

Monitoring and modelling of aquatic environments how is artificial intelligence used ?

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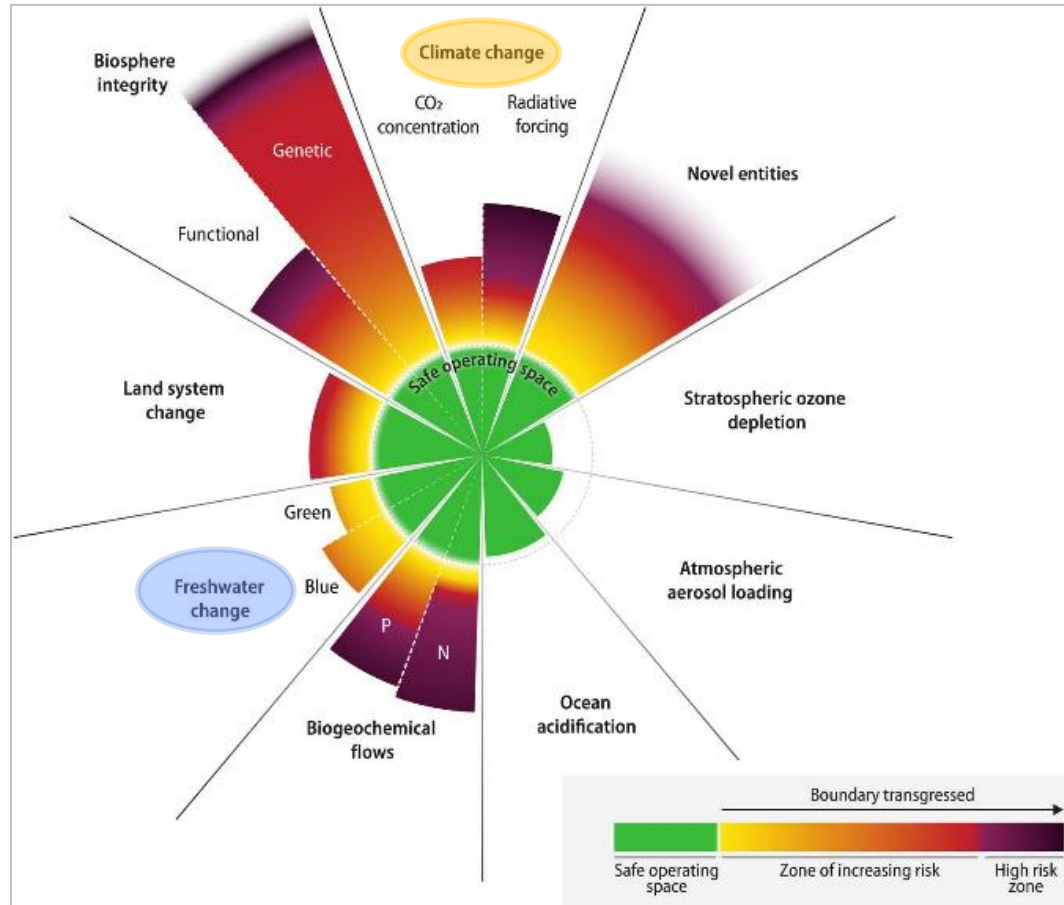
LEESU

Ecole nationale des ponts et chaussées

- Water, Environment & Urban systems
- Urban hydrology
- Research fields
 - *Monitoring and modelling of inland waters*
 - Alternative solutions for urban water management
 - Nature-based solutions

Inland waters

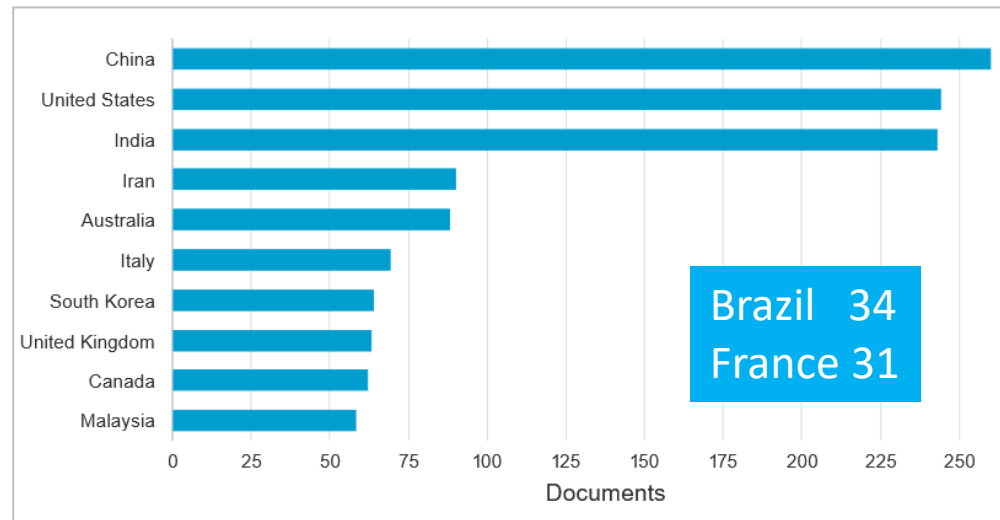
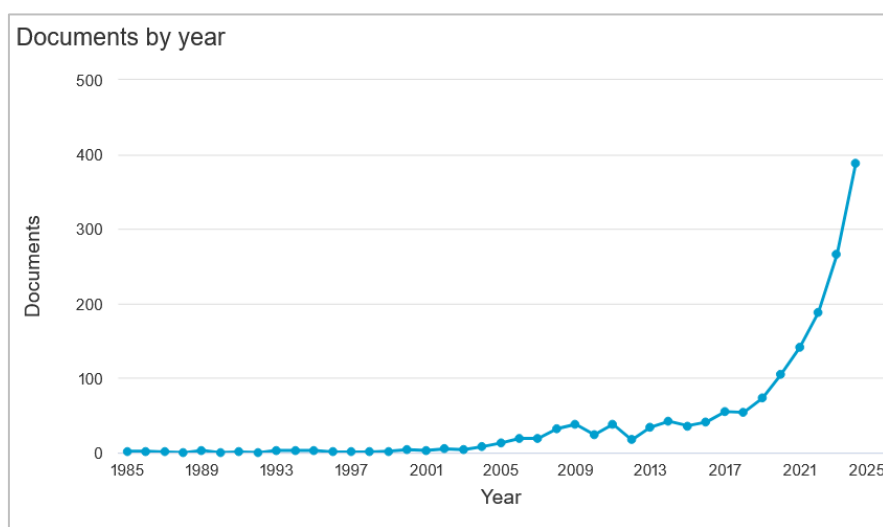
- Planetary limits
- Freshwaters & Climate change



