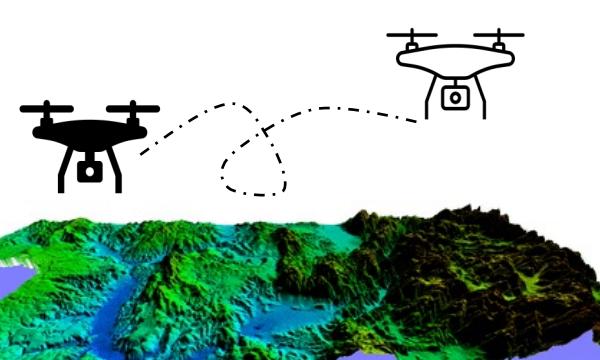




The use of innovative technology for fluvial monitoring

Gravel Bed Rivers 9 - 13 January 2023



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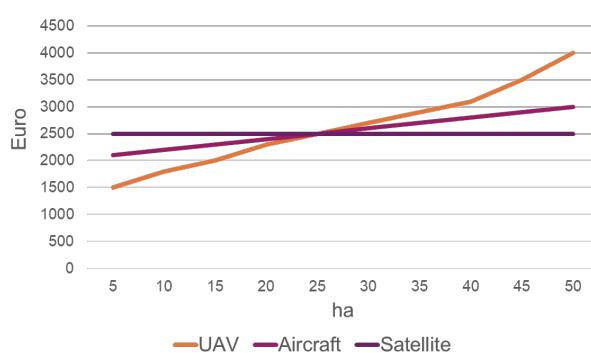
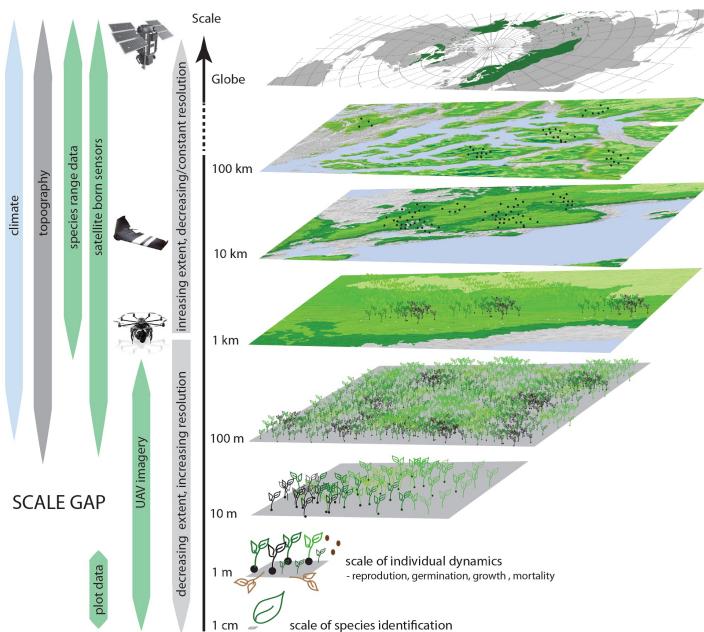


PONTIFICIA
UNIVERSIDAD
CATÓLICA
DE CHILE

Environmental monitoring at a glance

2

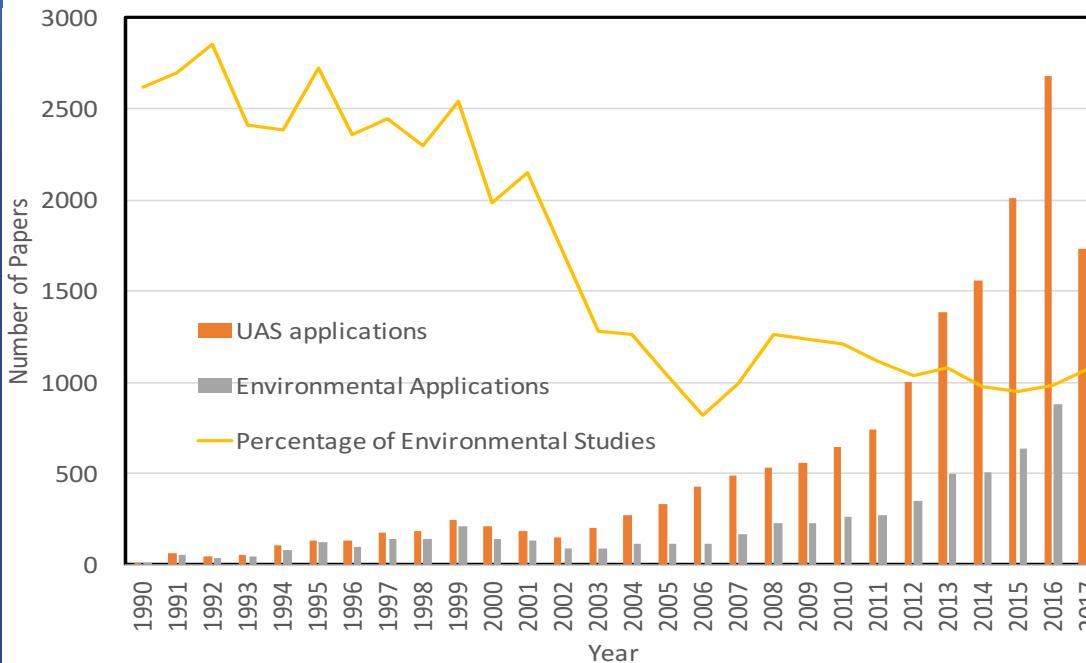
Different sensing alternatives. Which one should I use?



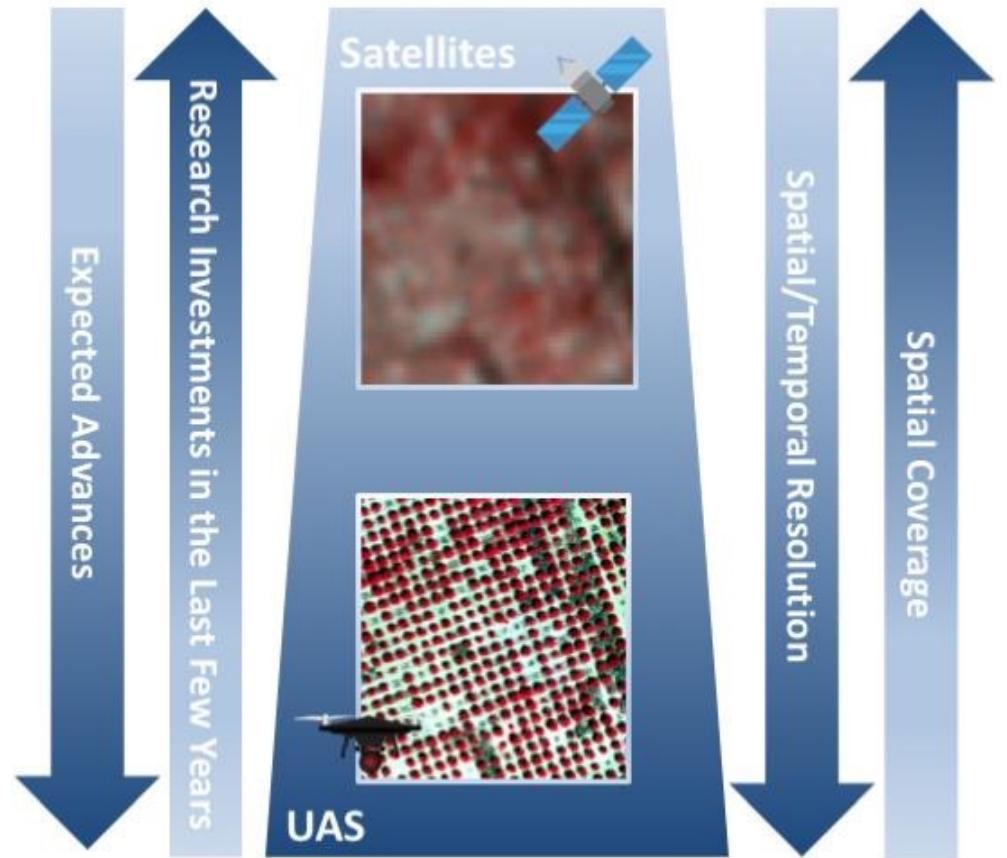
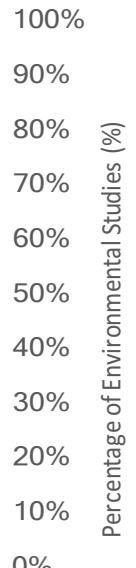
Environmental monitoring at a glance

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)



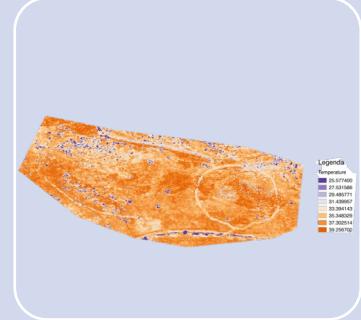
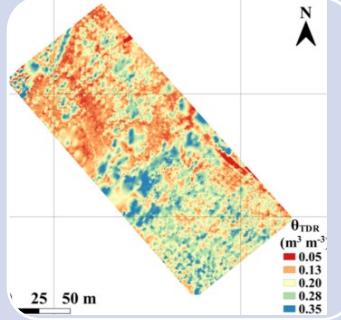
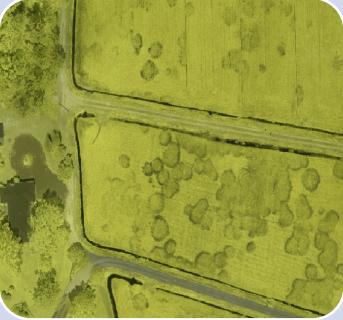
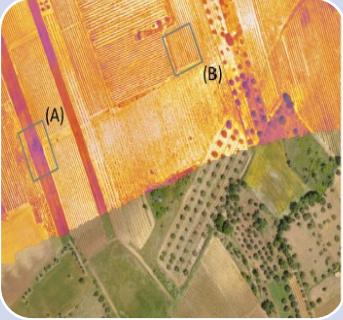
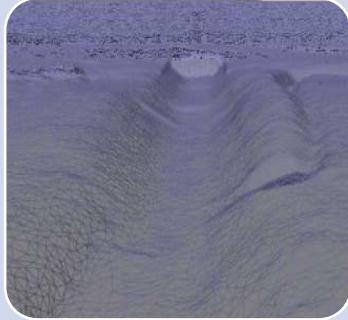
Multirotor 4-6 Kg



UAS applications for Environmental monitoring

5

Some (of many!) applications...



