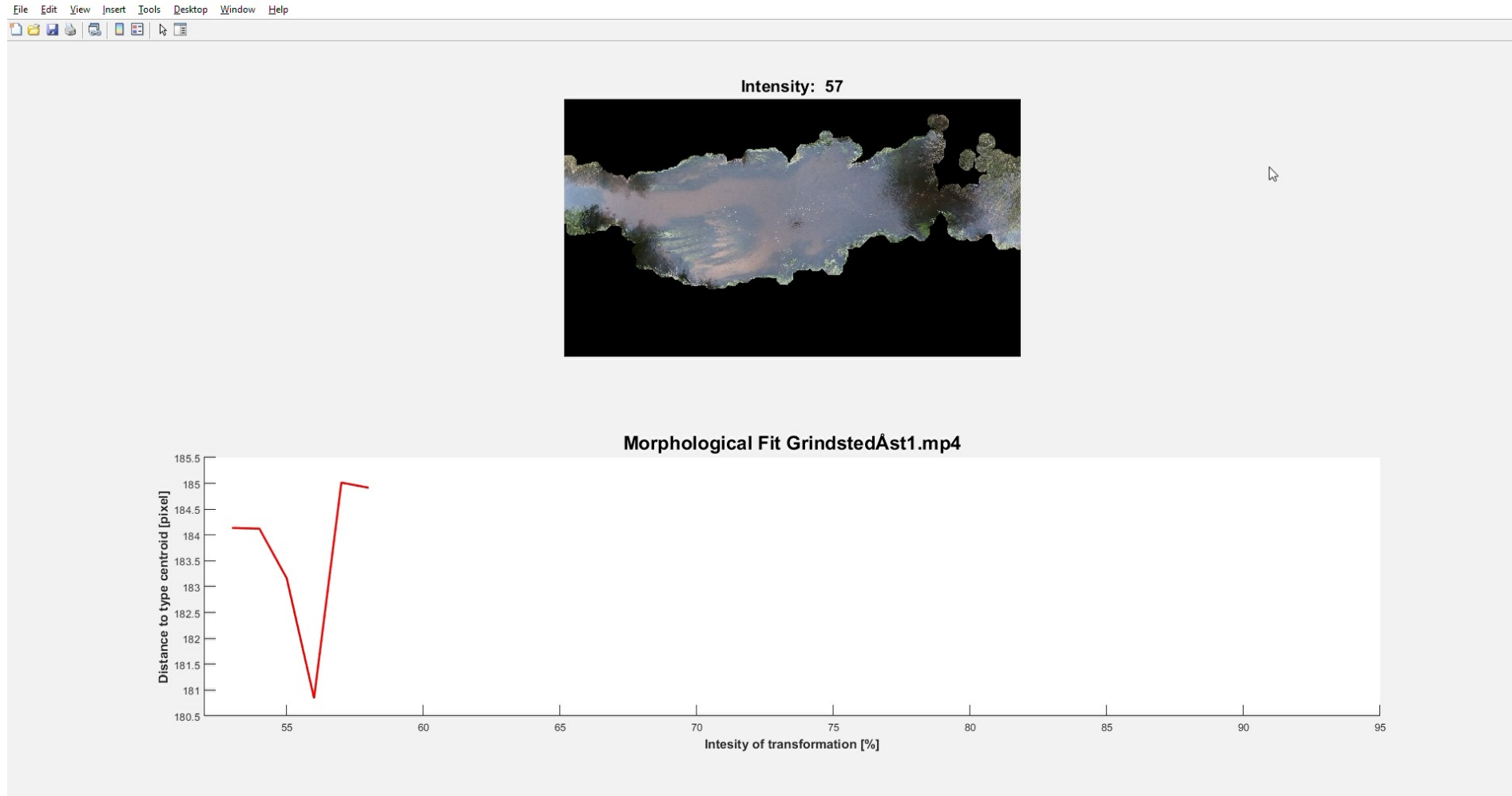




# Water extent segmentation

24