Rockström et al., (2009). A safe operating space for humanity. *Nature*, <https://doi.org/10.1038/461472a>

Bibliographic analysis

- Scopus
- ‘Artificial Intelligence’
AND hydrology OR limnology OR 'water AND quality
AND monitoring' OR 'modelling’
- 1989 – 2024 : 1670 documents



Artificial Intelligence and monitoring and modelling aquatic environments

- Big data
- Rare long-term timeseries on aquatic environments
- Increasing use of online sensors and measuring devices

2 examples

- Microbiological contamination of urban rivers : the Seine river in Paris
- Blooms of toxic species in urban lakes : Lake Créteil

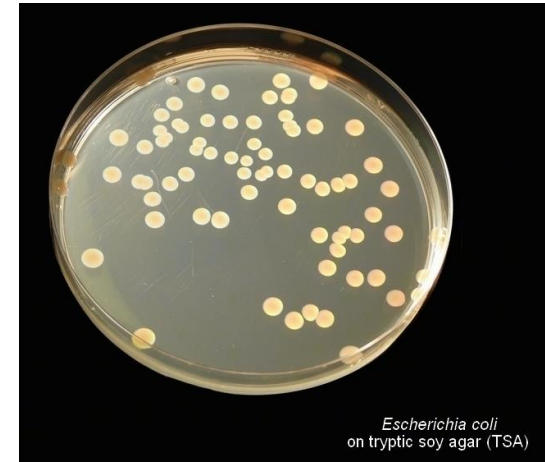
Microbiological contamination of rivers

- Heavy rainfall episodes
 - Fecal contamination of the river
 - *Overpass of the regulatory threshold?*
-
- Triathlon in the River Seine
 - *Will the competition take place?*



Forecast of microbiological contamination

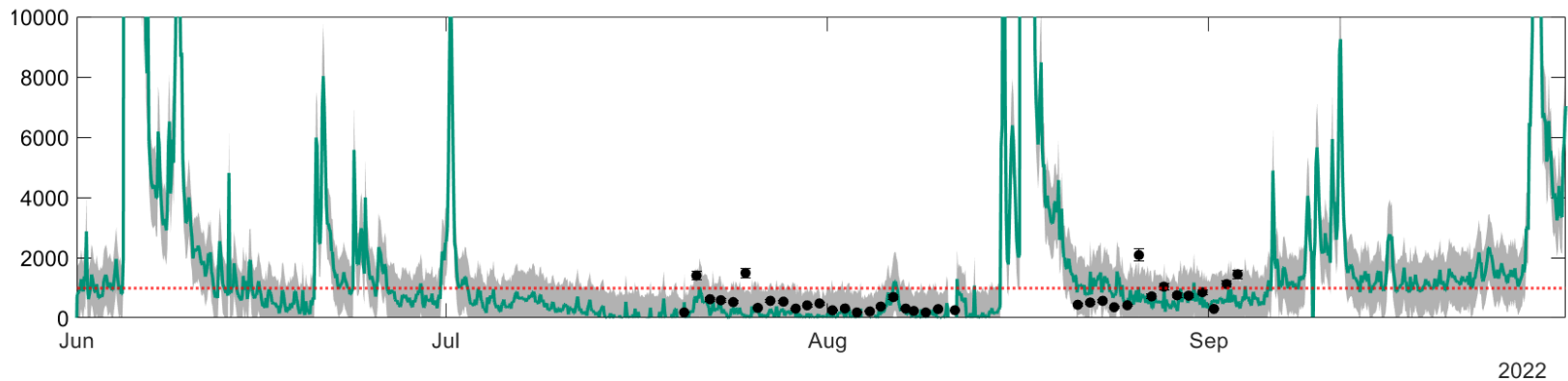
- *E.coli* ∈ fecal indicator bacteria
- Lab analysis $\geq 12\text{h} - 24\text{h}$
- New monitoring devices
- High-frequency monitoring
- Forecast based on hydrological and weather conditions