DSMs

Crop water
stress
detection

Vegetation
status

Soil water
content

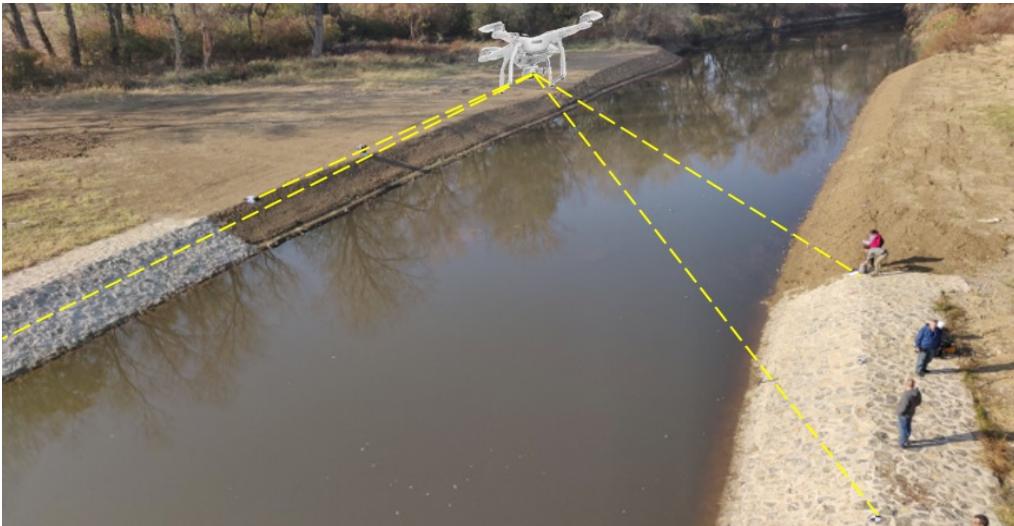
Archaeological
surveys

fluvial
monitoring

Outline

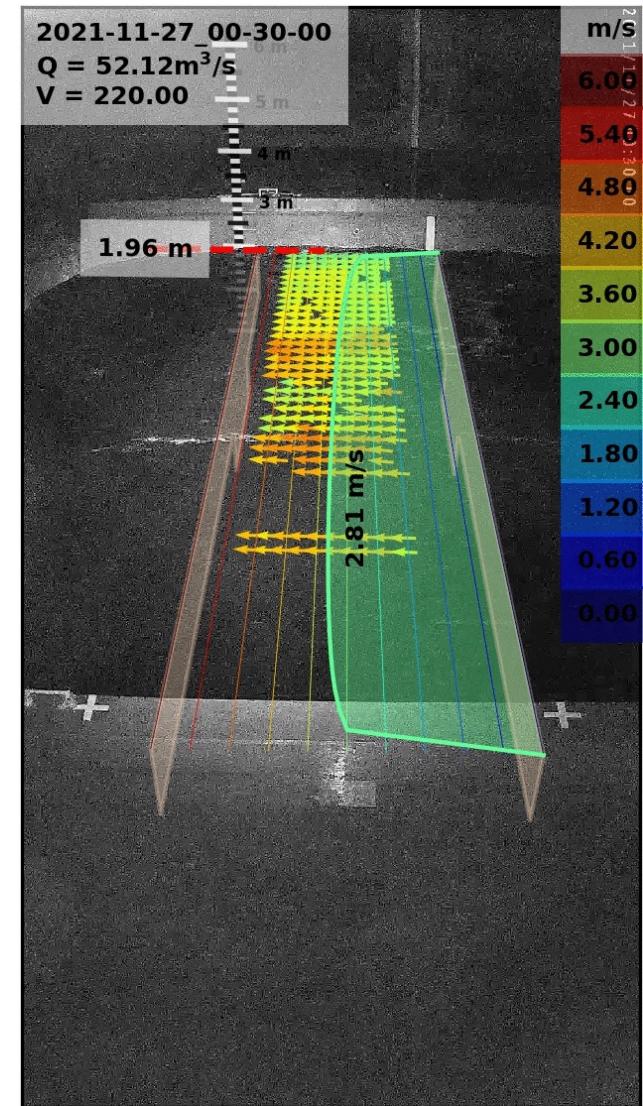
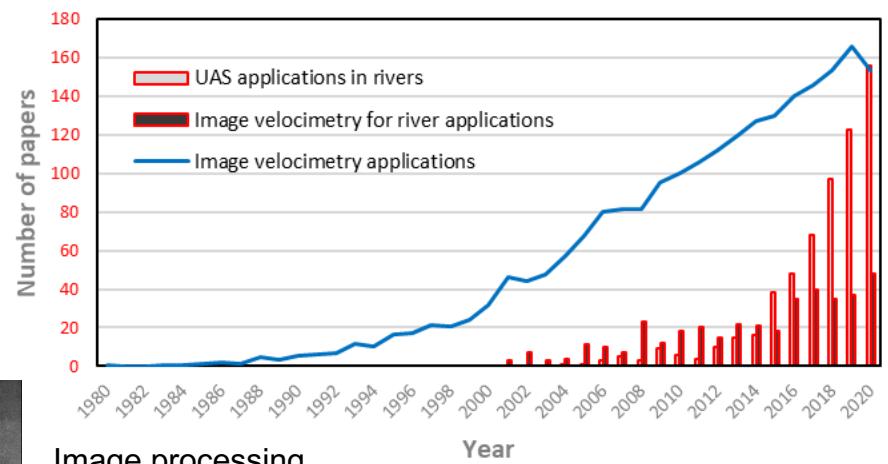
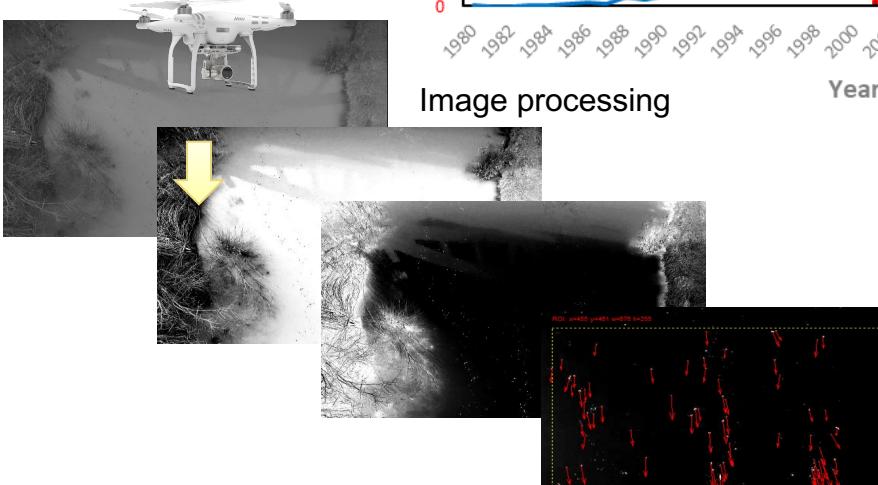
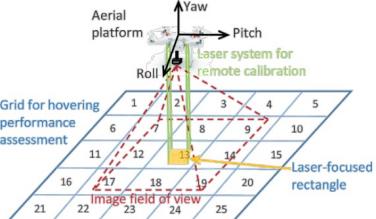
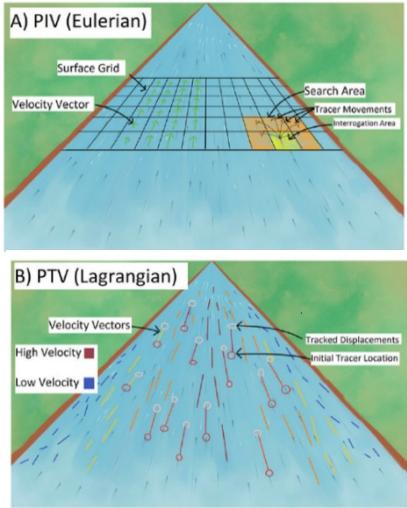
6

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



Streamflow monitoring

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- ✓ **Scope:** Streamflow monitoring: velocities, discharge, and water levels.
- ✓ **Versatility:** smartphones, fixed cameras, UASs.

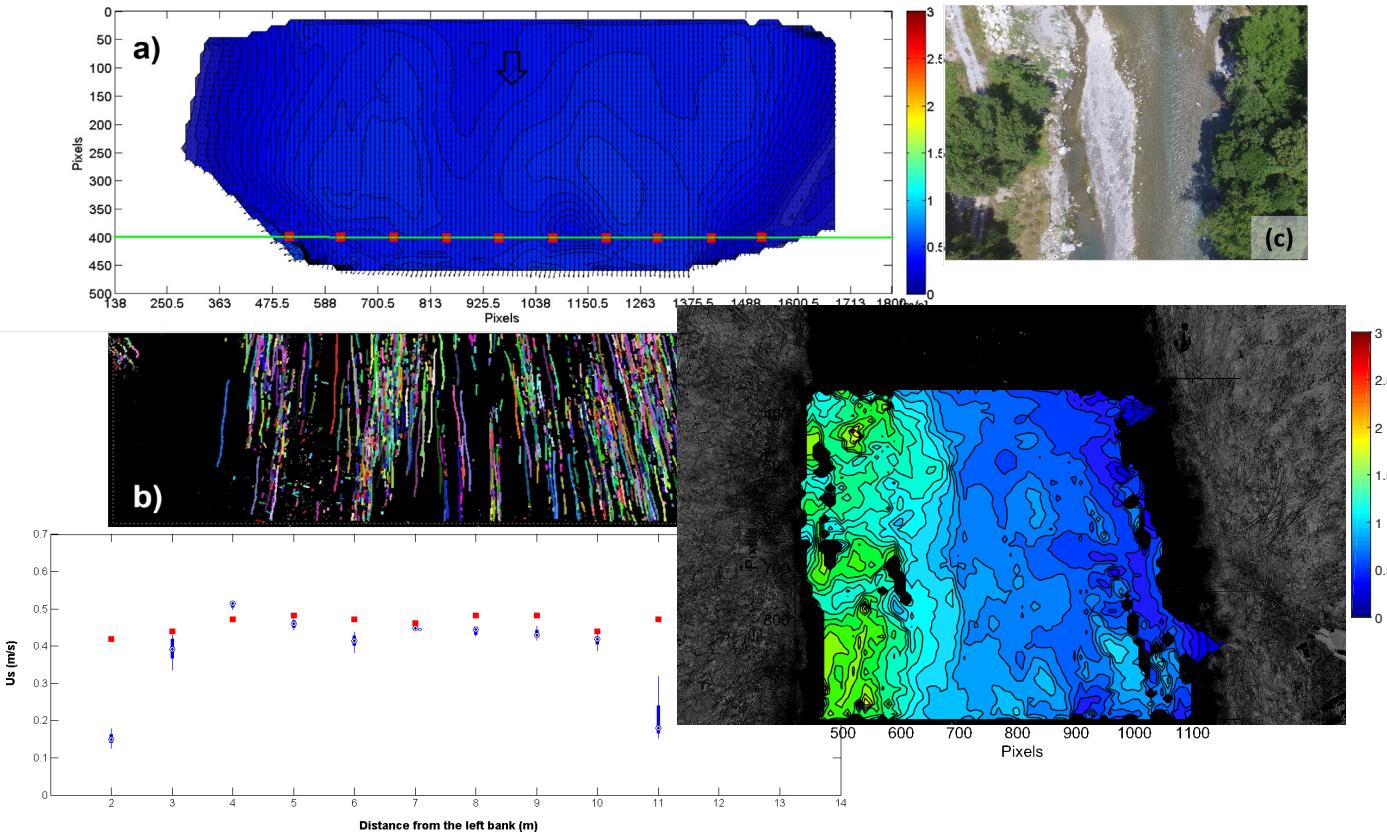
<http://www.photrack.ch/dischargekeeper.html>



