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IMAGE PROCESSING FOR STUDYING SIZE SEGREGATION IN BEDLOAD TRANSPORT







Hugo Lafaye de Micheaux^{1,2*}, Christophe Ducottet¹ and Philippe Frey²

Communauté **UNIVERSITÉ Grenoble Alpes**

(1) Université de Lyon, UJM-Saint-Etienne, CNRS, UMR5516, Laboratoire Hubert Curien F-42023 Saint-Etienne, France

(2) Univ. Grenoble Alpes, Irstea, UR ETGR, 2 rue de la Papeterie-BP 76, F-38402 St-Martin-d'Hères, France

*E-mail: hugo.lafaye-de-micheaux@irstea.fr

1. Context and objectives

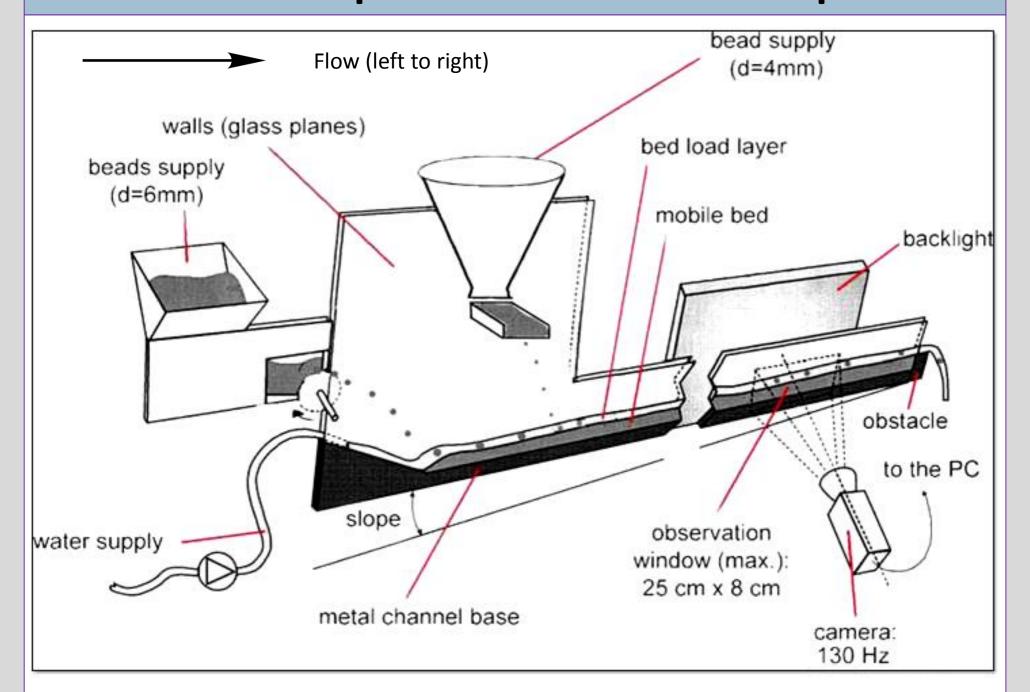
Global aim: studying bedload transport thanks to experiments with two-size beads in a water flow.

L Main objective: track beads over long time sequences to better understand size segregation responsible for complex morphology structures.

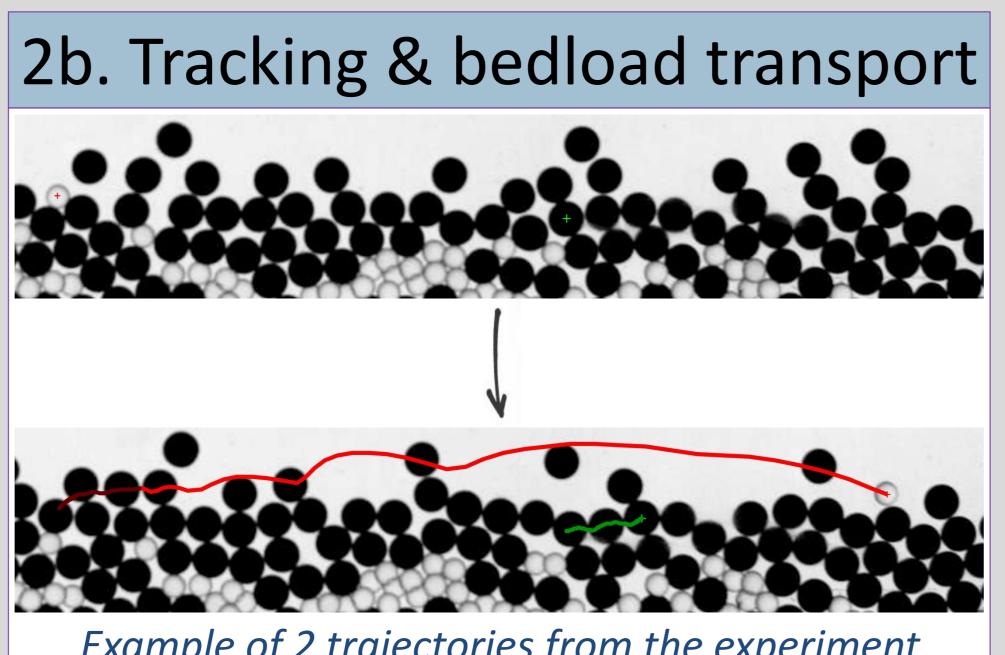
The idea: propose an online particle filter-based tracking algorithm (framework from [2])

- Include adapted multiple motion models with known mechanical dynamics to anticipate bead locations.
- Introduce an observation model from a conditional likelihood to handle detection errors.