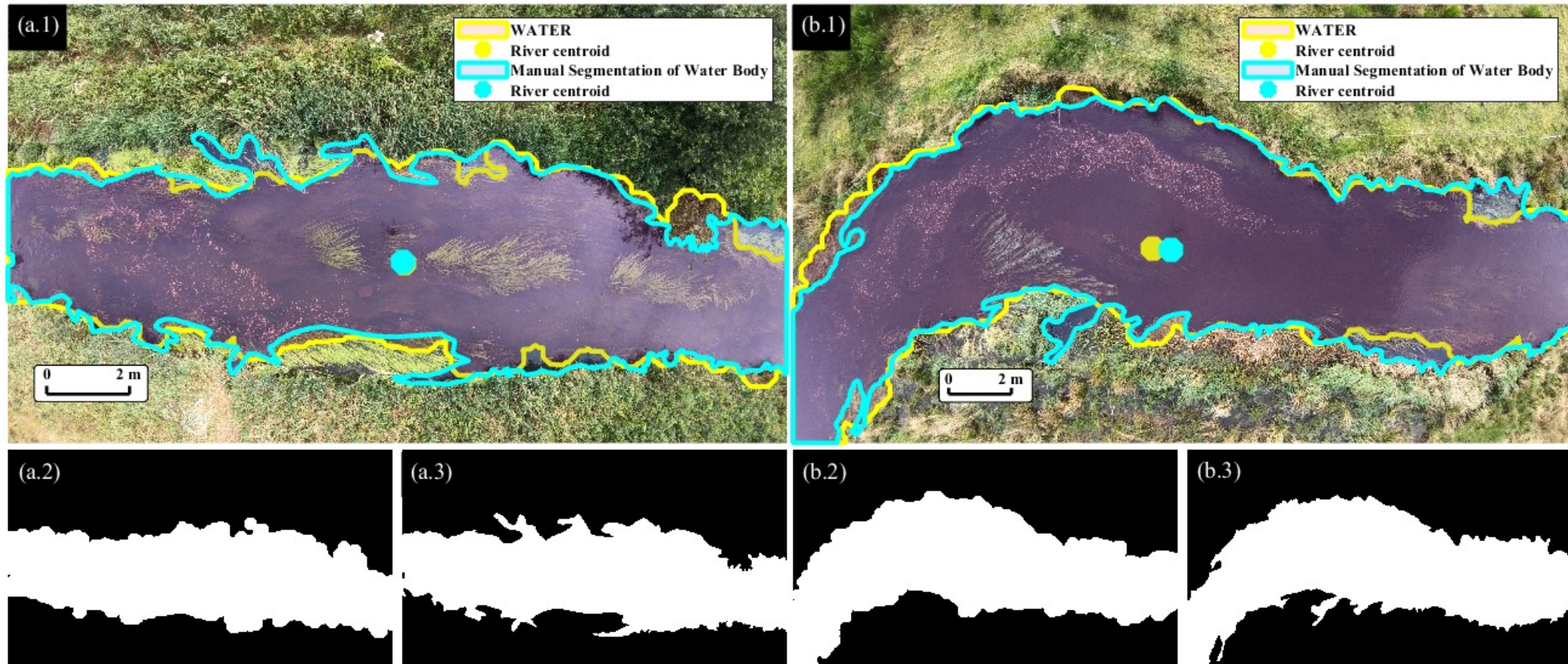
## Water AuTomatic sEgmentation in Rivers (WATER)