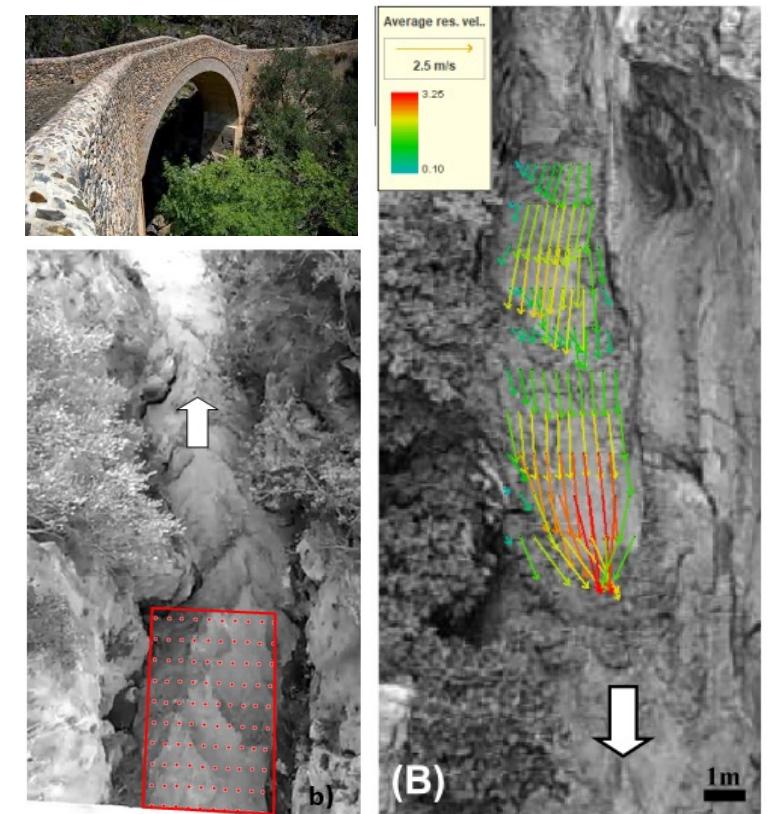
Streamflow monitoring

8

Field experience & Analysis



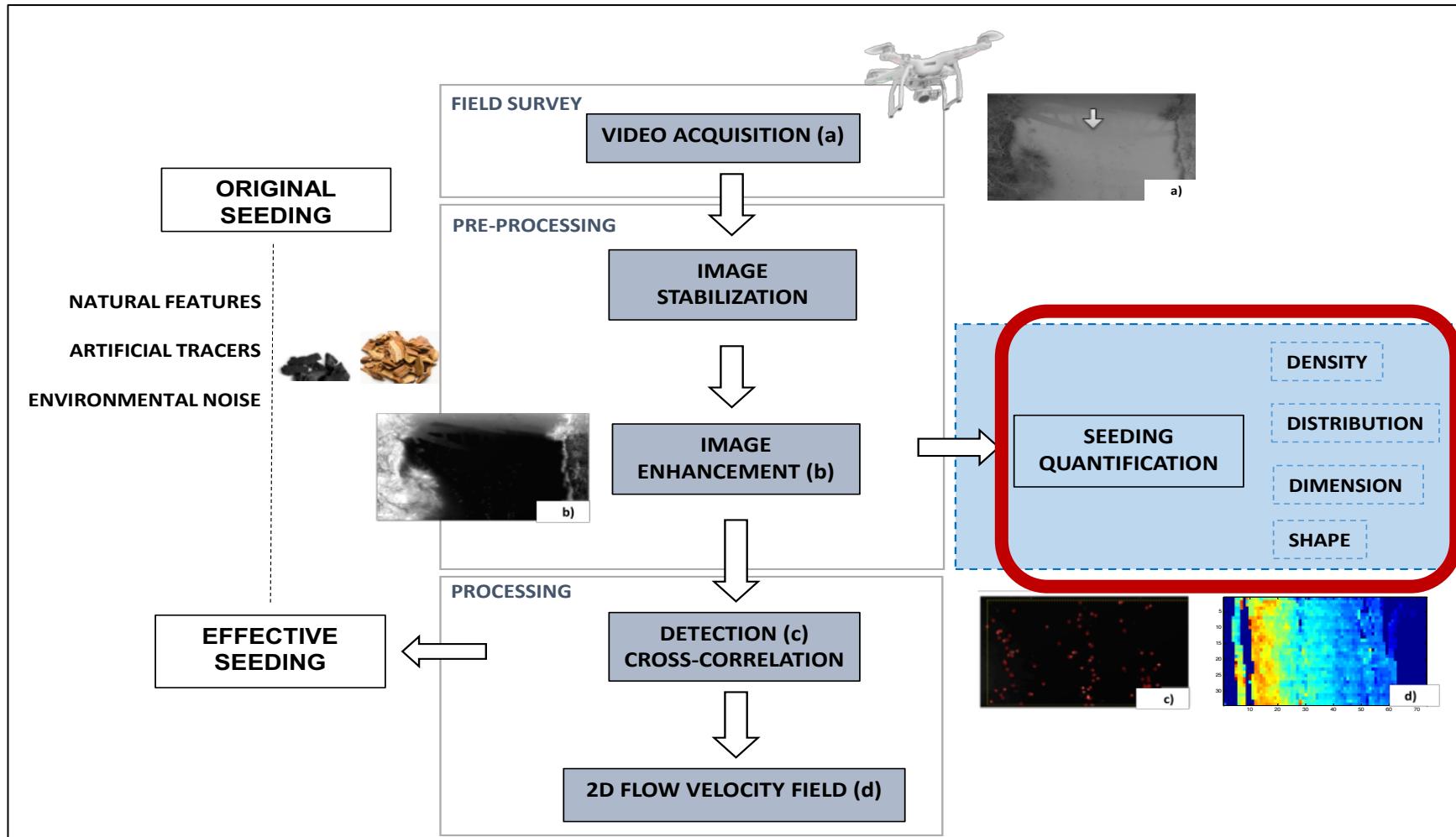
Torrente Raganello (Civita, Italy)



Streamflow monitoring

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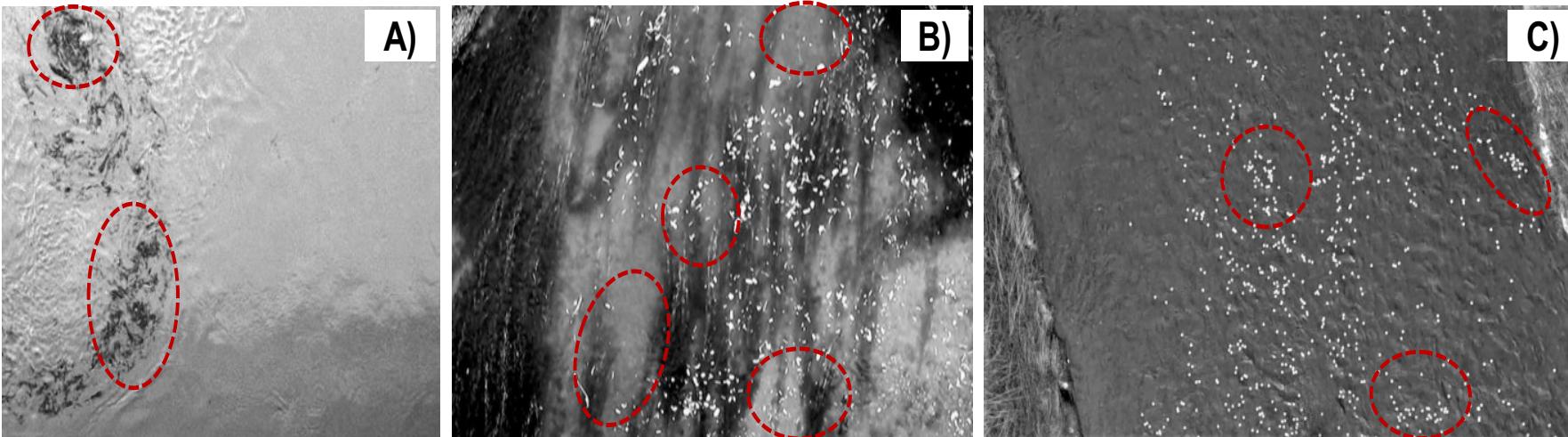
Workflow for image-velocimetry analysis



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**

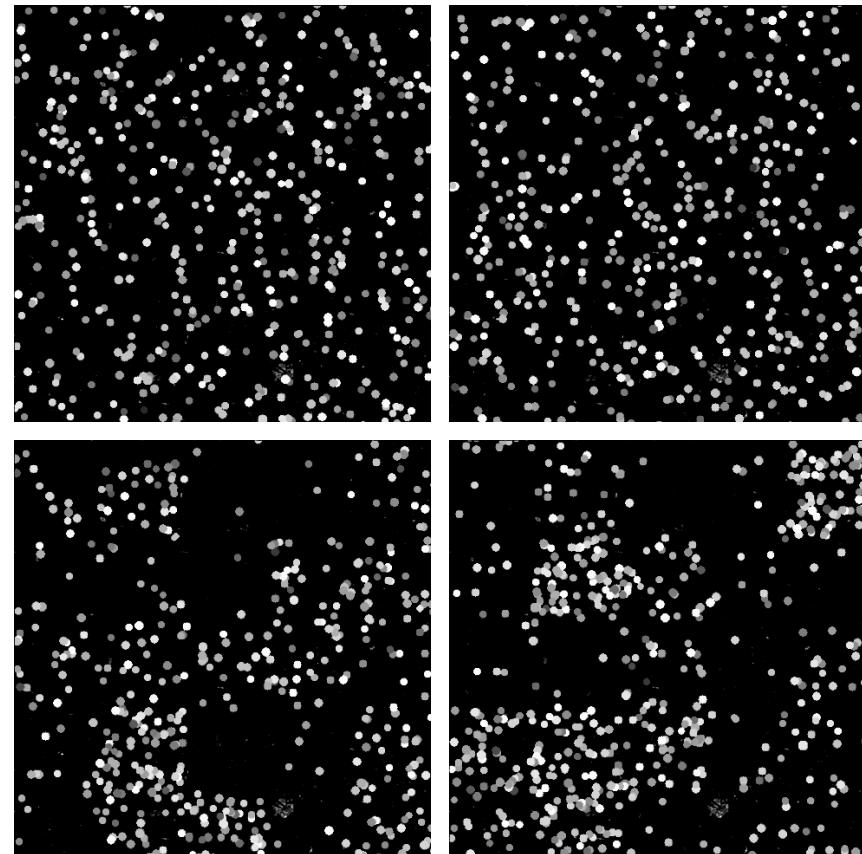
Seeding Quantification

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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 λ and level of aggregation v .
- 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

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Numerical framework: Analysis

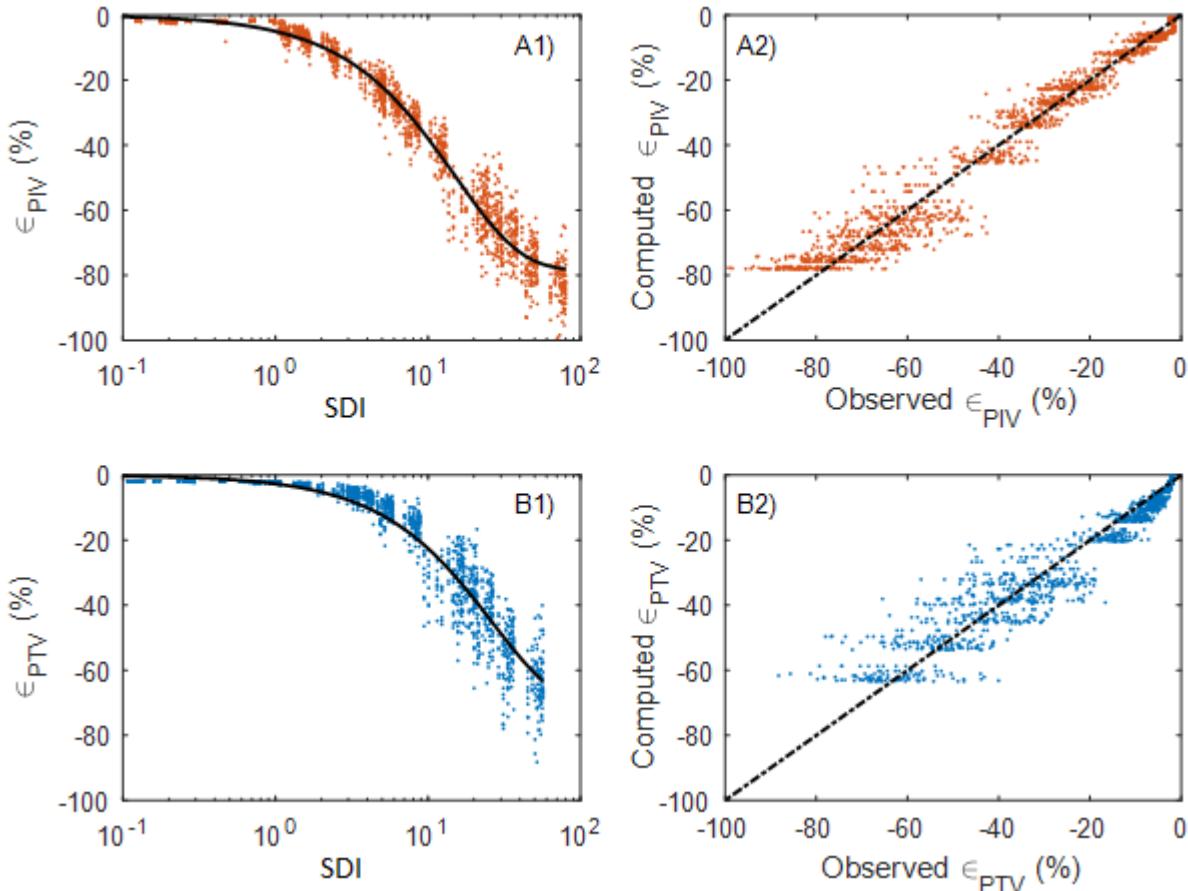


Fig. Numerical simulations of synthetically generated particles that present different aggregation levels.

