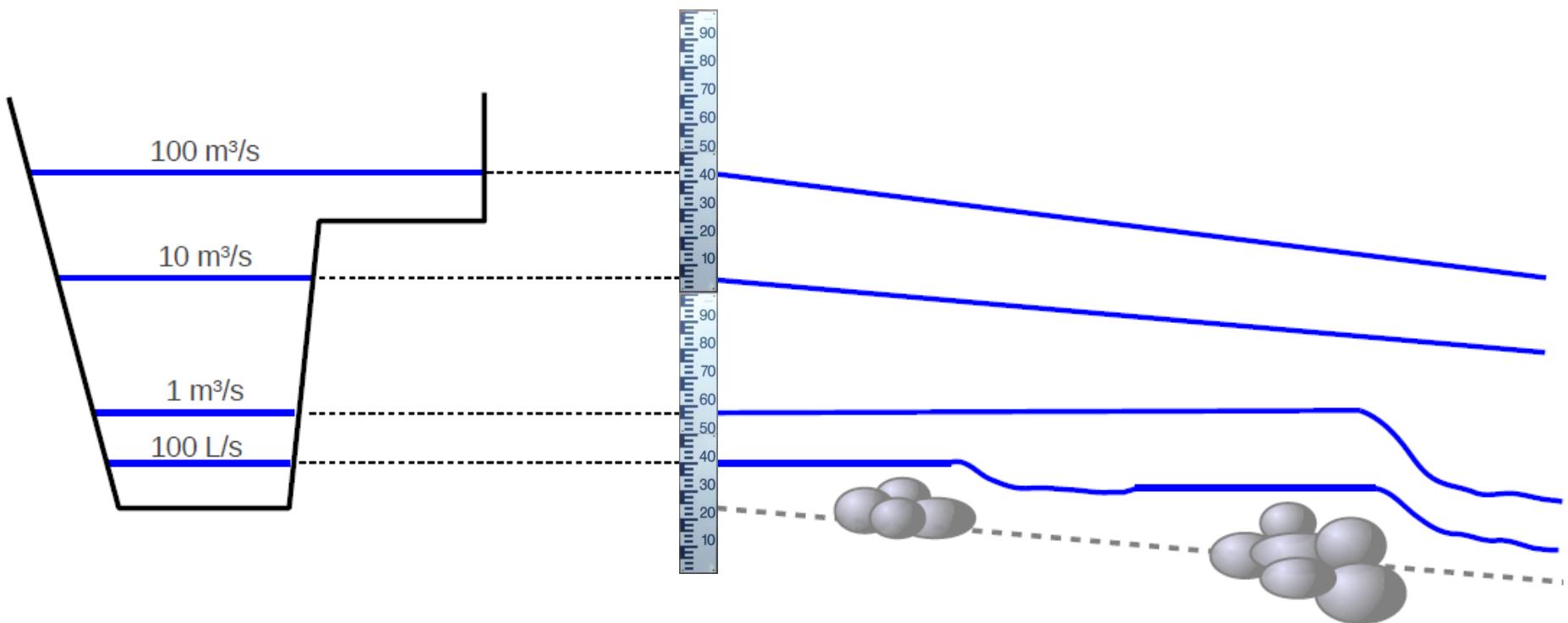


The Bayesian estimation of rating curves: principles of the BaRatin method



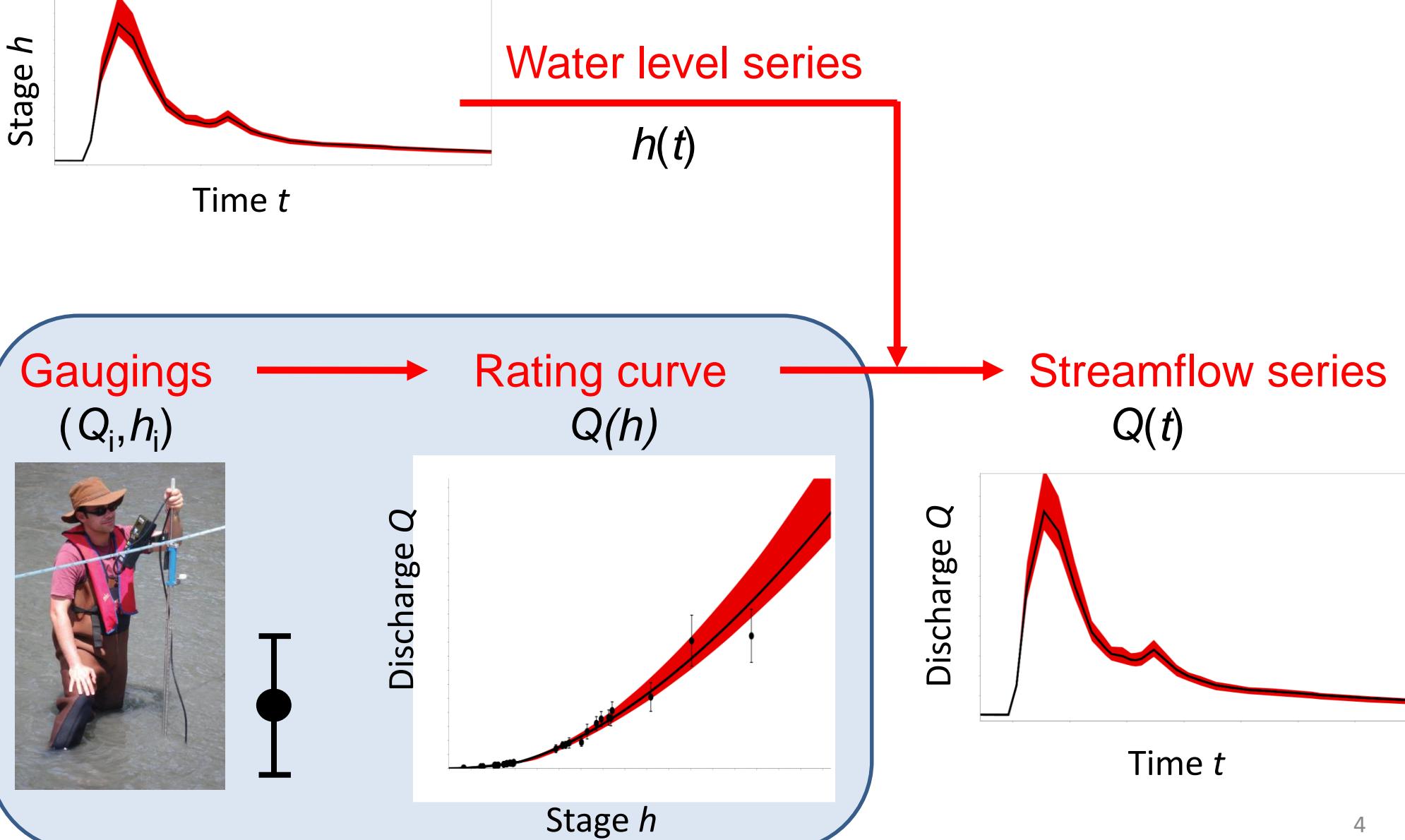
The BaRatin method for rating curves

- ✓ Introduction
- ✓ Hydraulic principles behind the rating curve
- ✓ Rating curve estimation
- ✓ Going further

The BaRatin method for rating curves

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- ✓ Hydraulic principles behind the rating curve
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Establishing probabilistic streamflow data



BaRatin (Bayesian rating curves)

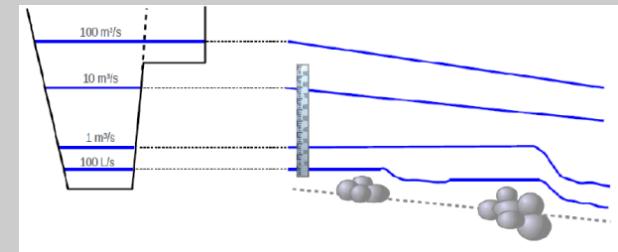
Develop a practical calibration technique to:

- Combine data (measurements) and hydraulic knowledge
- Make expert knowledge and assumptions easier to defend and review
- Account for data and hydraulics uncertainties
- Provide discharge uncertainties