# Water extent segmentation

## 25 Human (manually-based) vs Machine (automatic)

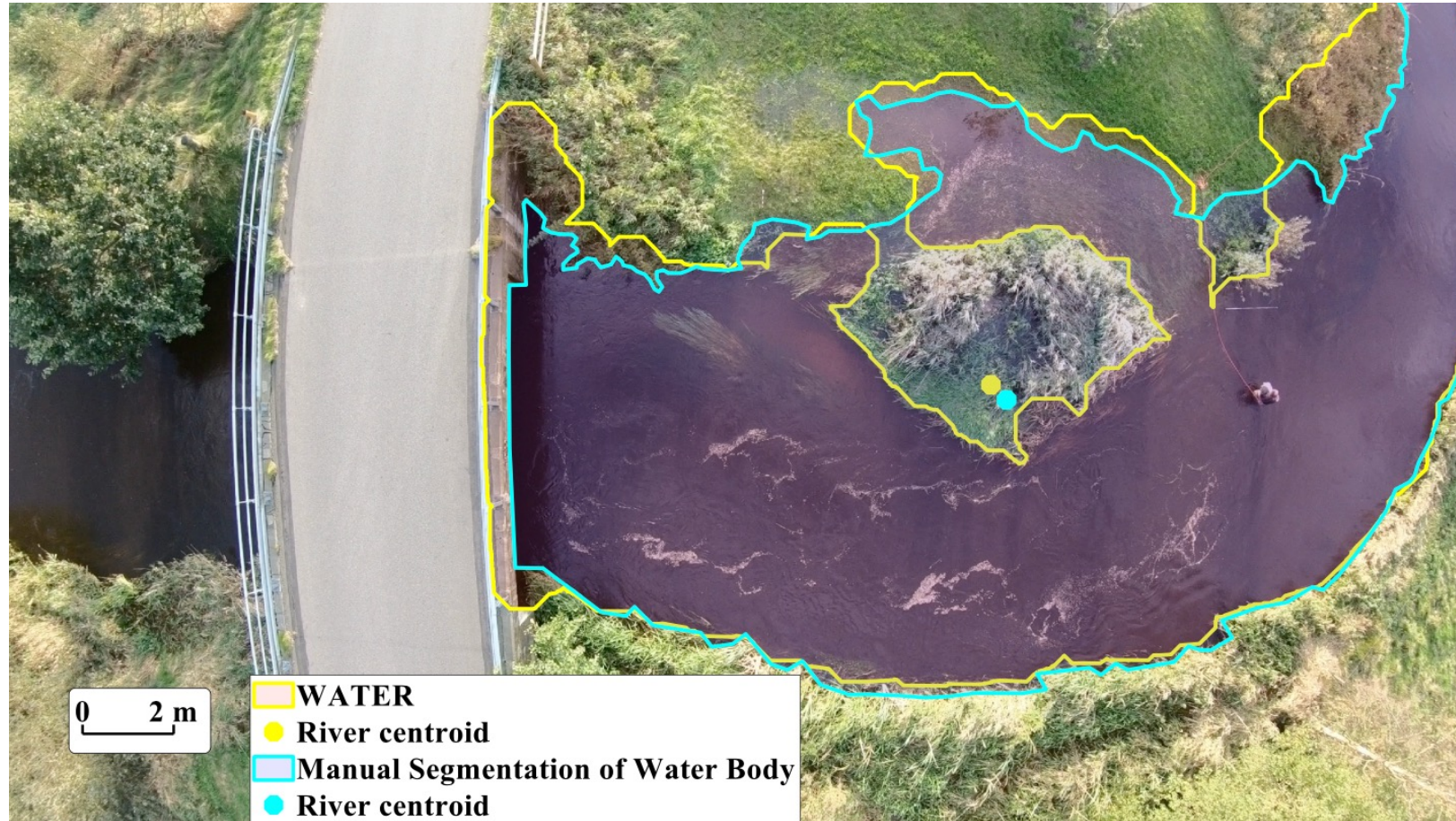