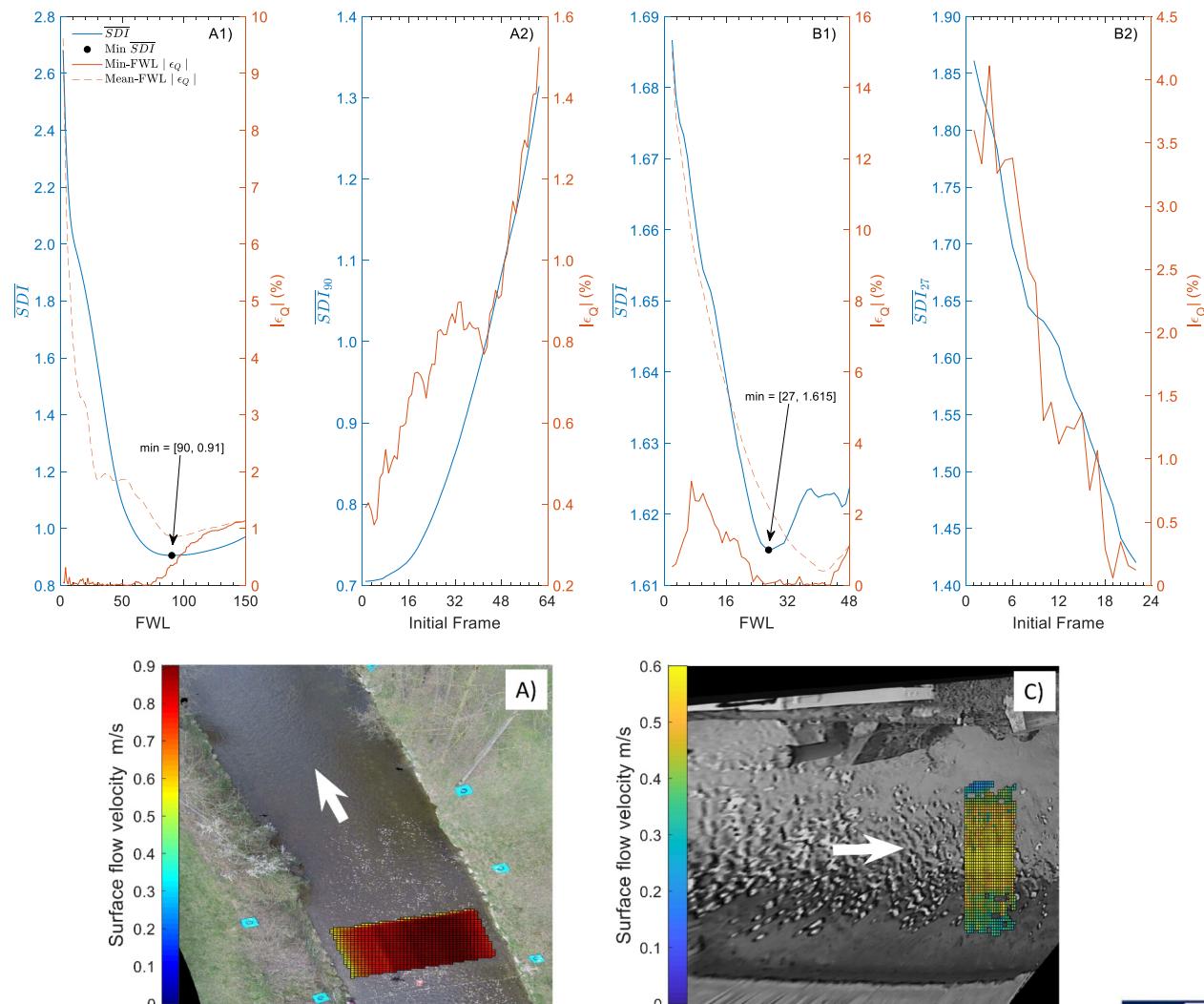
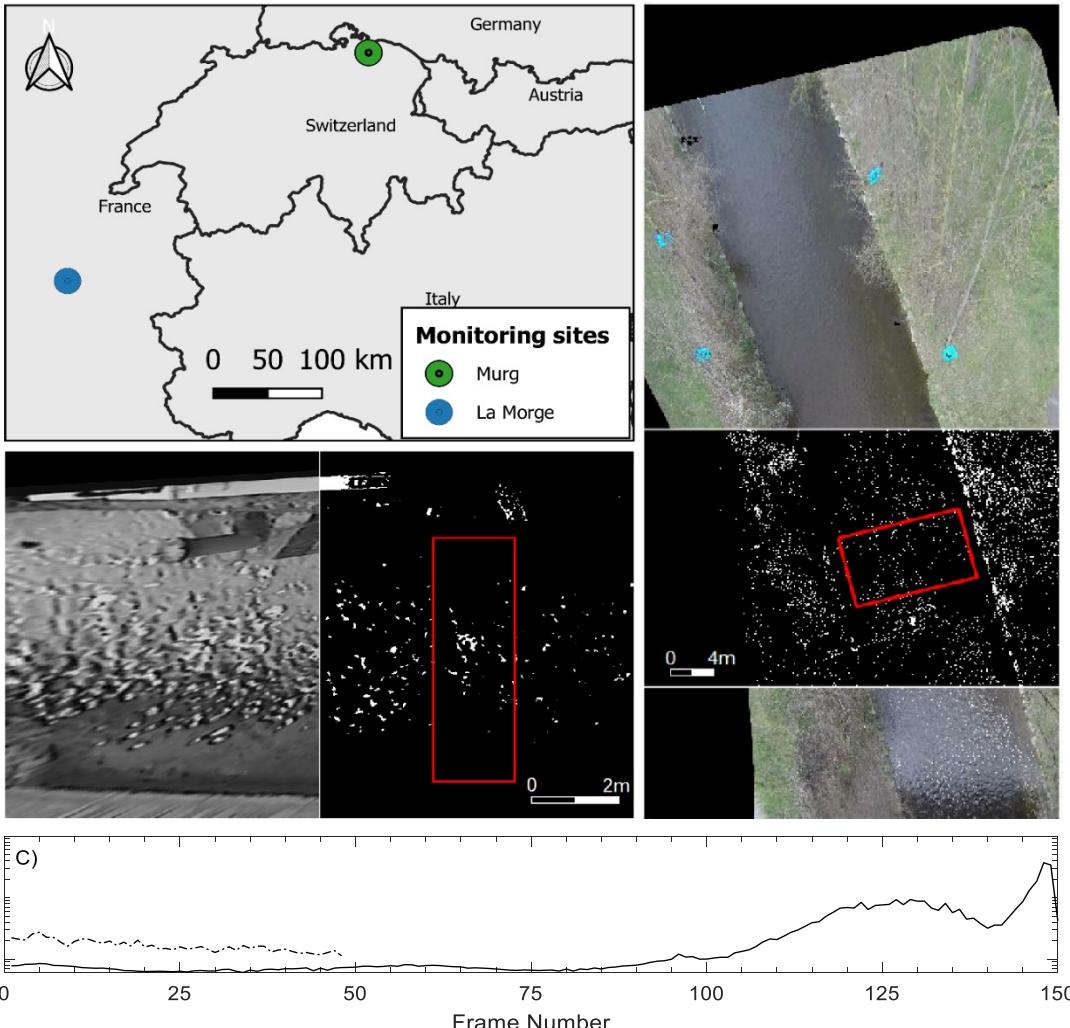
$$SDI = D^{*0.1} / \left(\frac{\rho}{\rho_c D^{*1}} \right)$$

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

SDI at field conditions

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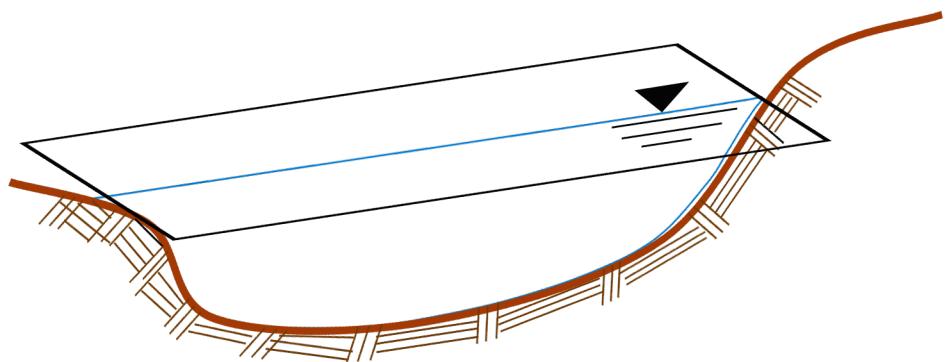
SDI applications: 1.0



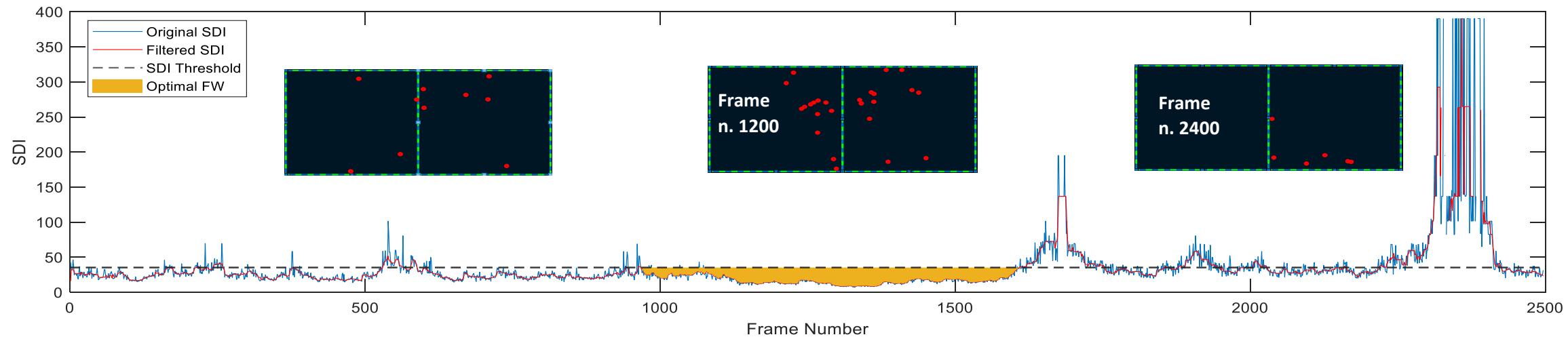
SDI at field conditions

14

SDI applications: 2.0



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



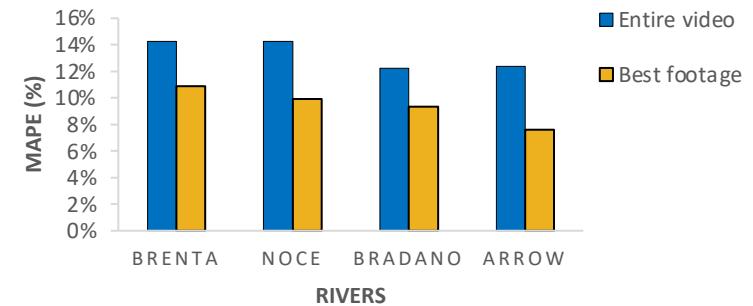
SDI at field conditions

15

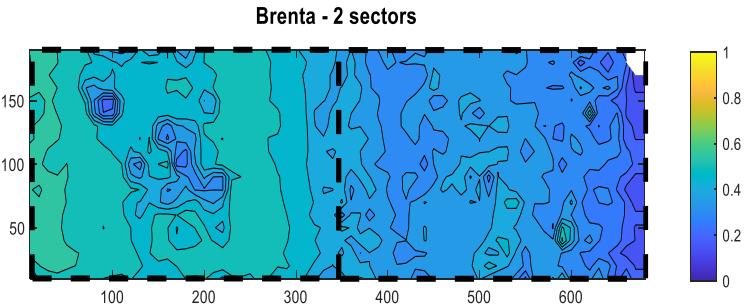
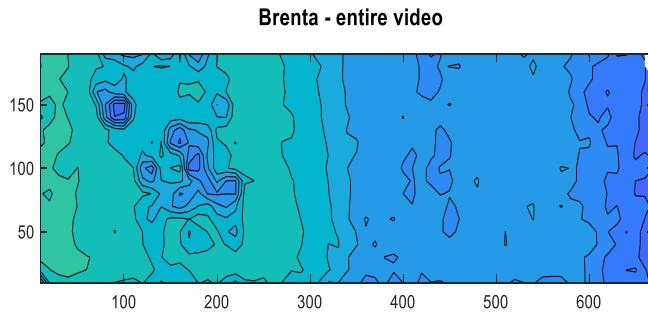
SDI applications: 2.0

- 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).

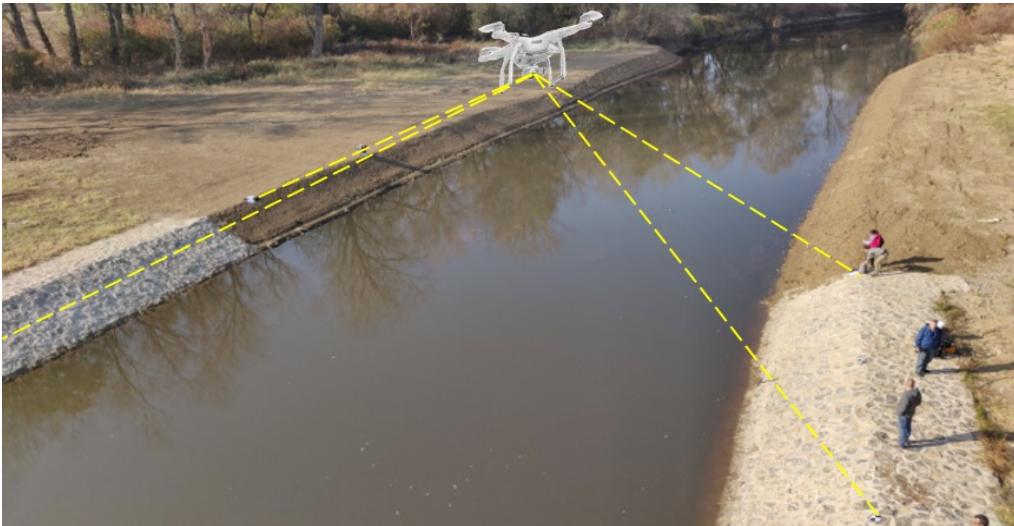


U component [m/s]