DATA
MEASUREMENTS
STATISTICS



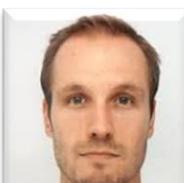
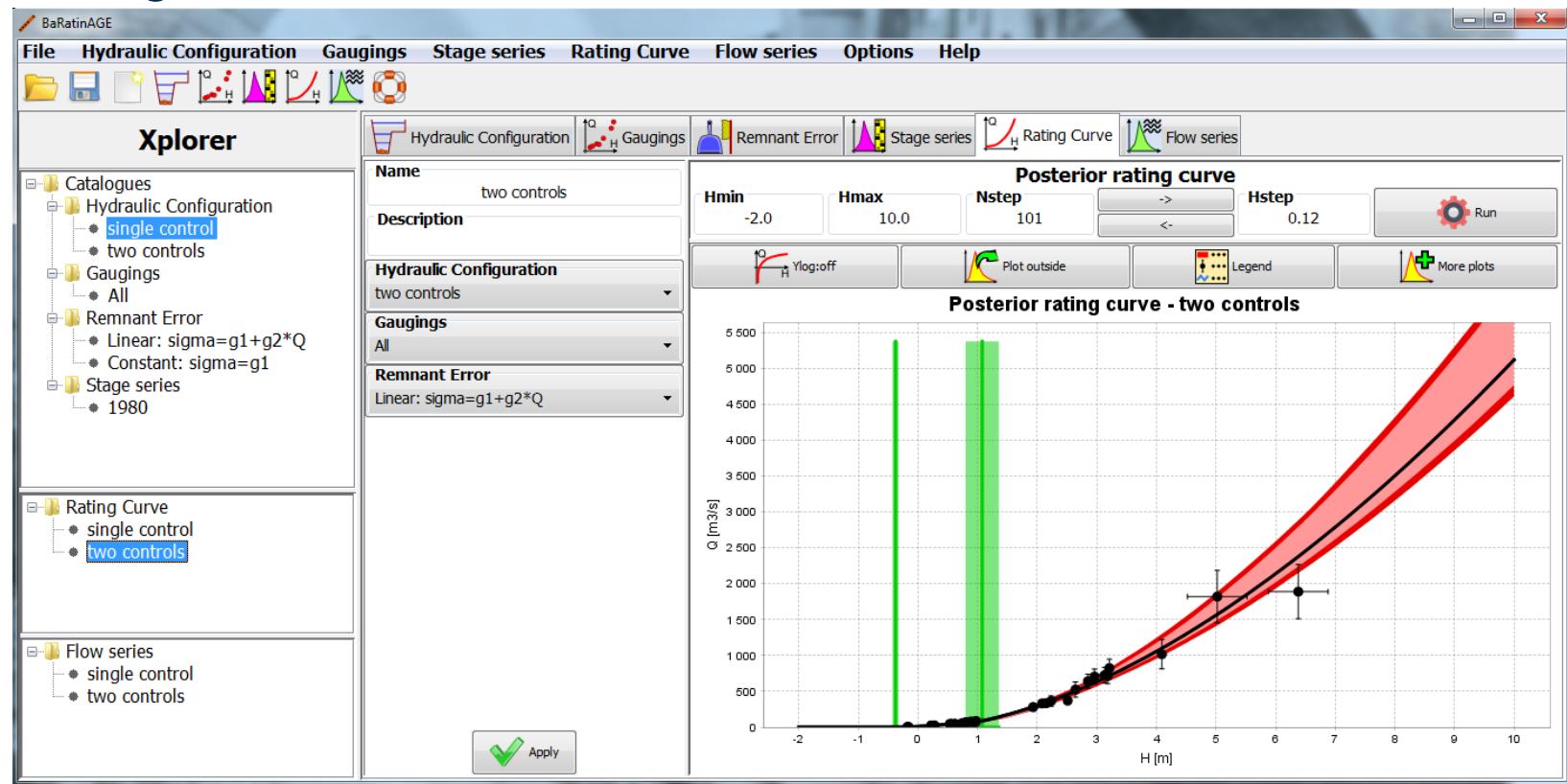
MODEL
CONTROLS
HYDRAULICS



BaRatinAGE software



- Graphical interface (Java) and user manual
- Freely available in French, English and other languages
- Open-source (GPL3), codes available on GitHub
- ~200 registered users