# Water extent segmentation

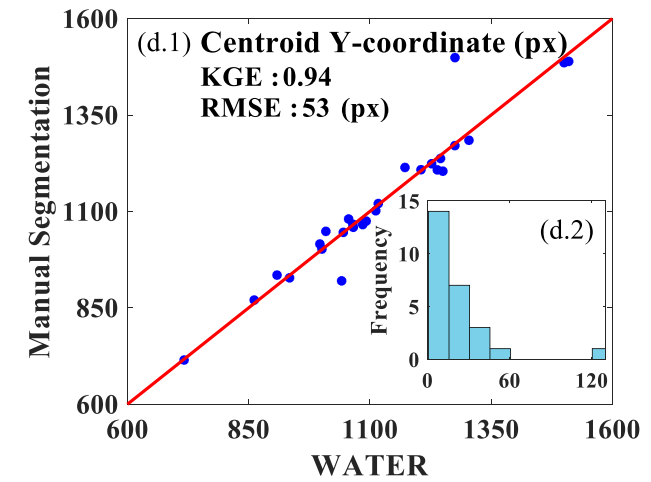
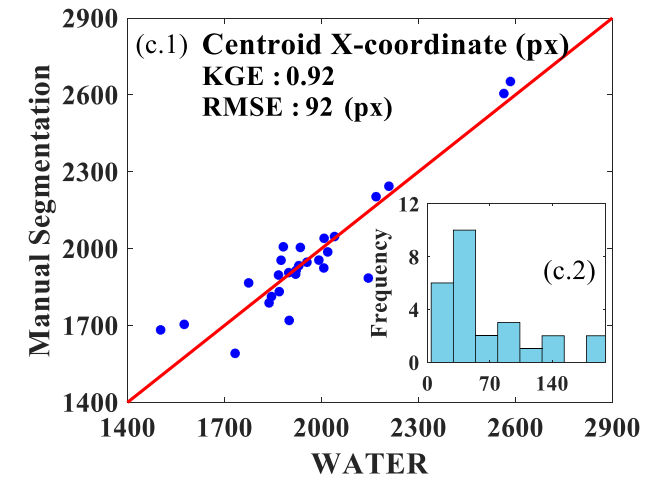
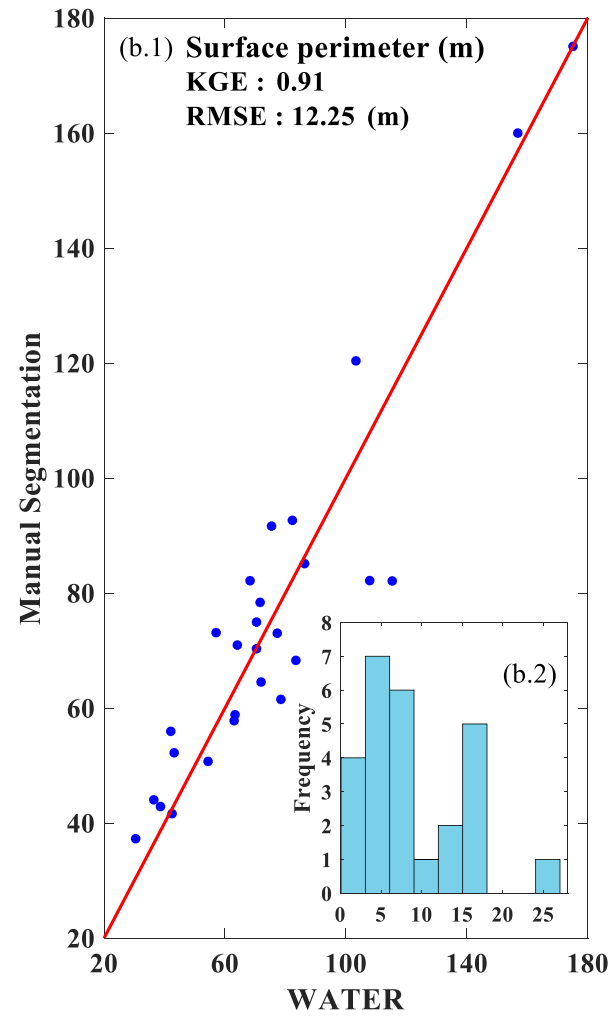
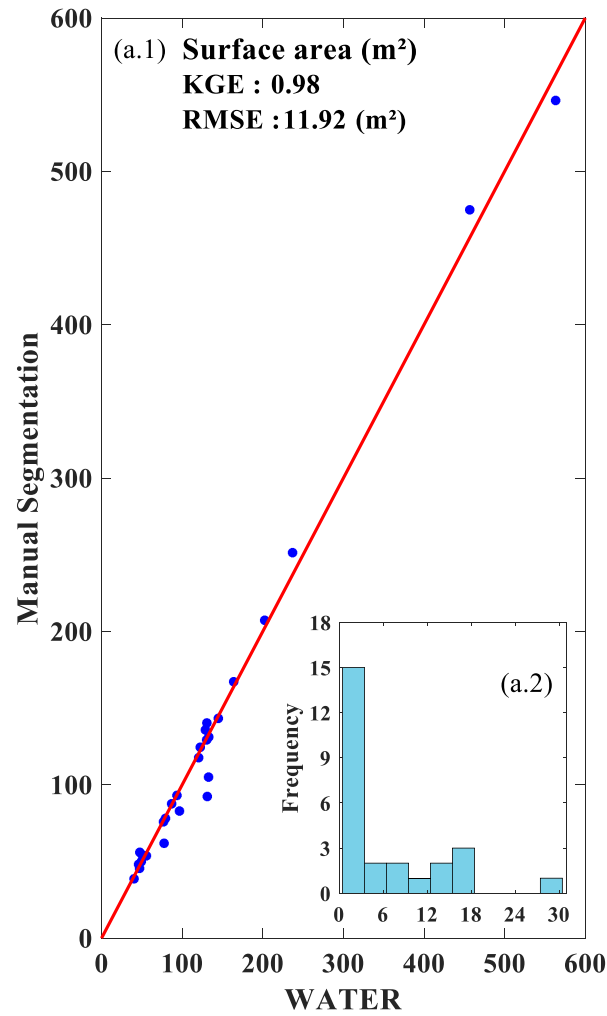
26