High-frequency monitoring of *E.coli* in the Seine

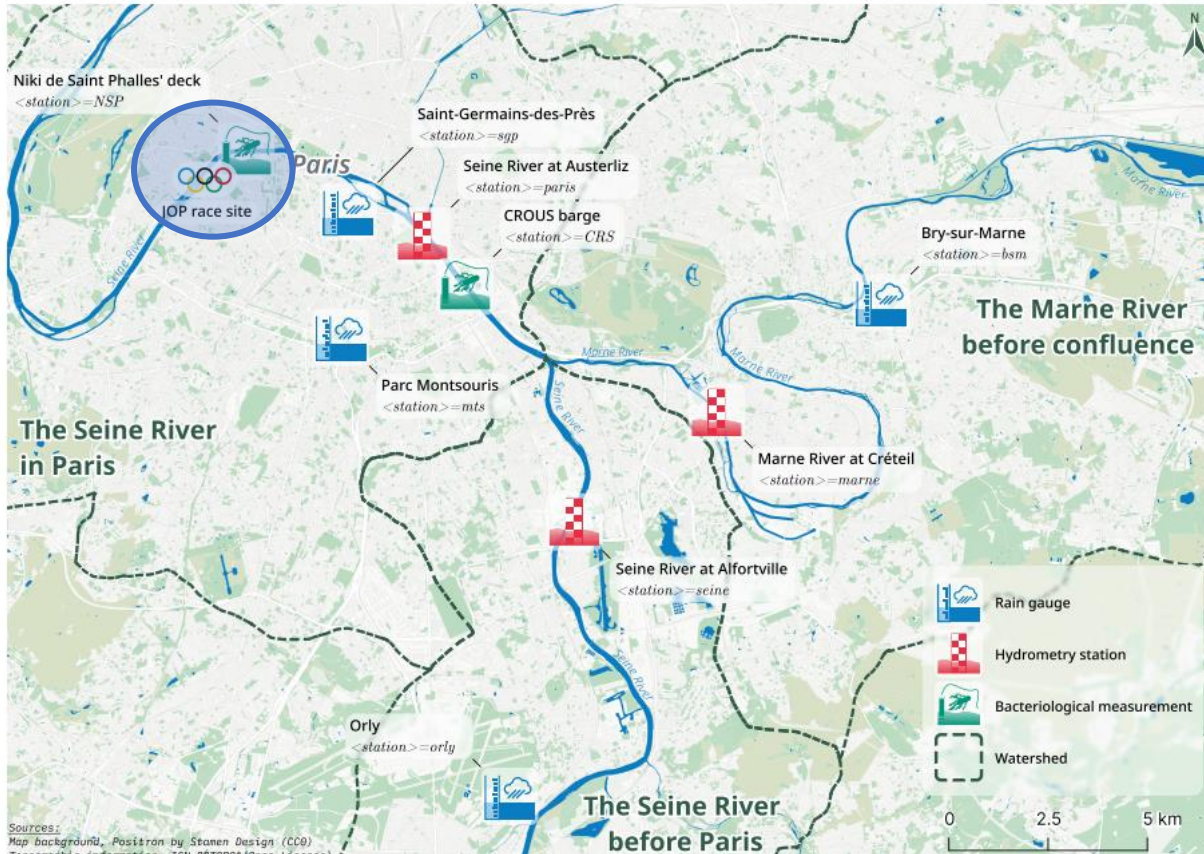


Arthur Guillot-Le Goff
Rémi Carmigniani
ENPC, LEESU & LHSV



$\Delta t = 2h$

Rain and hydrological data



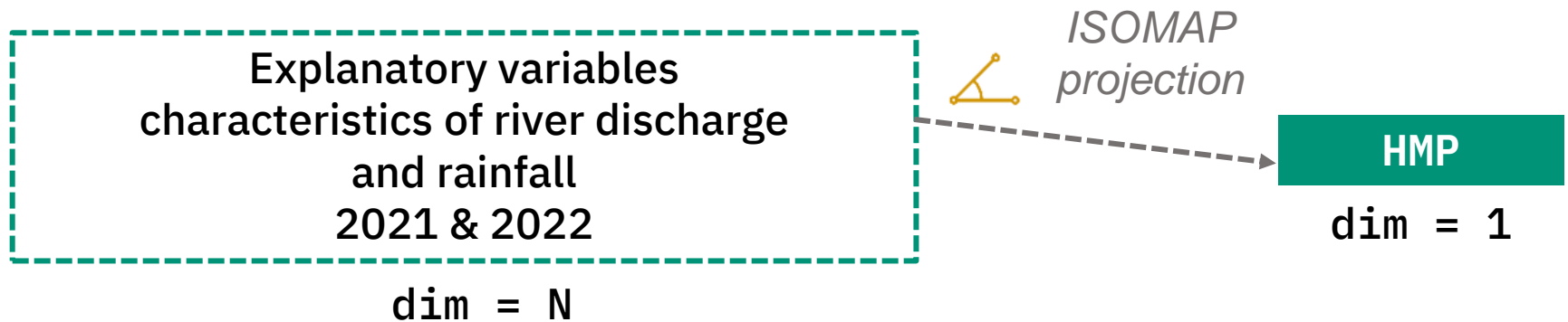
Explanatory variables of FIB
→ Meteo & Hydro

- **SEINE in PARIS**
 - Rain : *Saint-Germains-des-Près; Montsouris*
 - Hydro : *Pont d'Austerlitz*
- **SEINE Upstream**
 - Rain : *Orly*
 - Hydro : *Alfortville*
- **MARNE**
 - Rain : *Bry-sur-Marne*
 - Hydro : *Créteil*