Outline

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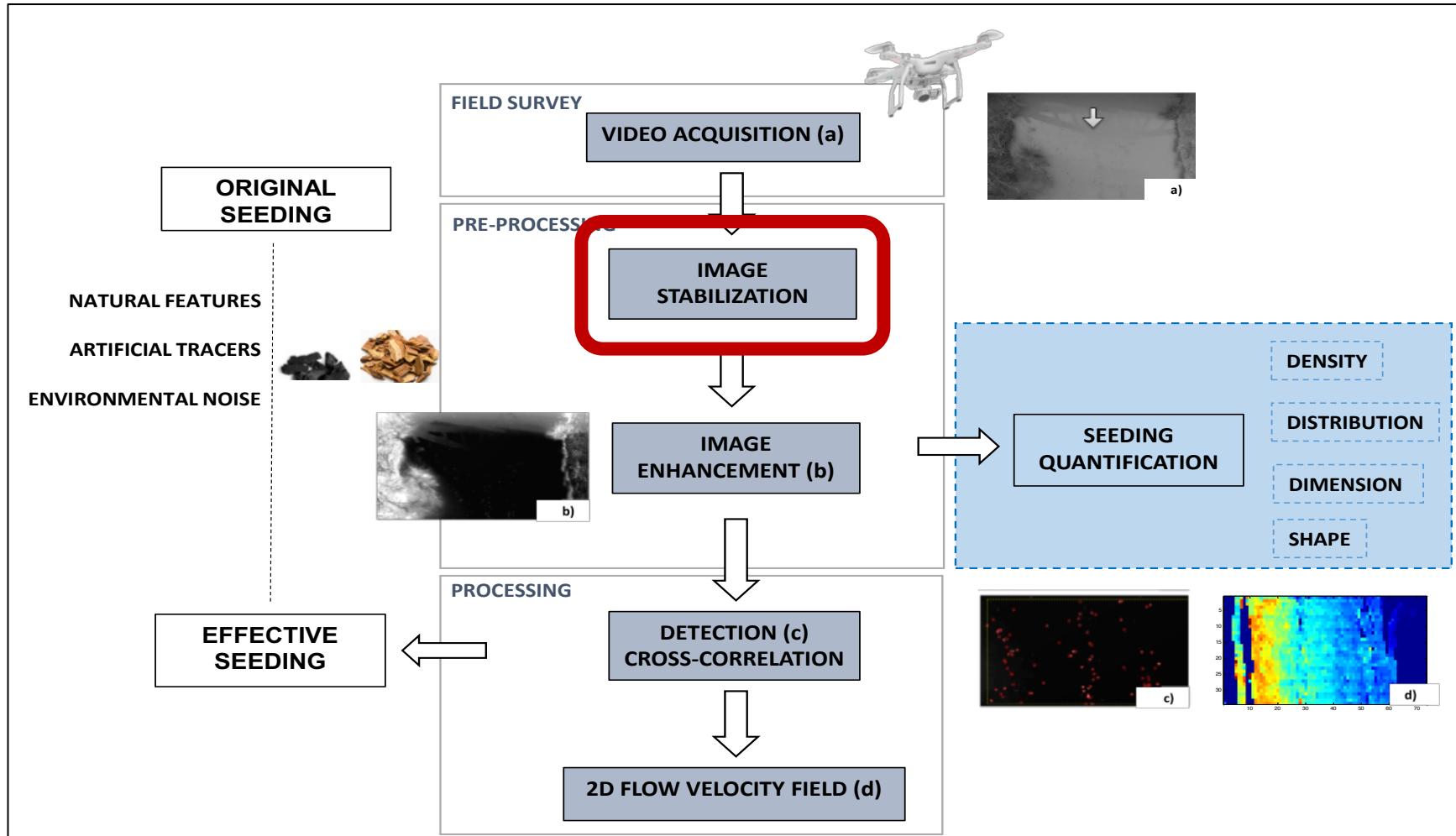
1. Seeding quantification for image velocimetry applications
2. Footage stabilisation for image velocimetry applications
3. Water extent segmentation



Streamflow monitoring

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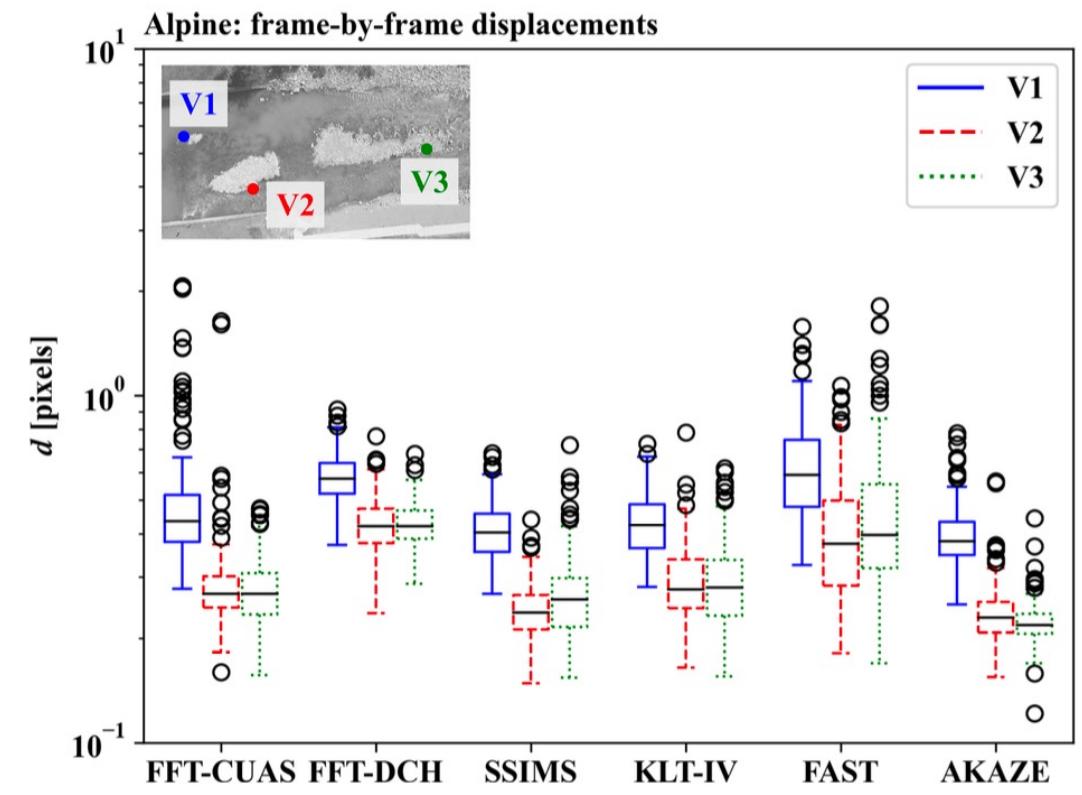
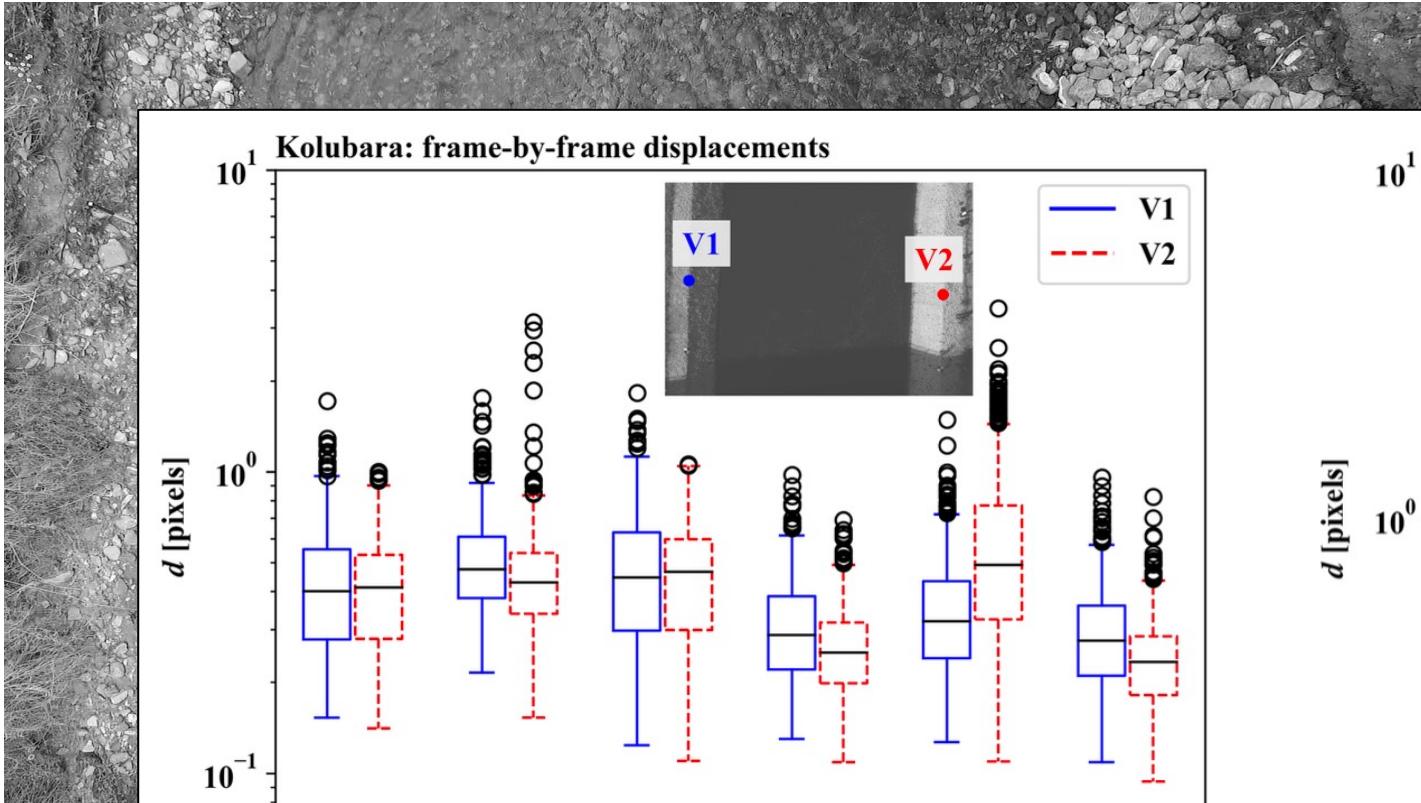
Workflow for image-velocimetry analysis



Streamflow monitoring

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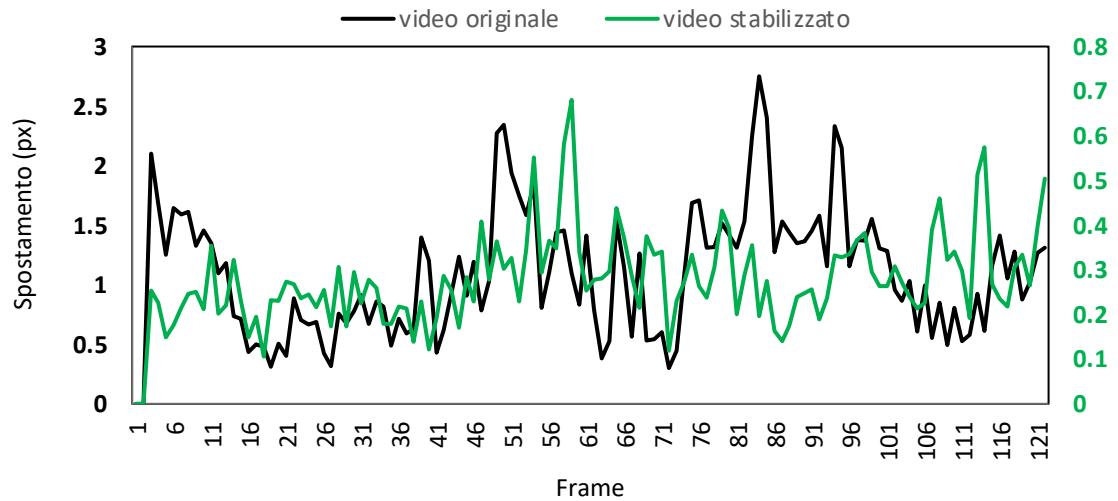
Stabilisation issues