What happens when a vegetated bar is in the middle of the river?



# Water extent segmentation

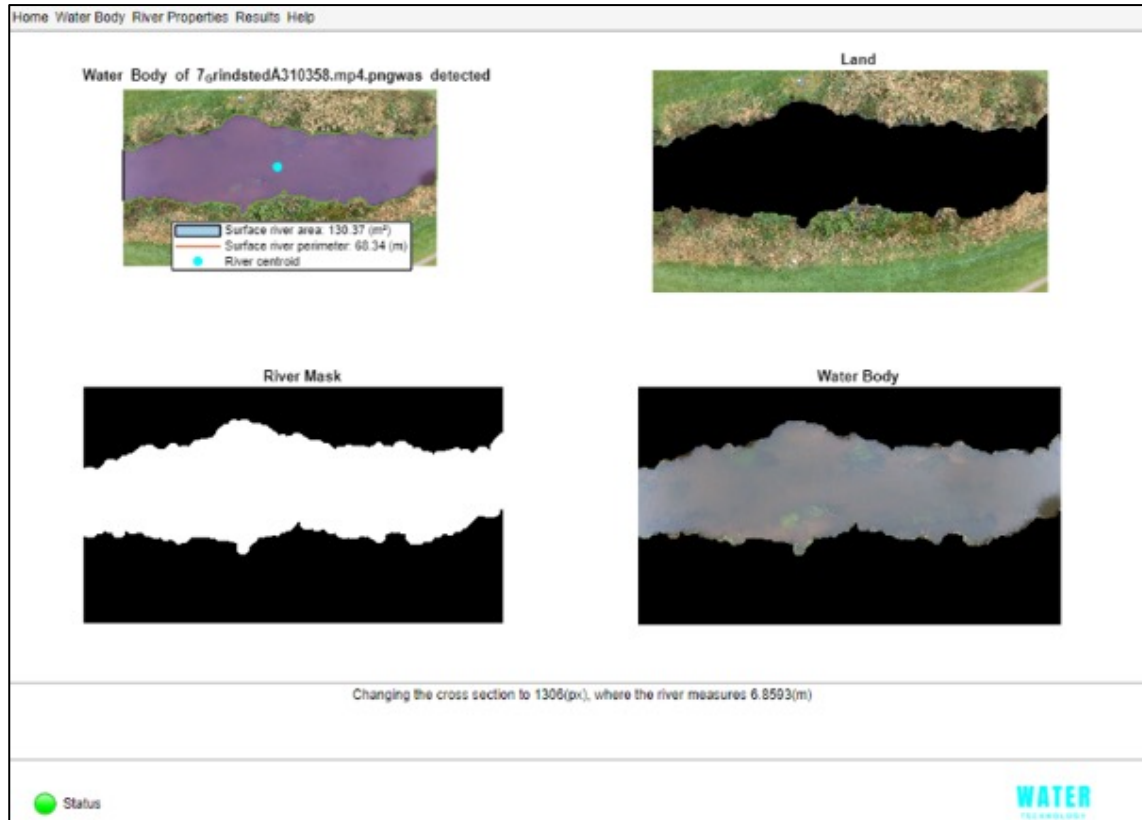
27





# Water extent segmentation

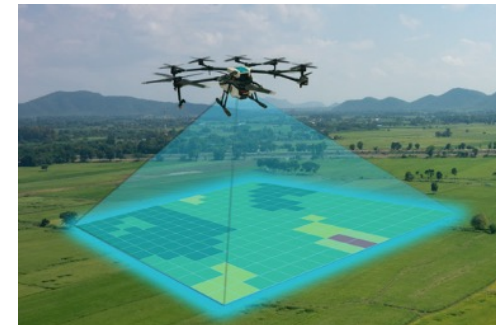
## 28 Software development and Graphical User Interface (GUI)