2a. Experimental setup



Experimental arrangement sketch (modified from [1])



Example of 2 trajectories from the experiment

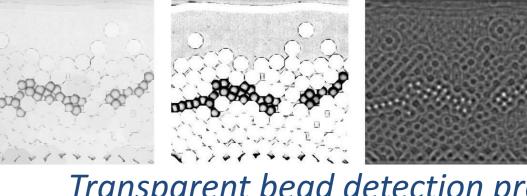
Tracking bedload sediment transport to study:

- Velocities, concentrations, bedload granular rheology
- Size segregation and associated morphology

3. Multi-model tracking algorithm [3]

Stage 1: Object detector

- Use specific morphological operations (erosion, hconvex, cross-corr,...)
- Measure motion states based on neighborhood and velocities
- Return observation state $z_t = \{x_t^{obs}, u_t^{obs}, s_t^{obs}\}$



Transparent bead detection procedure

Stage 2: Data association

Perform greedy algorithm on best matching combinations

Search perimeter depends on motion model

Stage 3: Particle filtering

Objective: Estimate state $c_t = \{x_t, u_t, s_t\}$ of targets according to observations z_t 1 target \Rightarrow cloud of N particles, 1 particle \Rightarrow 1 state $c_t^i = \{x_t^i, u_t^i, s_t^i\}$

Based on a Sequential Monte Carlo approach (SMC) and a Markov Chain:

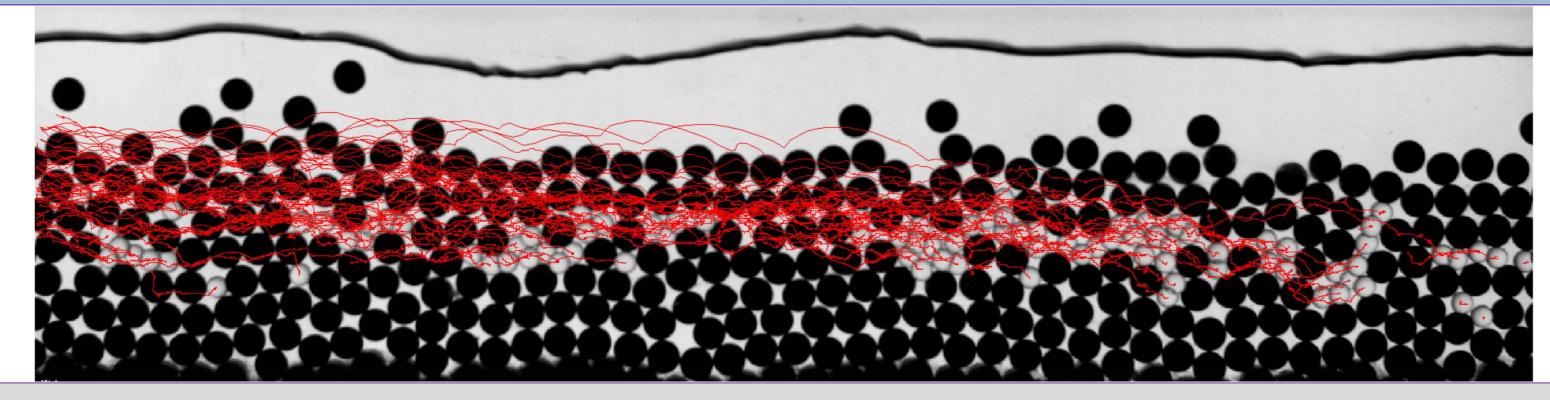
- 1. Prediction: predict target state according to evolution model on particles
- 3 motion models based on bedload dynamics:
 - Resting not moving - $(x, y)_t = (x, y)_{t-1} + \varepsilon_{rest}^{(x,y)}$
 - sliding on others $(x,y)_t = (x,y)_{t-1} + (u,0)_{t-1}\Delta t + \varepsilon_{roll}^{(x,y)}$ Rolling
 - Saltating bouncing on others $(x,y)_t = (x,y)_{t-1} + (u,v)_{t-1} \Delta t + \varepsilon_{salt}^{(x,y)}$

Correction Observation Prediction

2. Correction: correct predicted state thanks to observations

- Particle importance weighting
- $w_t^i = p(\boldsymbol{z}_t | \boldsymbol{c}_t^i) = p(\boldsymbol{x}_t^{oos} | \boldsymbol{x}_t^i)$ vrais. velocity
- Normalize weights and resample particles
- 3. Final target state estimation by averaging resampled particles

4. Example of tracking results



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5. Conclusions & Perspectives

- > New online particle filter-based tracking algorithm based on multiple dynamic models:
 - Input of object mechanical dynamics helps approaching real trajectories.
 - Allows studying bedload transport with high confidence.

Perspectives: apply to long sequences of sediment transport to study high and lower frequency phenomena

References

- [1] Hergault, V., Frey, P., Métivier, F., Barat, C., Ducottet, C., Böhm, T. & Ancey, C. (2010). Image processing for the study of bedload transport of two-size spherical particles in a supercritical flow. Experiments in fluids, 49, 1095-1107.
 - [2] Isard, M. & Blake, A. (1998). A mixed-state condensation tracker with automatic model-switching. *IEEE ICCV*, 107-112.
- [3] Lafaye de Micheaux, H., Ducottet, C. & Frey, P. (2018). Multi-model particle filter-based tracking with switching dynamical state to study bedload transport. Accepted by journal Machine Vision and Applications, Springer.