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Single hydro-meteorological indicator of the rainfall events

- Reduction of the dataset dimension
- Non-linear method: ISOMAP
- Global indicator
 - Global hydro-meteo parameter (HMP)

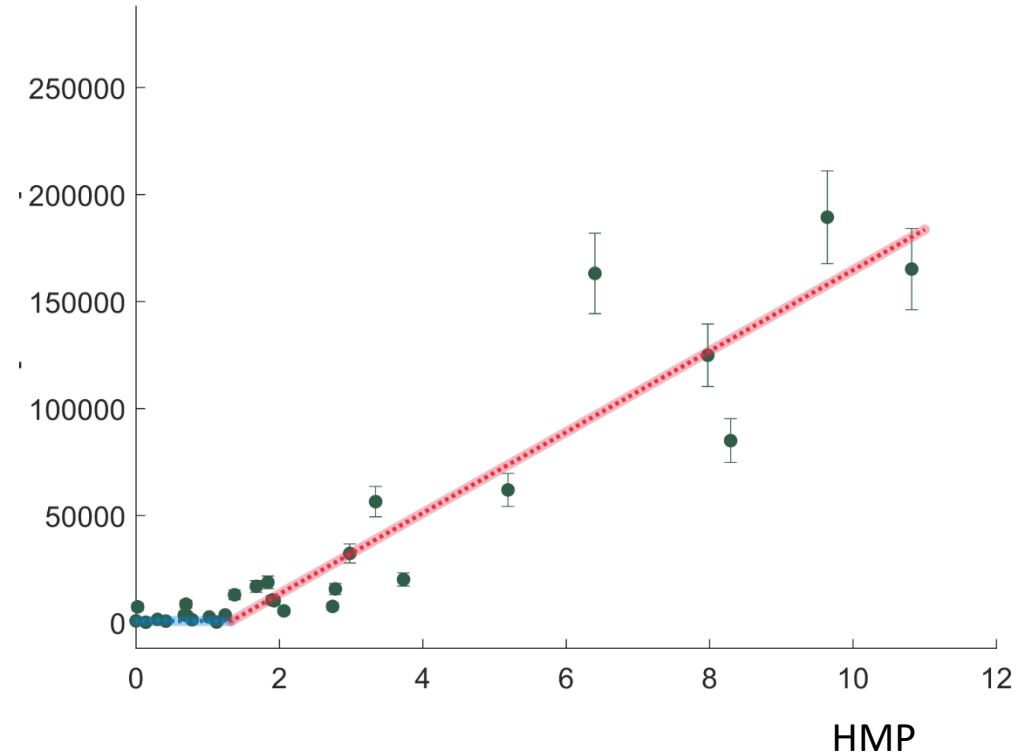


Relationship between HMP and the contamination level

Threshold effect

$HMP \leq \text{threshold} \rightarrow$ no contamination

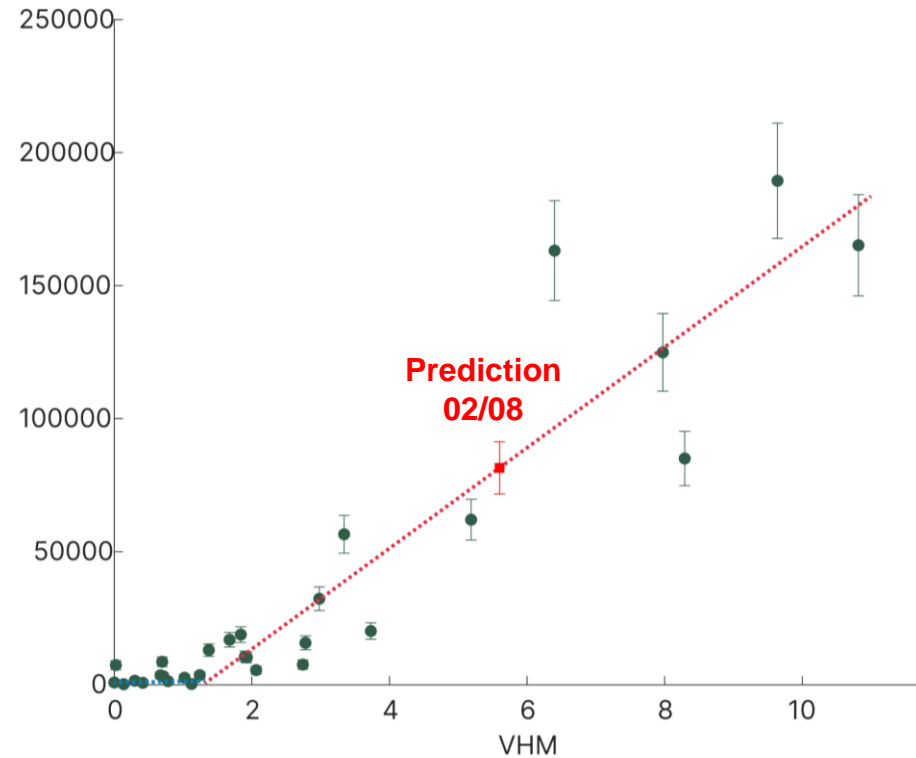
$HMP > \text{threshold} \rightarrow$ increasing contamination