# Main UAS limitations and Future perspectives

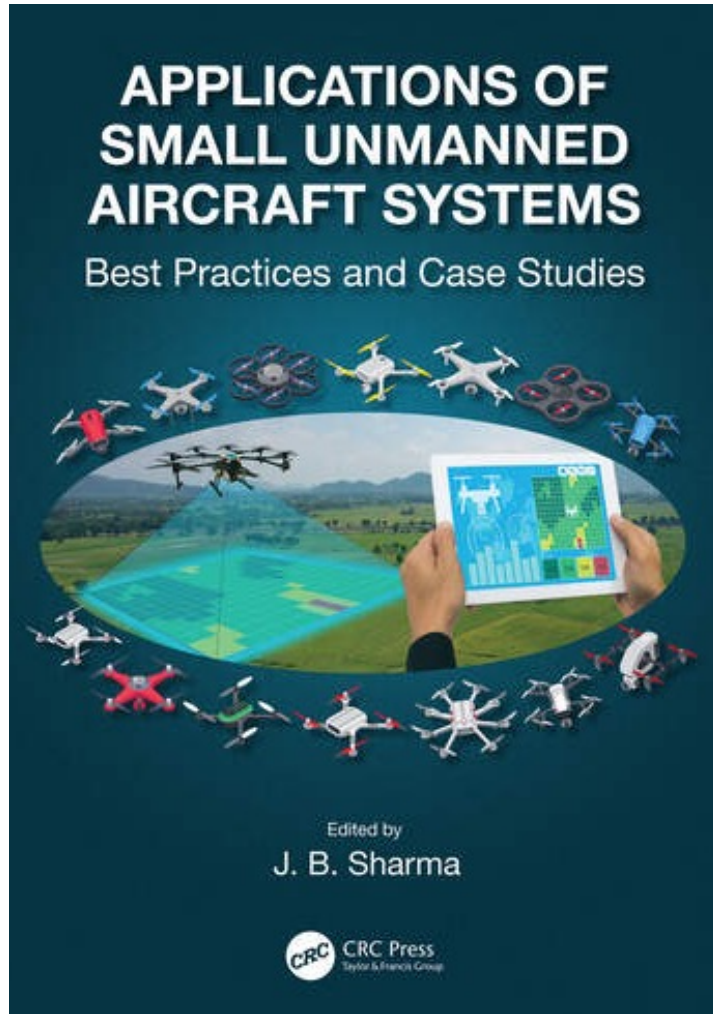
29

- ❑ UAS **maximum payload**, which limits the ability to use multiple sensors and communication hardware;
- ❑ National **flight regulations**, which limit the use of UAS, particularly in urban areas;
- ❑ The need for **continuous power supply** for frequent flight missions;
- ❑ The inability to fly in **extreme meteorological conditions**;
- ❑ The **amount of data** to manage.



# UAS books (recommended)

30



**Applications of Small Unmanned Aircraft Systems**  
1st Edition - 2022  
Best Practices and Case Studies  
Edited By J.B. Sharma



## Unmanned Aerial Systems for Monitoring Soil, Vegetation, and Riverine Environments



**Edited by**  
Salvatore Manfreda and Eyal Ben Dor



Earth Observation Series

**Unmanned Aerial Systems for Monitoring Soil, Vegetation, and Riverine Environments**  
1st Edition - January 2023  
Editors: Salvatore Manfreda, Ben Dor Eyal





# Codes and Data availability (some of them!)

31

- Pizarro, A., Latorre, M. A. G., & Alcayaga, H. (2022, December 8). Automatic Segmentation of Water Bodies Using RGB Data: A Physically-Based approach. <https://doi.org/10.17605/OSF.IO/3JXFD>
- Pizarro, A., Dal Sasso, S. F., & Manfreda, S. (2022, March 1). VISION: VIdeo StabilisatIOn using automatic features selection. <https://doi.org/10.17605/OSF.IO/HBRF2>
- Dal Sasso SF, Pizarro A, Pearce S, Maddock I, Manfreda S. 2021. Increasing LSPIV performances by exploiting the seeding distribution index at different spatial scales (Version 0.1). [codes] OSF. <https://doi.org/10.17605/OSF.IO/3AJNR>
- Pizarro, A., Dal Sasso, S. F., & Manfreda, S. (2020, September 28). Refining image-velocimetry performances for streamflow monitoring: Seeding metrics to errors minimisation. <https://doi.org/10.17605/OSF.IO/B7EAW>
- Pizarro, A., Dal Sasso, S. F., Perks, M. T., and Manfreda, S. 2020. Identifying the optimal spatial distribution of tracers for optical sensing of stream surface flow (Version 0.1), [codes], OSF, <https://doi.org/10.17605/OSF.IO/8EGQW>