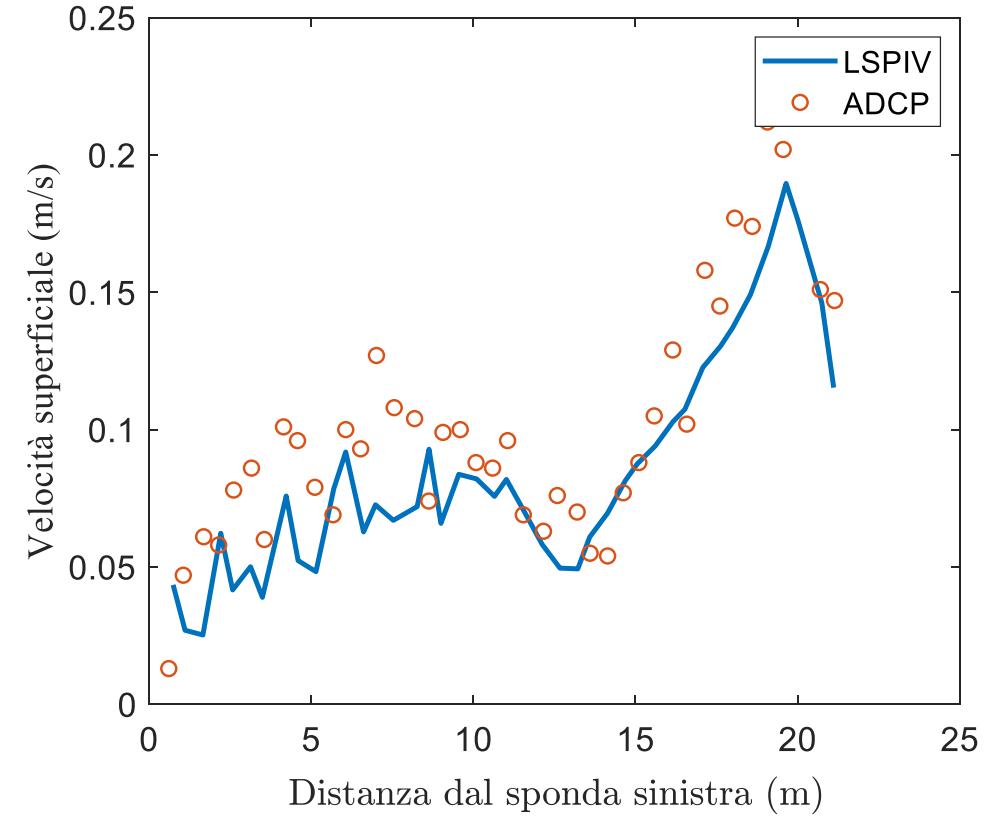
Stabilisation for image velocimetry

19

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



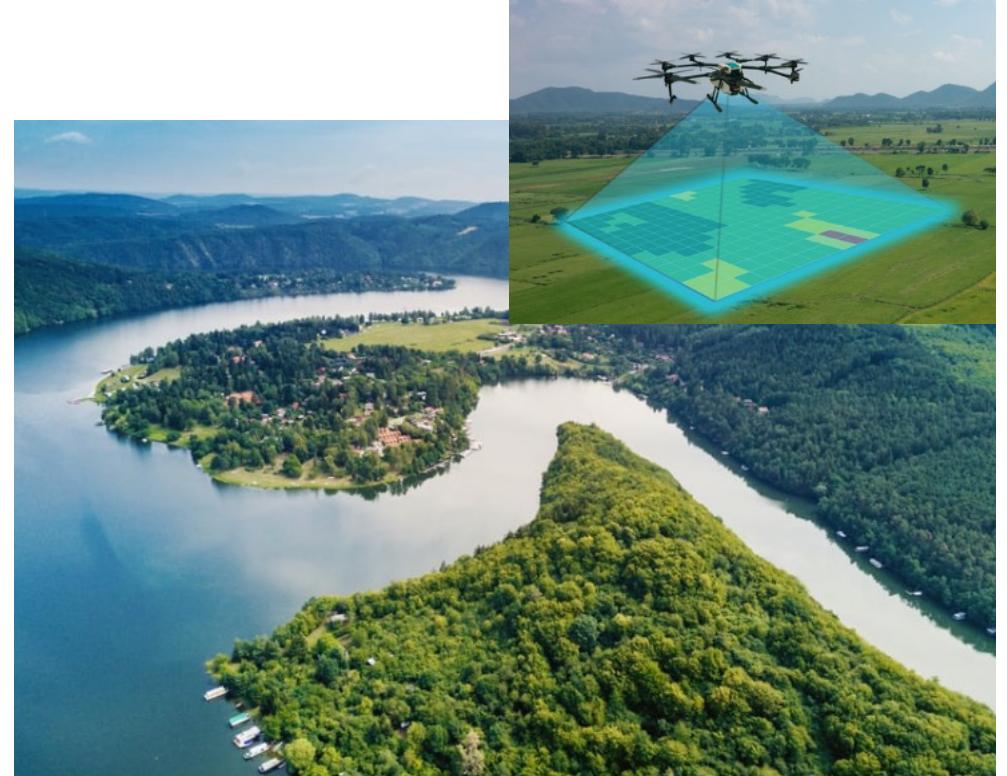
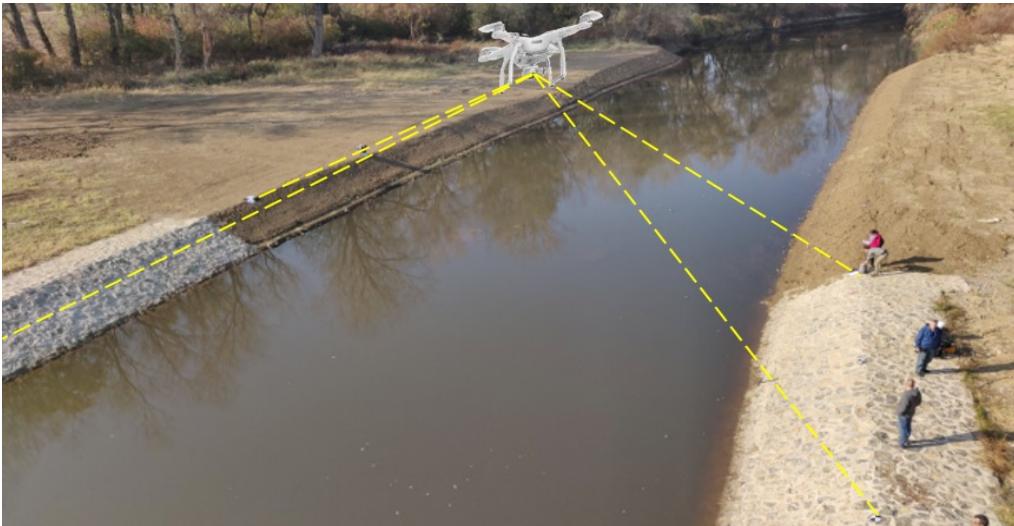
Sezione	Componente di Velocità	RMSE (m/s)		Riduzione errore (%)
		Video non stabilizzato	Video stabilizzato	
S1	v	0.07	0.06	8%
	u	0.02	0.02	-1%
	m	0.07	0.07	7%
S2	v	0.08	0.07	8%
	u	0.03	0.03	6%
	m	0.08	0.07	8%



Outline

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



Water extent segmentation

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How do we (computers) see rivers?

Humans:

- Colour
- Sound
- Water movement
- etc



Computers:

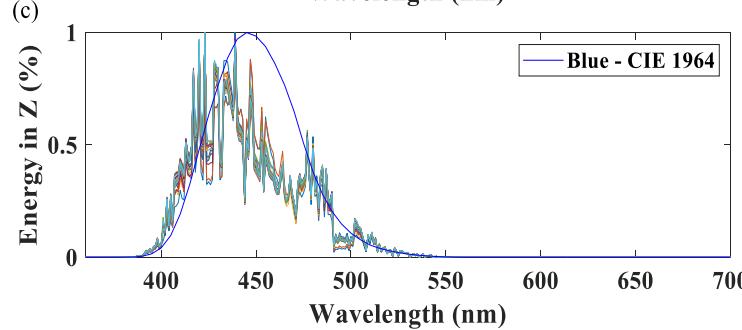
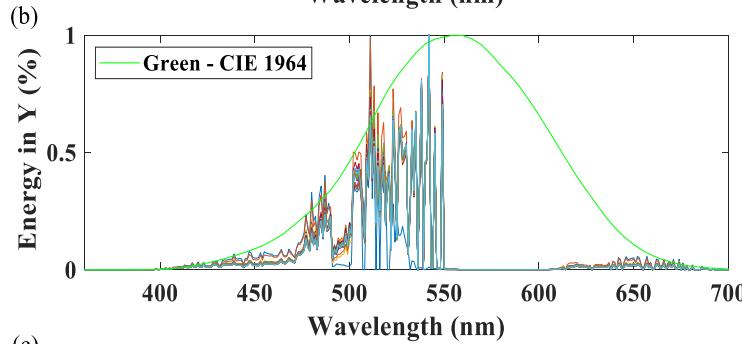
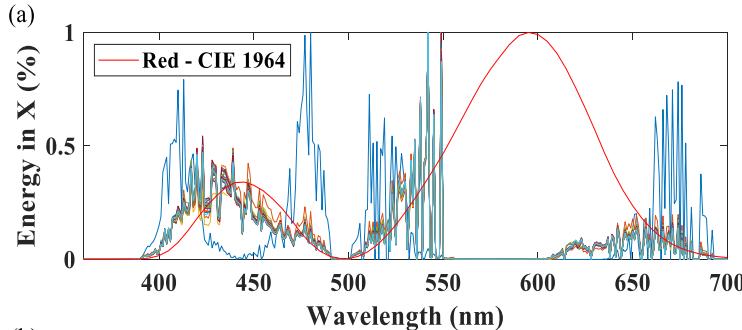
- Ones & Zeros

A grid of binary digits (0s and 1s) overlaid on the same aerial landscape image. The grid consists of 16 rows of 16 binary digits each. The pattern of 1s and 0s corresponds to the water extent segmentation, where 1s represent water and 0s represent land or vegetation. The binary code is as follows:
0111010011010011101001110100111010011101001110
1001110100111010011101001110100111010011101001110
101100010110001011000101100010110001011000101100010110
11000101100010110001011000101100010110001011000101100
111010011101001110100111010011101001110100111010011101
000101100010110001011000101100010110001011000101100010
001110100111010011101001110100111010011101001110100111
010011101001110100111010011101001110100111010011101001
011000101100010110001011000101100010110001011000101100
100010110001011000101100010110001011000101100010110001
101001110100111010011101001110100111010011101001110100
1101001110100111010011101001110100111010011101001110100
11

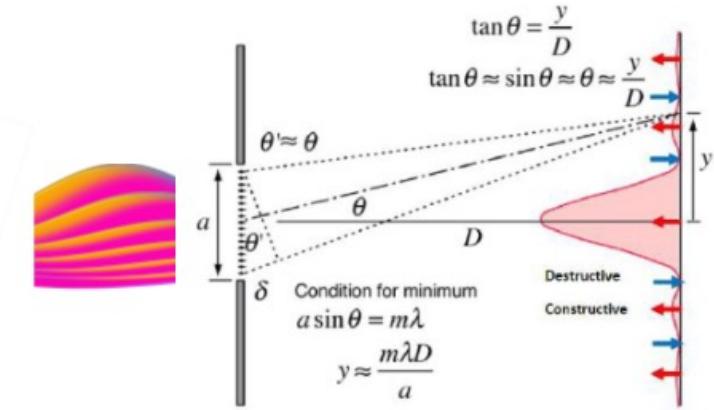
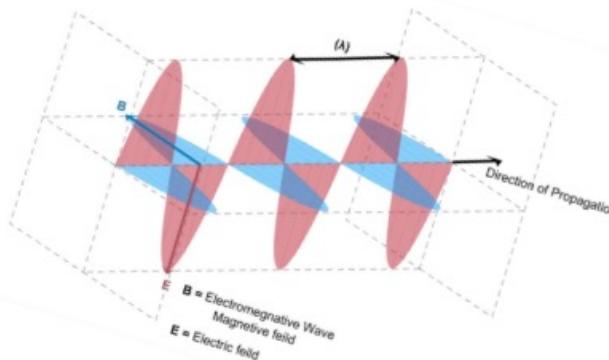
Water extent segmentation

22

Water extent segmentation by single slit diffraction correction?



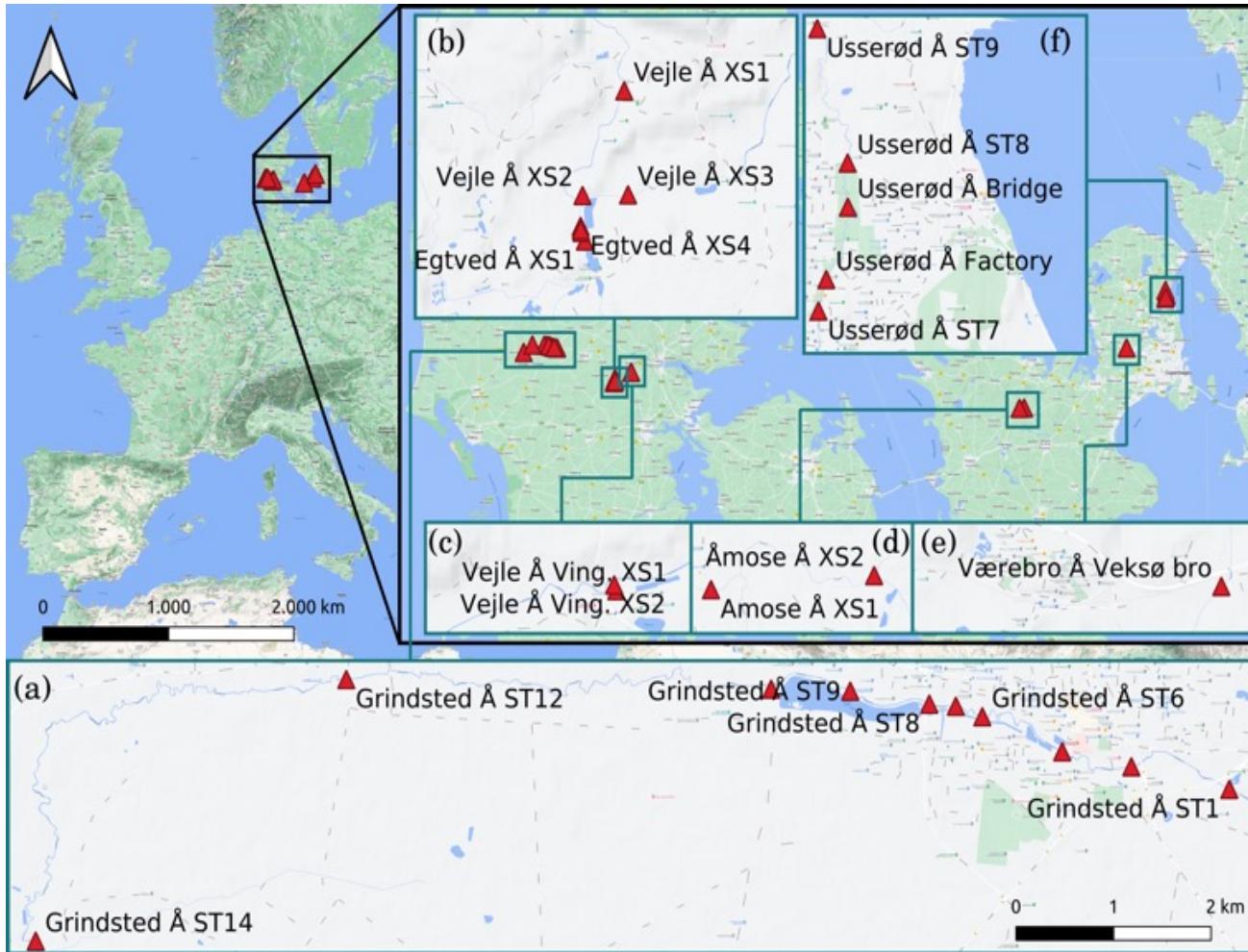
- Cameras are always affected by quantum interference
- CIE standard is a standardised way to represent colours
- What is acquired by the camera can be corrected to follow the CIE standard