Triathlon test event - August 2nd 2023

- ISOMAP using the 2021 and 2022 dataset
- HMP value on 02/08/2023

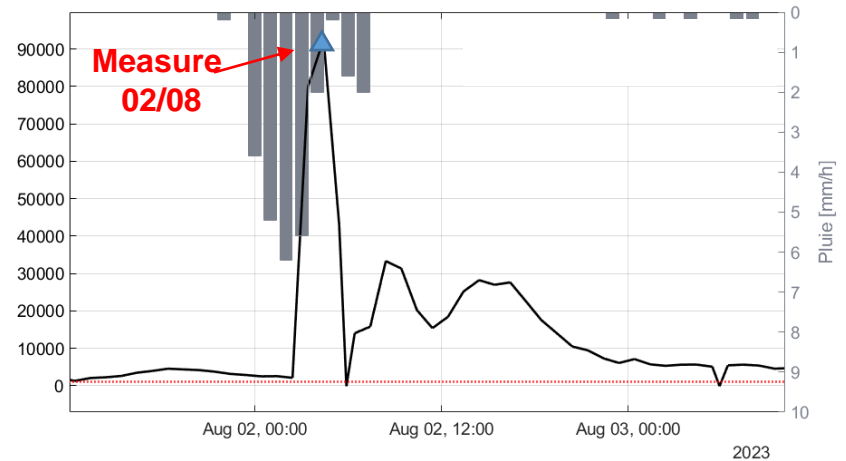
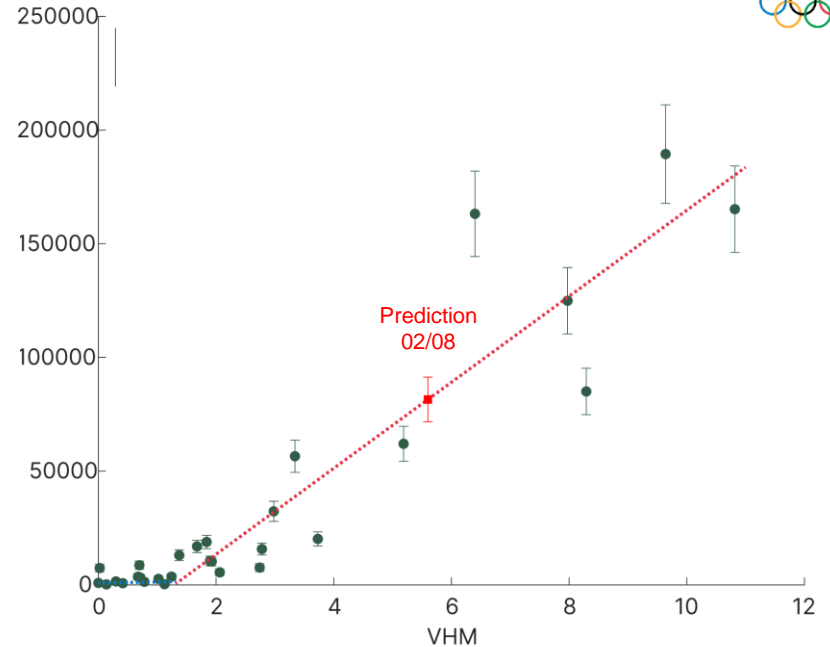