# References & Recommended papers [1/2]

32

- Bandini, F., Olesen, D., Jakobsen, J., Kittel, C. M. M., Wang, S., Garcia, M., and Bauer-Gottwein, P.: Technical note: Bathymetry observations of inland water bodies using a tethered single-beam sonar controlled by an unmanned aerial vehicle, *Hydrol. Earth Syst. Sci.*, 22, 4165–4181, 2018
- Dal Sasso, Pizarro, A. et al. (2021) “Increasing LSPIV performances by exploiting the seeding distribution index at different spatial scales”. *Journal of Hydrology*, 598, 126438. <https://doi.org/10.1016/j.jhydrol.2021.126438>
- Dal Sasso, S. F., A. Pizarro, C. Samela, L. Mita, and S. Manfreda, Exploring the optimal experimental setup for surface flow velocity measurements using PTV, *Environmental Monitoring and Assessment*, 190:460, (doi: 10.1007/s10661-018-6848-3) 2018.
- Dal Sasso, S. F., Pizarro, A. et al. (2020) “Metrics for the Quantification of Seeding Characteristics to Enhance Image Velocimetry Performance in Rivers”. *Remote Sensing*, 12(11), 1789; <https://doi.org/10.3390/rs12111789>
- Jolley MJ, Russell AJ, Quinn PF and Perks MT (2021) Considerations When Applying Large-Scale PIV and PTV for Determining River Flow Velocity. *Front. Water* 3:709269
- Manfreda, S.; McCabe, M.F.; Miller, P.E.; Lucas, R.; Pajuelo Madrigal, V.; Mallinis, G.; Ben Dor, E.; Helman, D.; Estes, L.; Ciruolo, G.; Müllerová, J.; Tauro, F.; De Lima, M.I.; De Lima, J.L.M.P.; Maltese, A.; Frances, F.; Caylor, K.; Kohv, M.; Perks, M.; Ruiz-Pérez, G.; Su, Z.; Vico, G.; Toth, B. On the Use of Unmanned Aerial Systems for Environmental Monitoring. *Remote Sens.* 2018, 10, 641.
- Manfreda, S.; McCabe, M.F.; Miller, P.E.; Lucas, R.; Pajuelo Madrigal, V.; Mallinis, G.; Ben Dor, E.; Helman, D.; Estes, L.; Ciruolo, G.; Müllerová, J.; Tauro, F.; De Lima, M.I.; De Lima, J.L.M.P.; Maltese, A.; Frances, F.; Caylor, K.; Kohv, M.; Perks, M.; Ruiz-Pérez, G.; Su, Z.; Vico, G.; Toth, B. On the Use of Unmanned Aerial Systems for Environmental Monitoring. *Remote Sens.* 2018, 10, 641.



# References & Recommended papers [2/2]

33

- Manfreda, S., S. F. Dal Sasso, A. Pizarro, F. Tauro, Chapter 10: New Insights Offered by UAS for River Monitoring, Applications of Small Unmanned Aircraft Systems: Best Practices and Case Studies, CRC Press, Taylor & Francis Group, 211, (doi: 10.1201/9780429244117-10) 2019.
- Paruta, A., P. Nasta, G. Ciruolo, F. Capodici, S. Manfreda, N. Romano, E. Bendor, Y. Zeng, A. Maltese, S. F. Dal Sasso and R. Zhuang, A geostatistical approach to map near-surface soil moisture through hyper-spatial resolution thermal inertia, IEEE Transactions on Geoscience and Remote Sensing, 2020.
- Perks, M., Pizarro, A. et al. (2020) "Towards harmonisation of image velocimetry techniques for river surface velocity observations". Earth Syst. Sci. Data, 12, 1545–1559, <https://doi.org/10.5194/essd-12-1545-2020>
- Pizarro, A., Dal Sasso, S. F., & Manfreda, S. (2022). VISION: Video Stabilisation using automatic features selection for image velocimetry analysis in rivers. SoftwareX, 19, 101173, <http://dx.doi.org/10.1016/j.softx.2022.101173>
- Pizarro, A., Dal Sasso, S. F., Manfreda, S. (2020) "Refining image-velocimetry performances for streamflow monitoring: Seeding metrics to errors minimisation". Hydrological Processes; 1–9. <https://doi.org/10.1002/hyp.1391916>.
- Pizarro, A. et al. (2020) "Identifying the optimal spatial distribution of tracers for optical sensing of stream surface flow". Hydrology and Earth System Sciences (HESS), <https://doi.org/10.5194/hess-24-5173-202015>.





**Thanks for your attention!**

**Questions?**