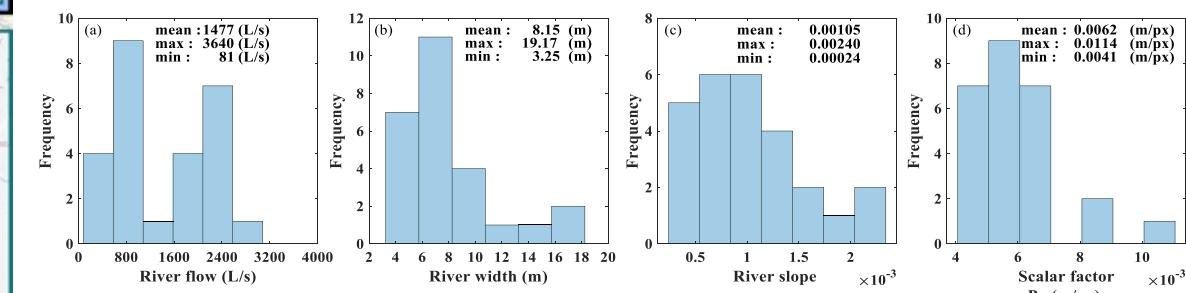
Water extent segmentation

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Case studies to test the model



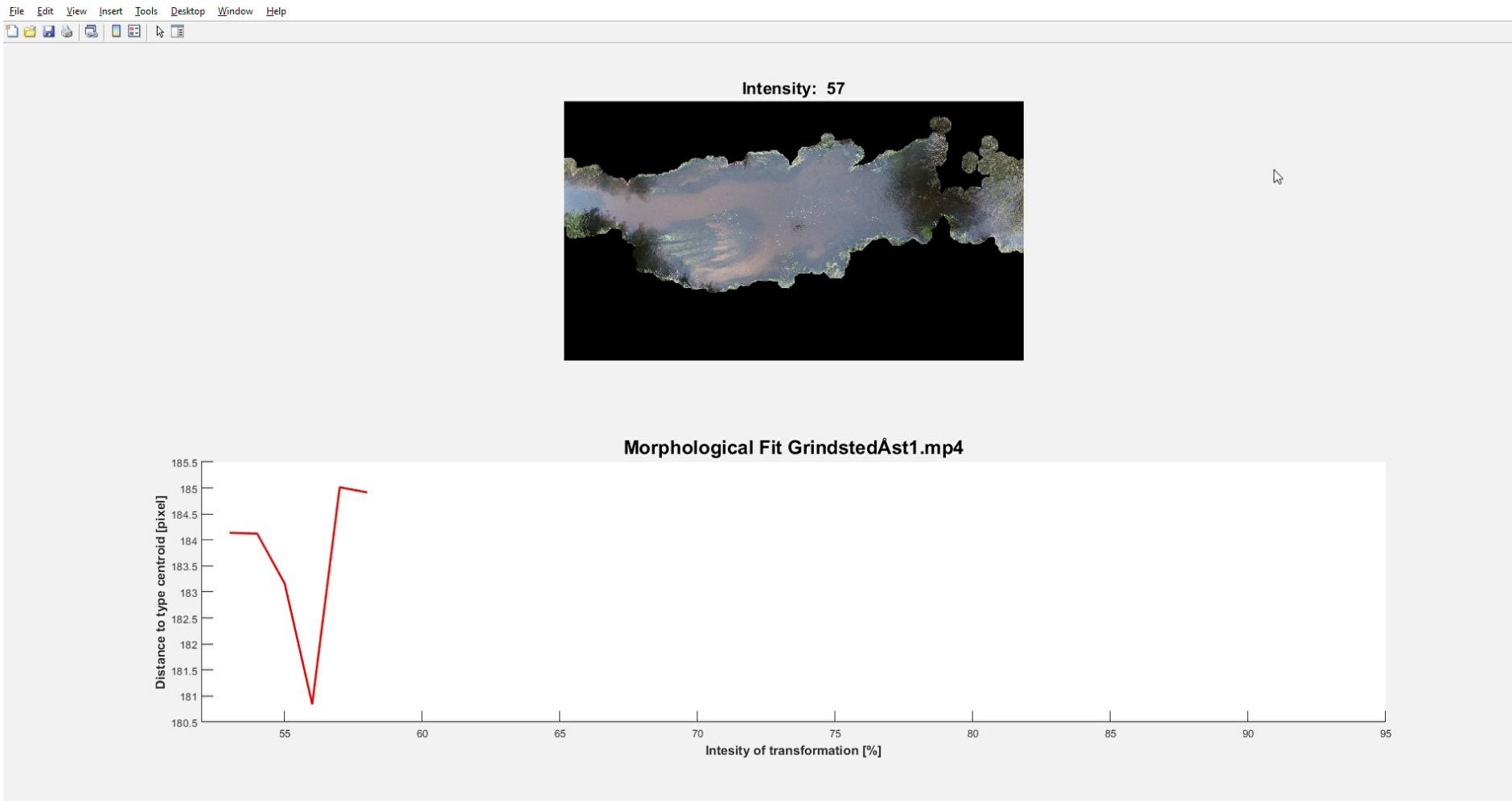
- 27 case studies located in Denmark
- Huge difference between them, e.g. Q ranges between 81 and 1477 L/s
- Riverbed in same cases has vegetation



Water extent segmentation

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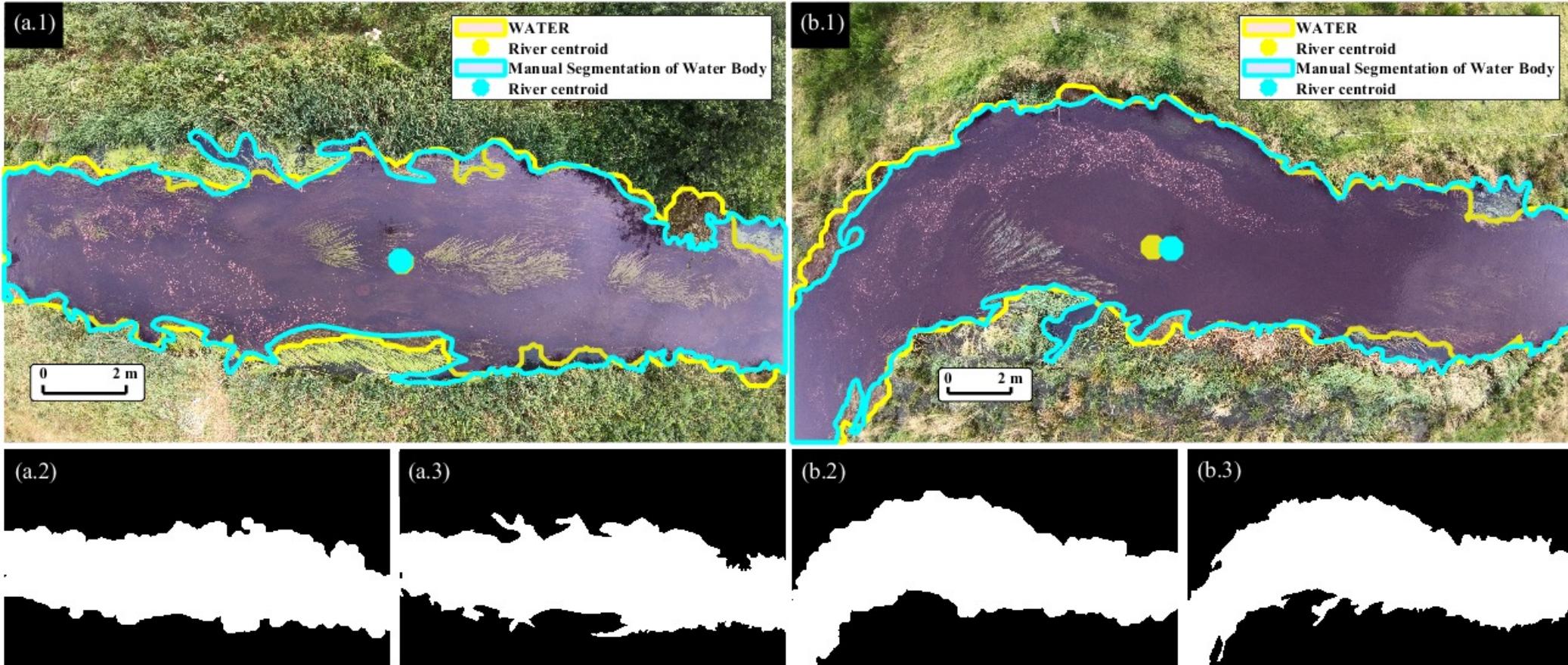
Water AuTomatic sEgmentation in Rivers (WATER)



Water extent segmentation

25

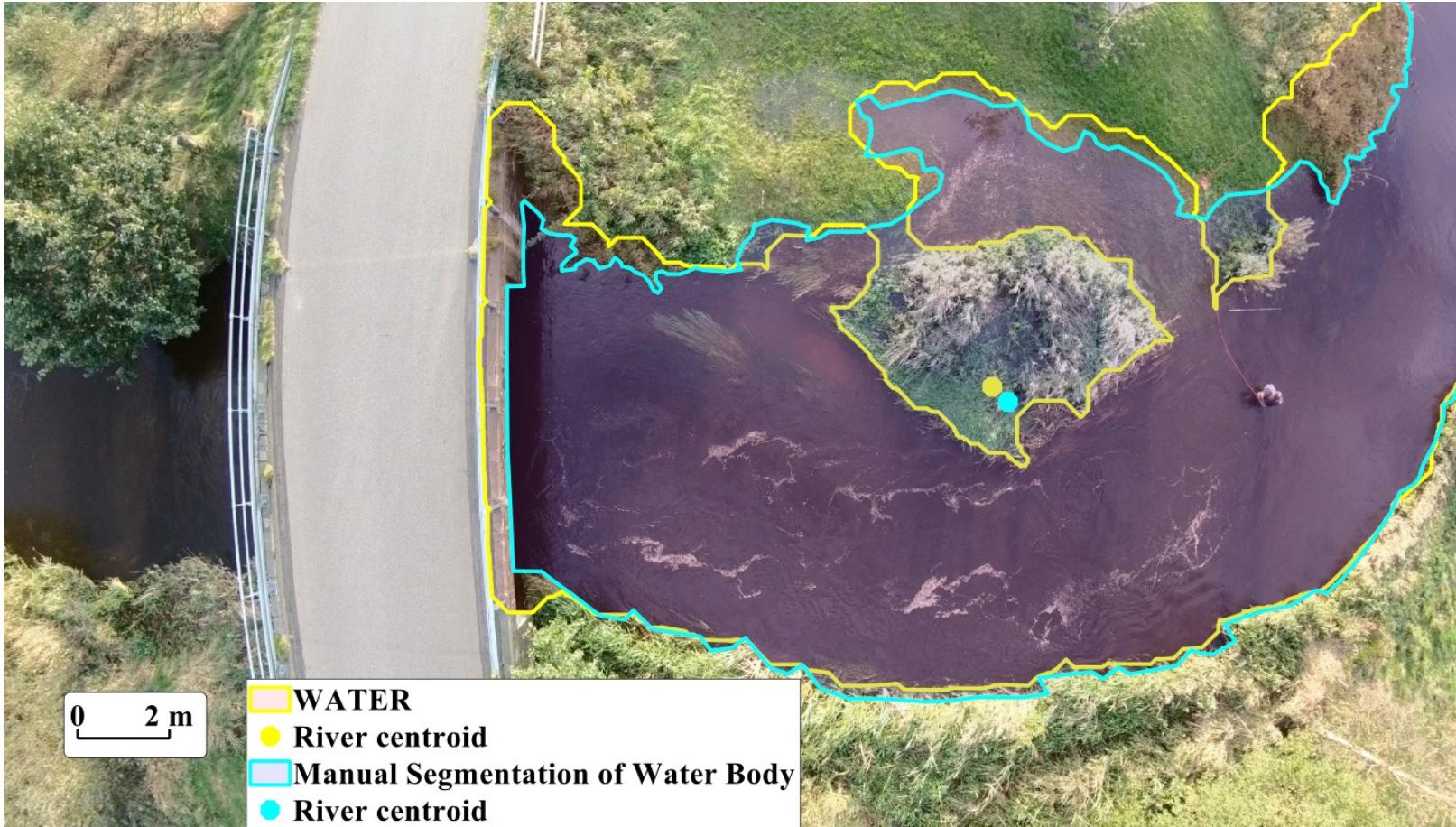
Human (manually-based) vs Machine (automatic)