PARIS 2024

Triathlon test event - August 2nd 2023

- ISOMAP using the 2021 and 2022 dataset
- HMP value on 02/08/2023

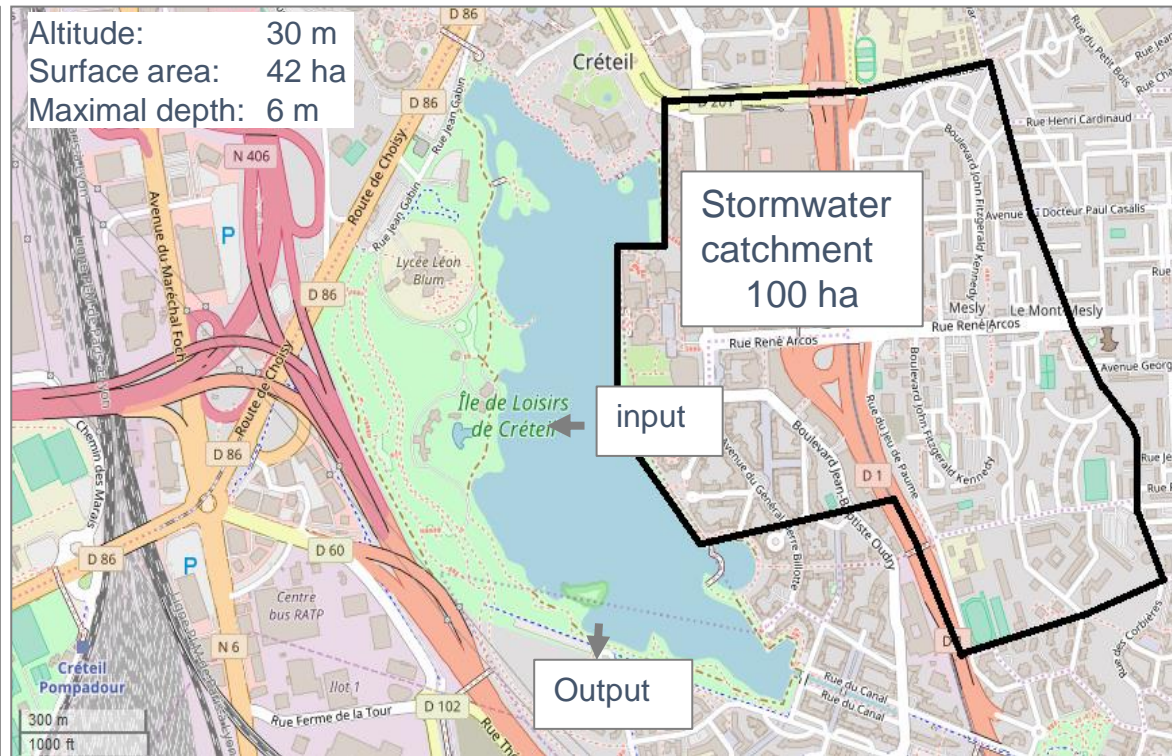
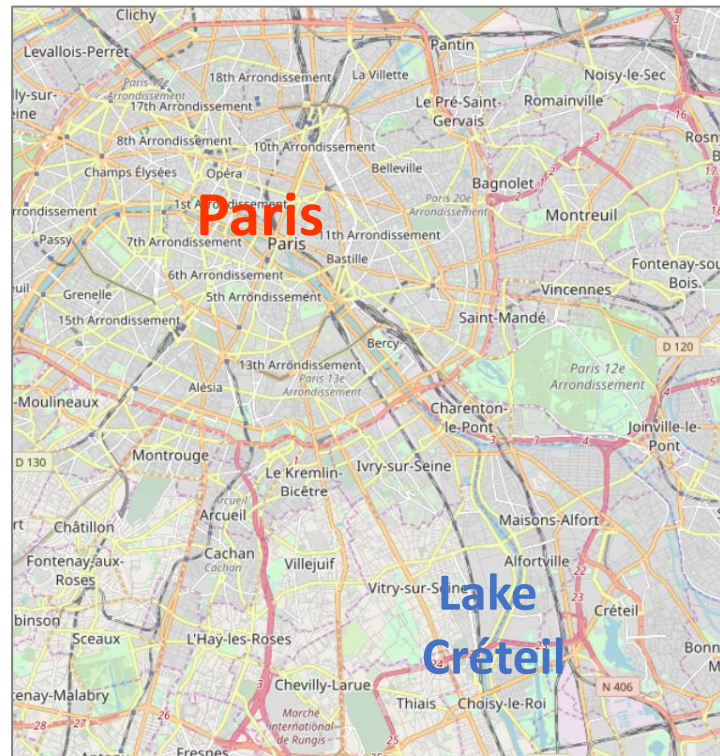


2-year timeseries
Correct prediction of *E.coli*
maximum concentration



Urban lakes
Bloom of potentially toxic cyanobacteria

Lake Créteil



- South-East of Paris
- Typical sandpit urban lake
- Recreational activities
- Cyanobacteria blooms

Lake Créteil monitoring station

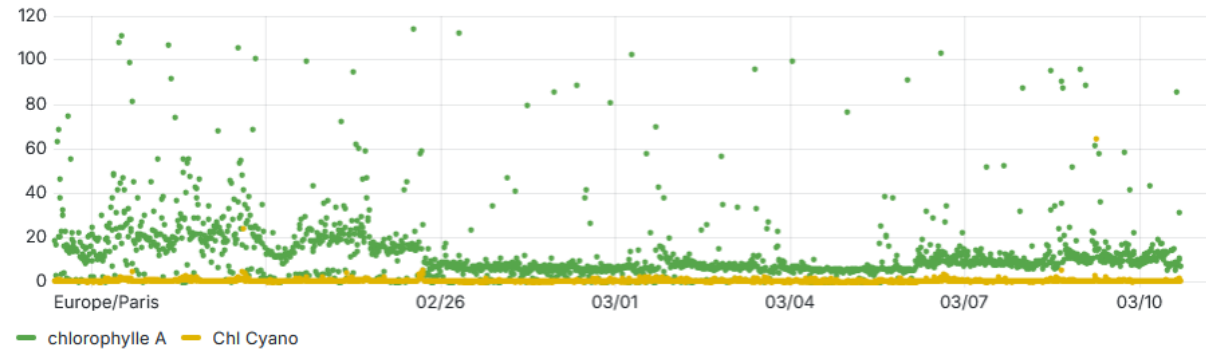
- Meteorological station
- Underwater sensors
 - Water temperature
 - Conductivity
 - O₂
 - Chlorophyll
 - Phycocyanin



Field timeseries

Station de mesure du lac de Créteil

Chlorophylle A & Phycocyanine (Cyanobactéries)



Oxygène dissous (% sat)