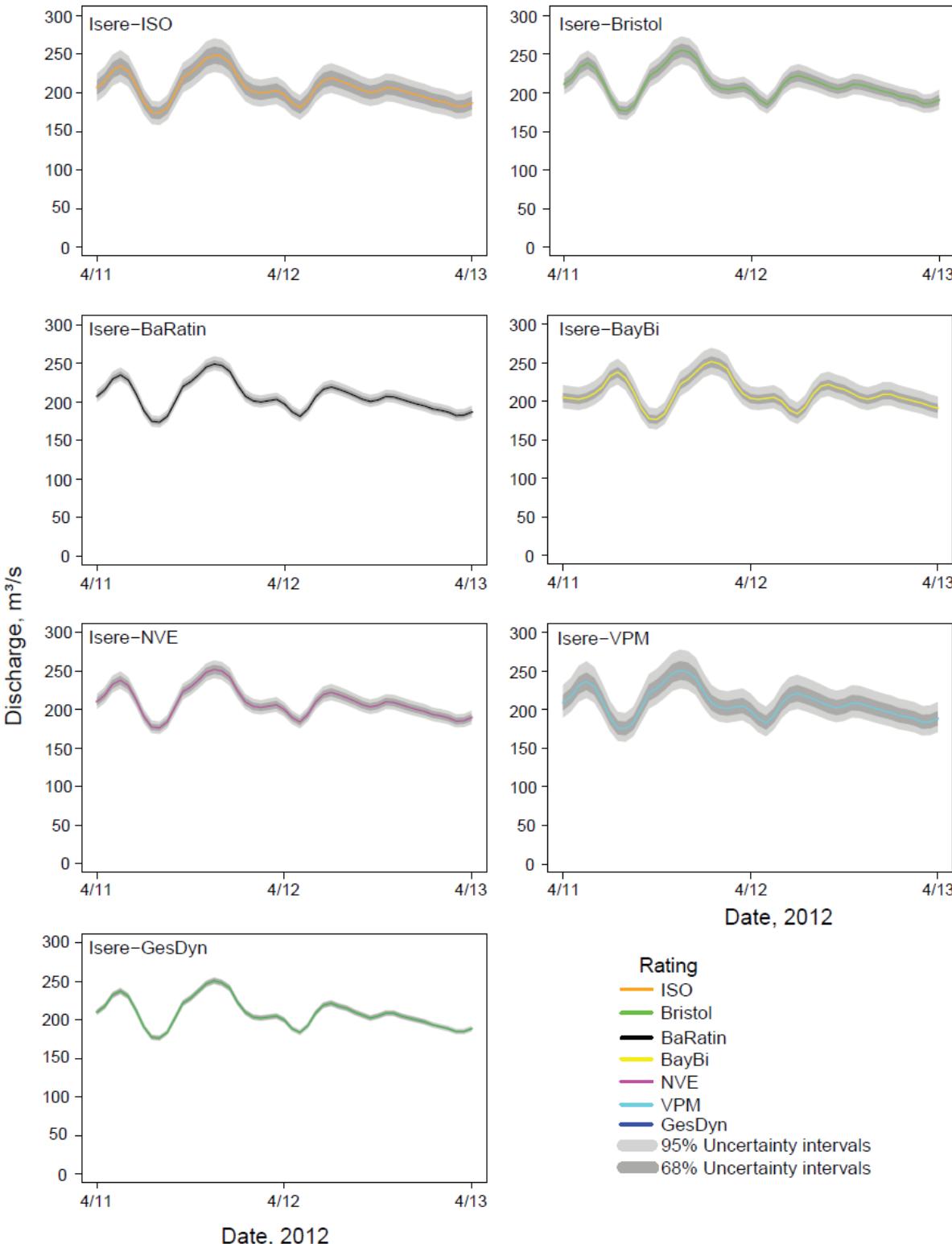


Rating curve uncertainty methods

New methods have been developed in the past decade to tackle practical and theoretical issues of the existing ISO/WMO method (which no hydrological service use routinely).

BaRatin is among the 3 that are used operationally, and likely the most widely released.