Water extent segmentation

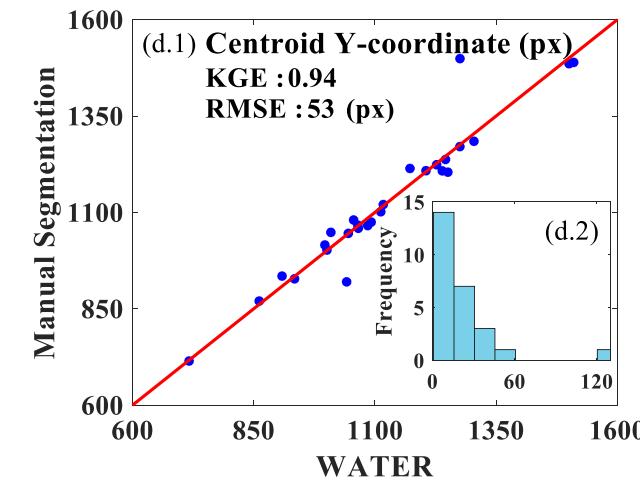
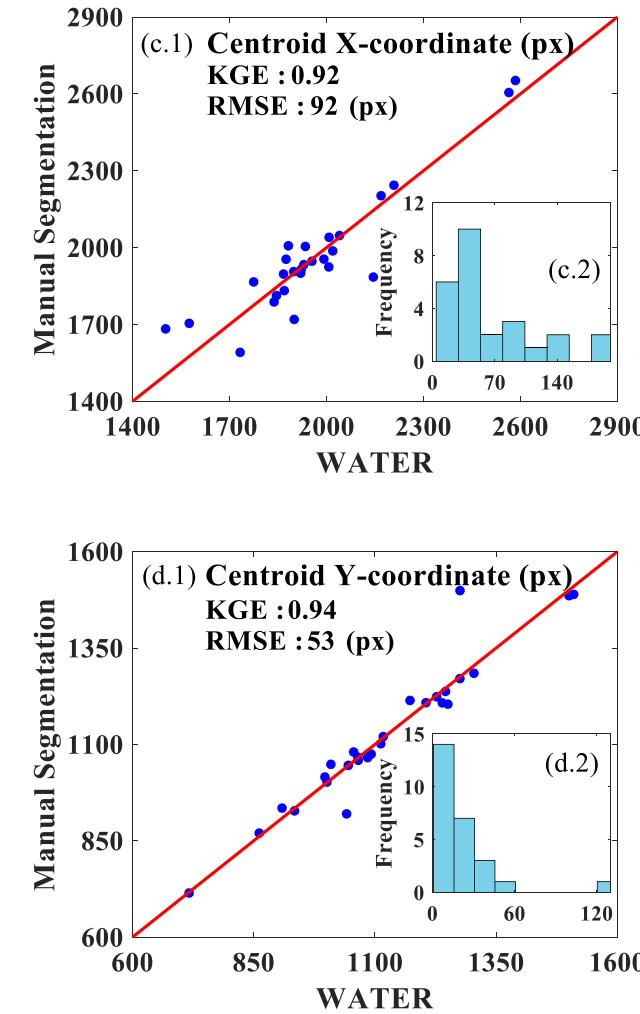
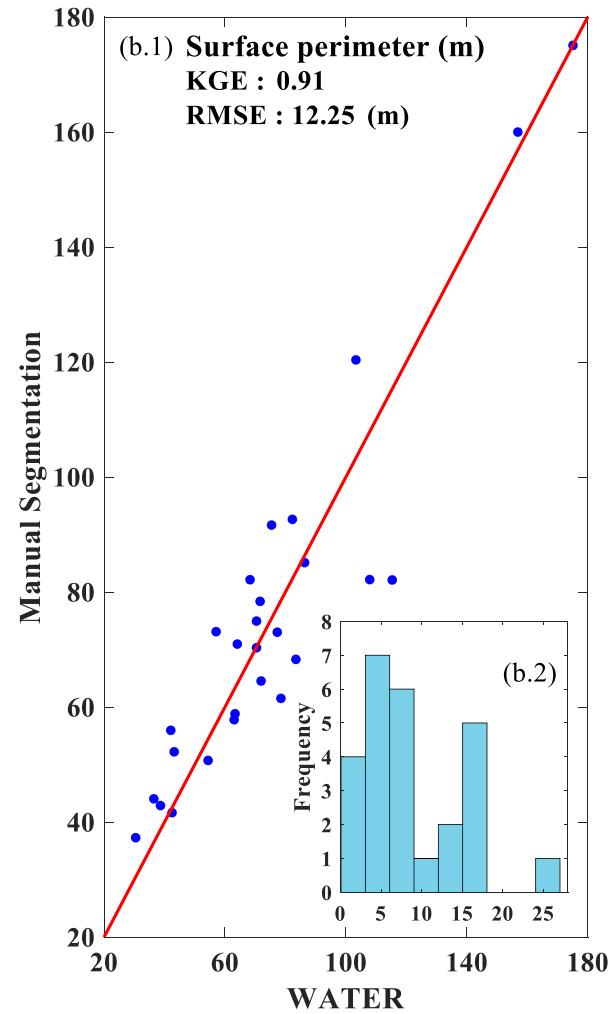
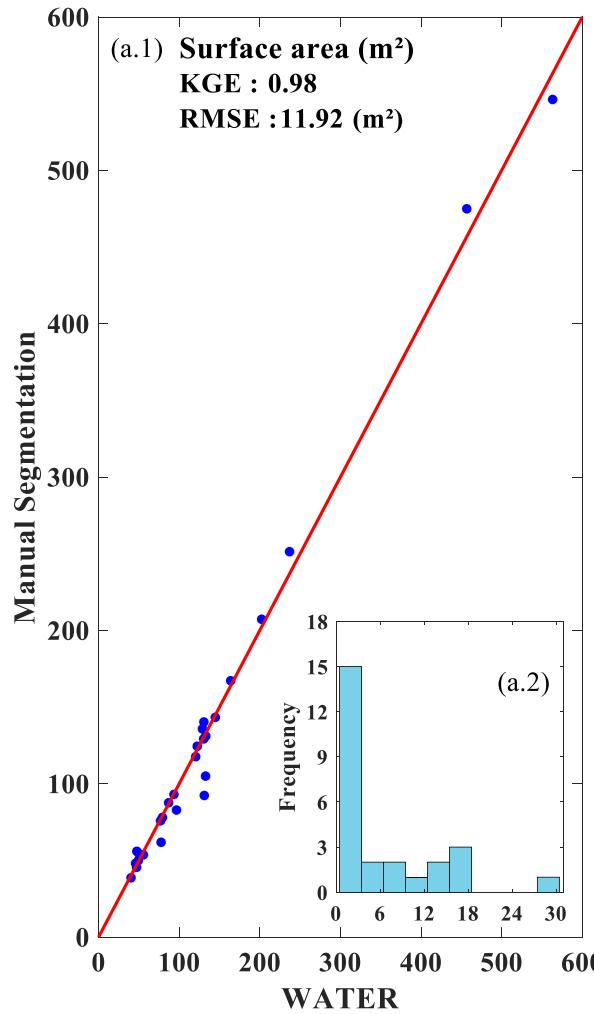
26

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



Water extent segmentation

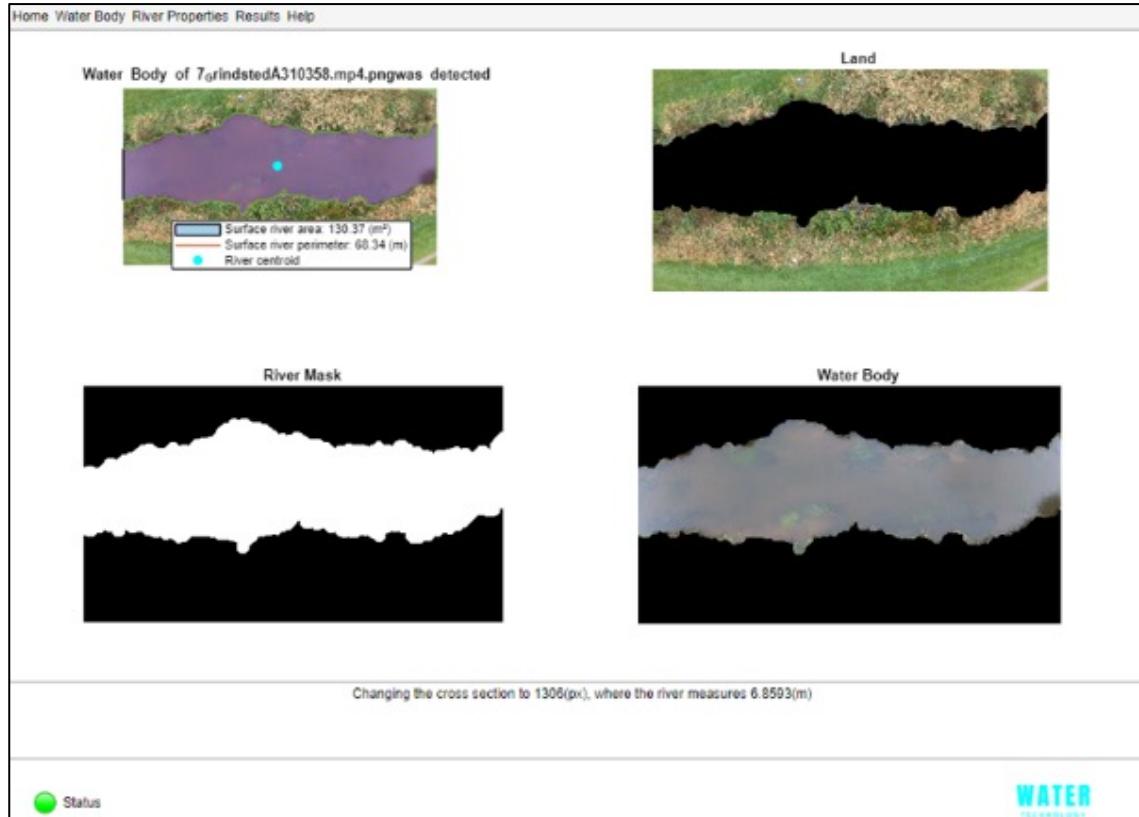
27



Water extent segmentation

28

Software development and Graphical User Interface (GUI)



Main UAS limitations and Future perspectives

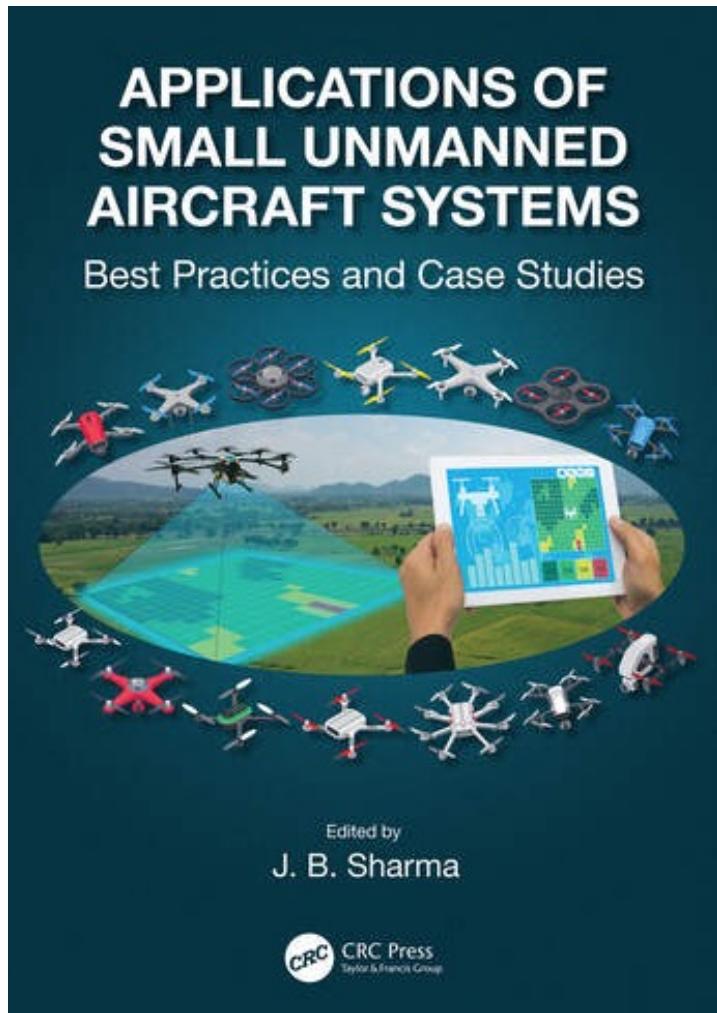
29

- 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



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
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- 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.; Ciraolo, 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.; Ciraolo, 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

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Thanks for your attention!

Questions?