Température de l'eau (°C) à 1.5m de profondeur

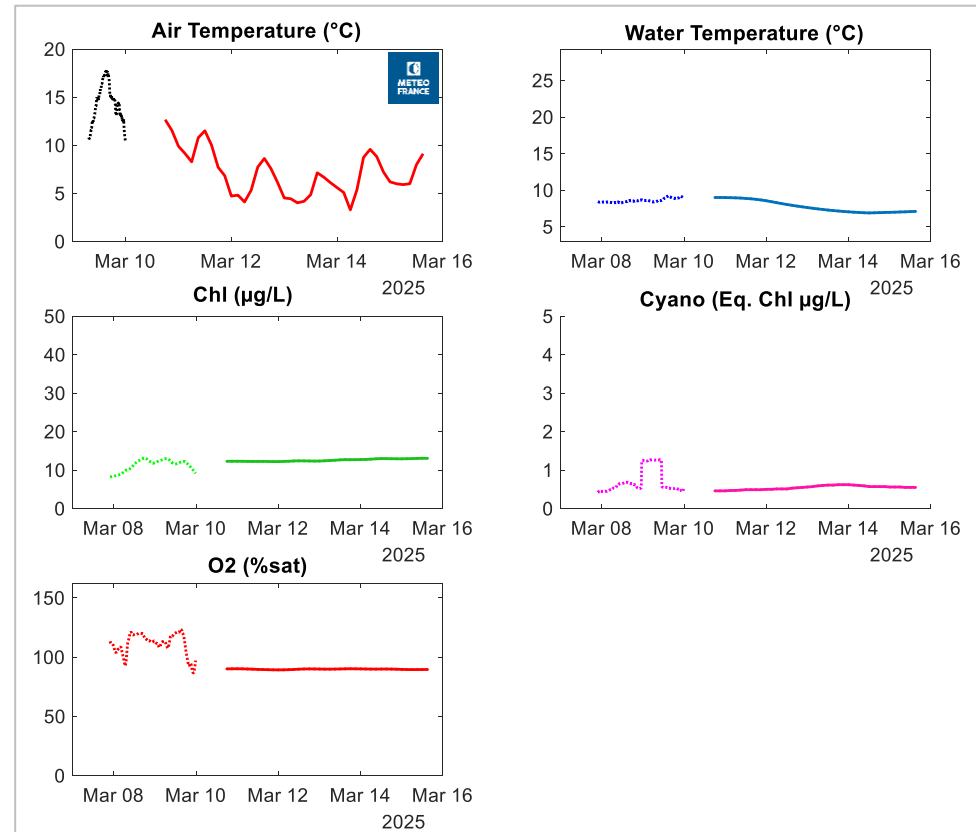


Température de l'air (°C)

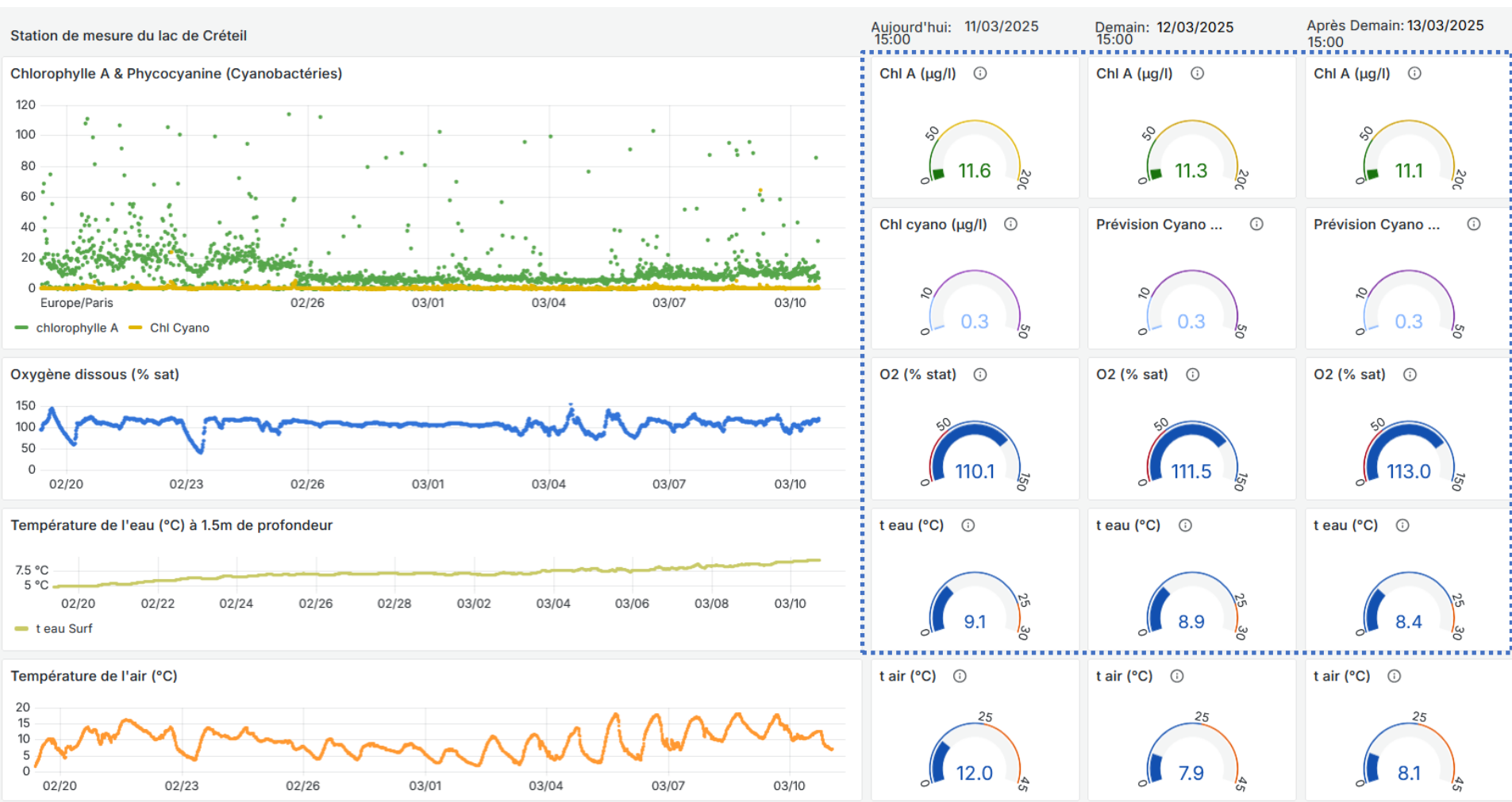


Short-term forecast of Chlorophyll and Cyanobacteria

- Non-linear autoregressive neural network
- Field data
 - Chlorophyll
 - Phycocyanin
 - Water temperature
 - Oxygen
- Meteorological forecast
 - Air temperature