Kiang et al. (2018)
Comparison of 7 methods for rating curve uncertainty



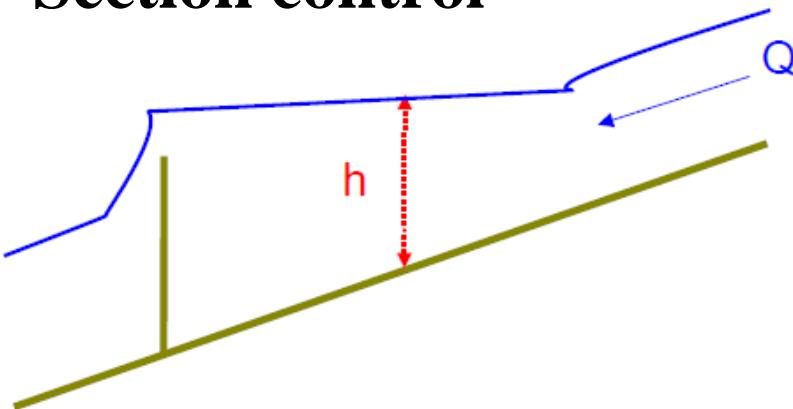
The BaRatin method for rating curves

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Hydraulic controls

- Physical properties of a channel which determine the relationship between stage and discharge at a location in the channel
(World Meteorological Organization, 2012)

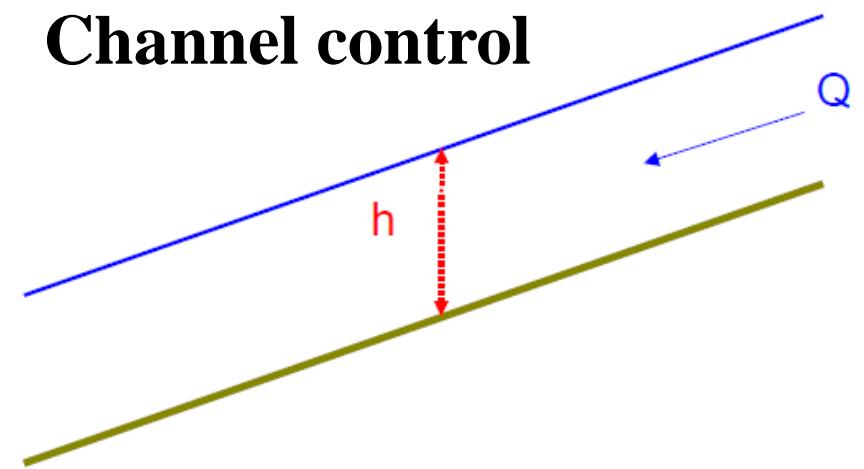
Section control



Fall (critical flow: choked flow)

Upstream water level ~ horizontal
« Emptying bucket »

Channel control



No fall (friction-dominated flow)