Field timeseries and prediction



Physic-informed neural ODE

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For a variable of interest $u(t)$ and an external forcing $g(t)$ we can write the following ODE system,

$$\frac{du}{dt}(t) = f_{\phi}^{(\text{PHY})}(u(t), g(t)) + f_{\theta}^{(\text{NN})}(u(t), g(t)),$$

where $f_{\phi}^{(\text{PHY})}$ is the known physics model with parameters ϕ and $f_{\theta}^{(\text{NN})}$ is a Neural Network with parameters θ .

- Physics constraints
- Require less data

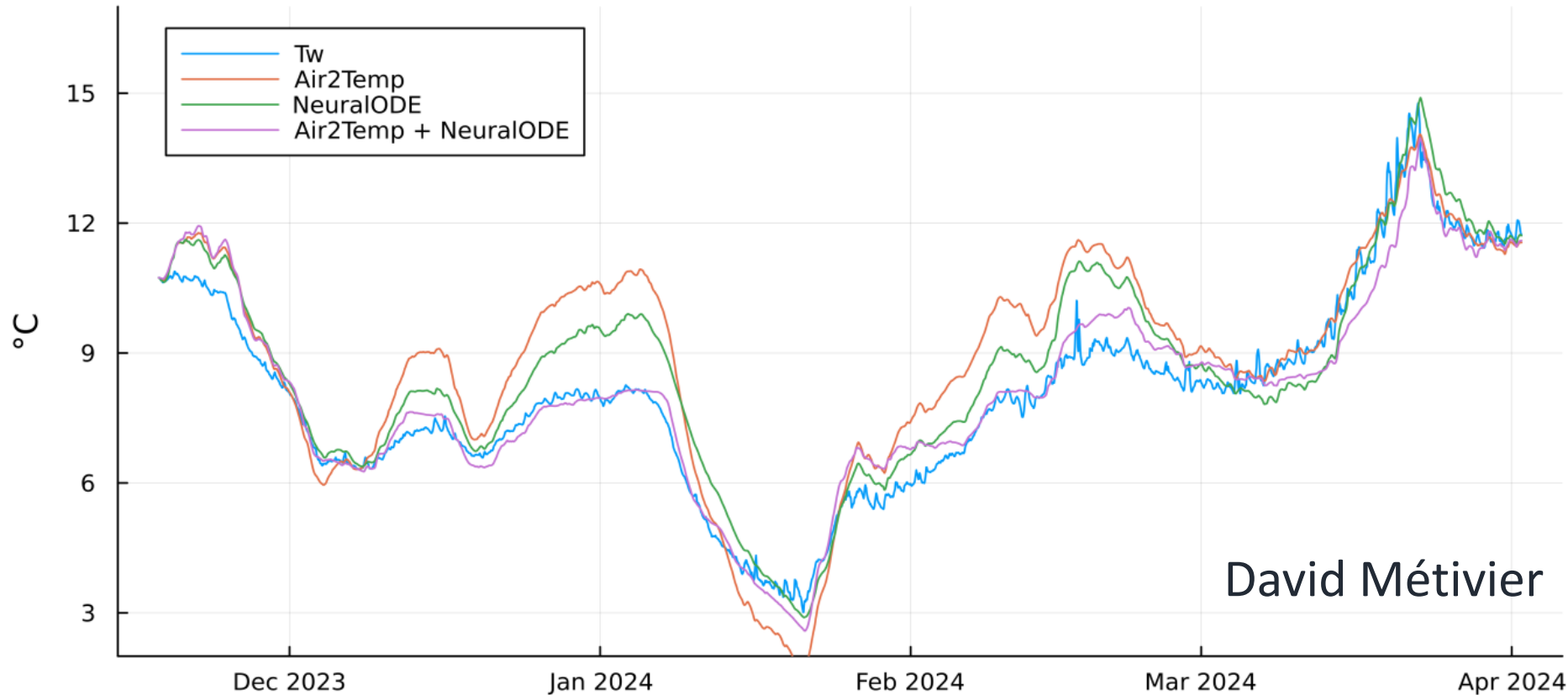
Application to lake water temperature

$u(t) = T_{\text{water}}(t)$ and the external forcing $g(t) = T_{\text{air}}(t)$.

We use the **Air2Temp** 4-parameters model (Piccolroaz et al. 2013)

$$f_{\phi}^{(\text{PHY})}(t) = \left(\phi_1 + \phi_2 T_{\text{air}}(t) - T_{\text{water}}(t) + \phi_3 T_{\text{water}}(t) \right) e^{(T_{\text{water}}(t) - T_{\text{ref}}) / \phi_4}$$

$f_{\theta}^{(\text{NN})}(T_{\text{water}}, T_{\text{air}}) = \text{Simple Dense NN}$ typically one or two layers of 16 neurons.



David Métivier

Thank you for your attention!
Obrigada!