Water level ~ parallel to riverbed

Section controls

*Mahurangi,
New Zealand*



Le Golo, France



Le Gapeau, France



L'Ardèche, France



Channel controls

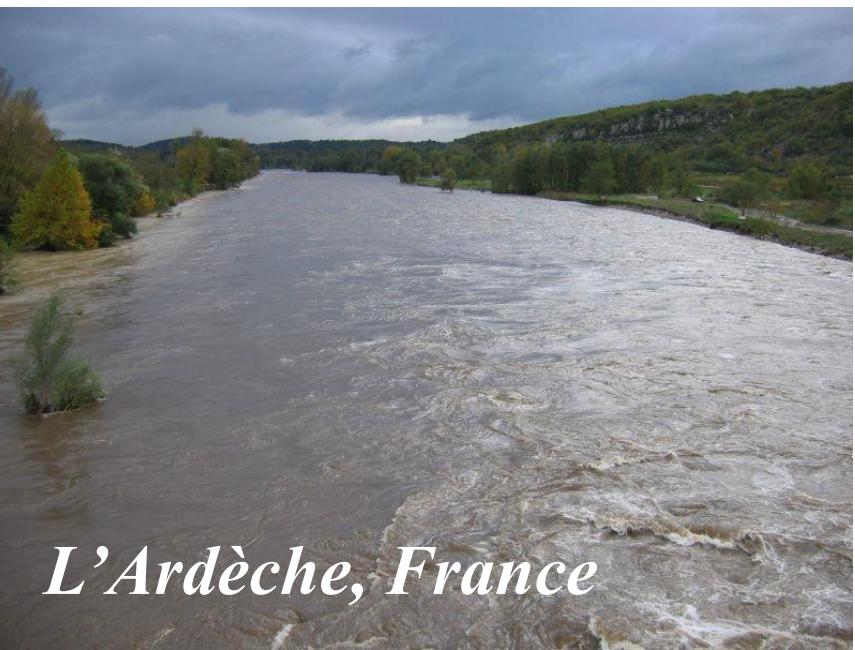


L'Isère, France

25/03/2009



La Sarre, France



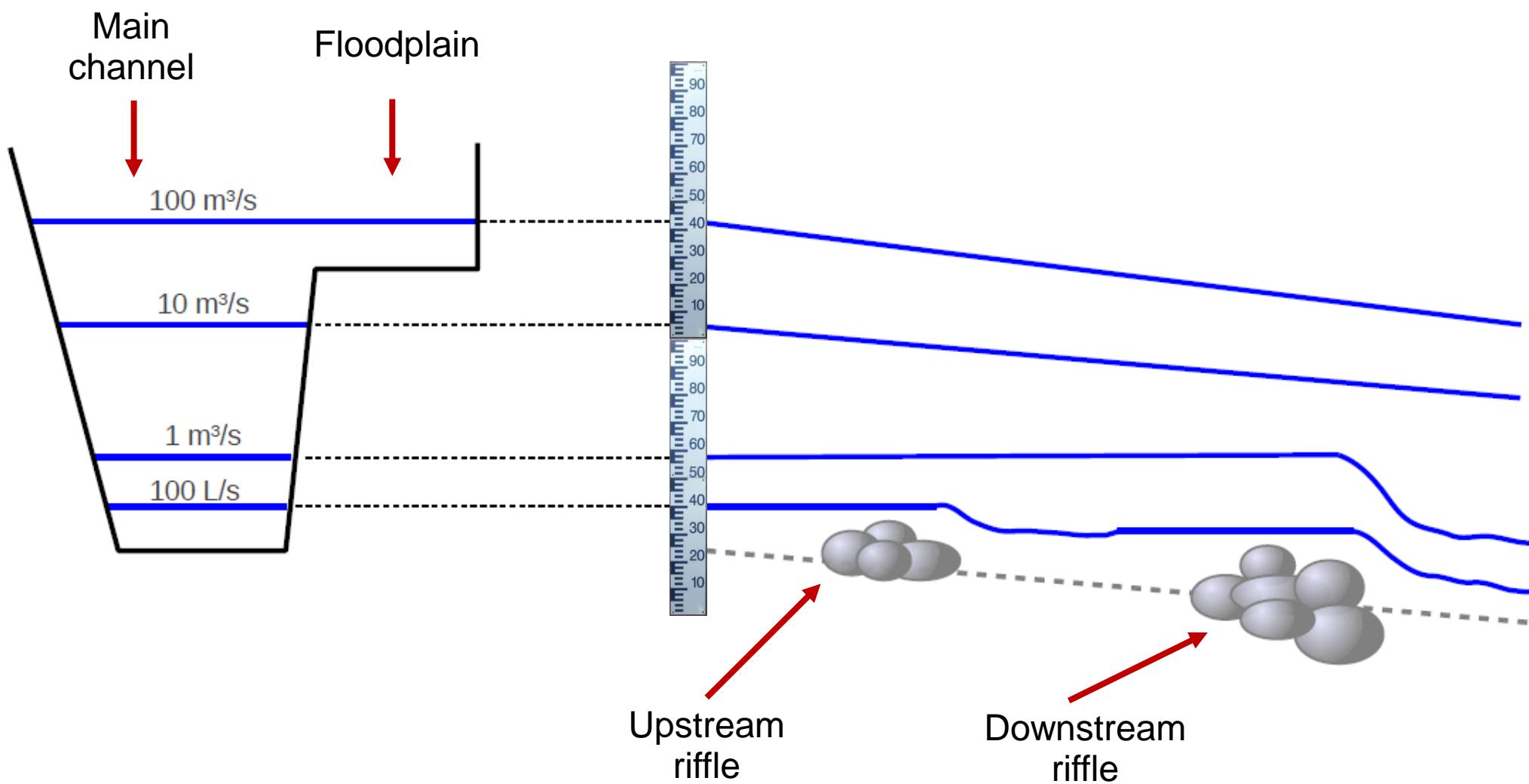
L'Ardèche, France



La Saône, France

Hydraulic controls

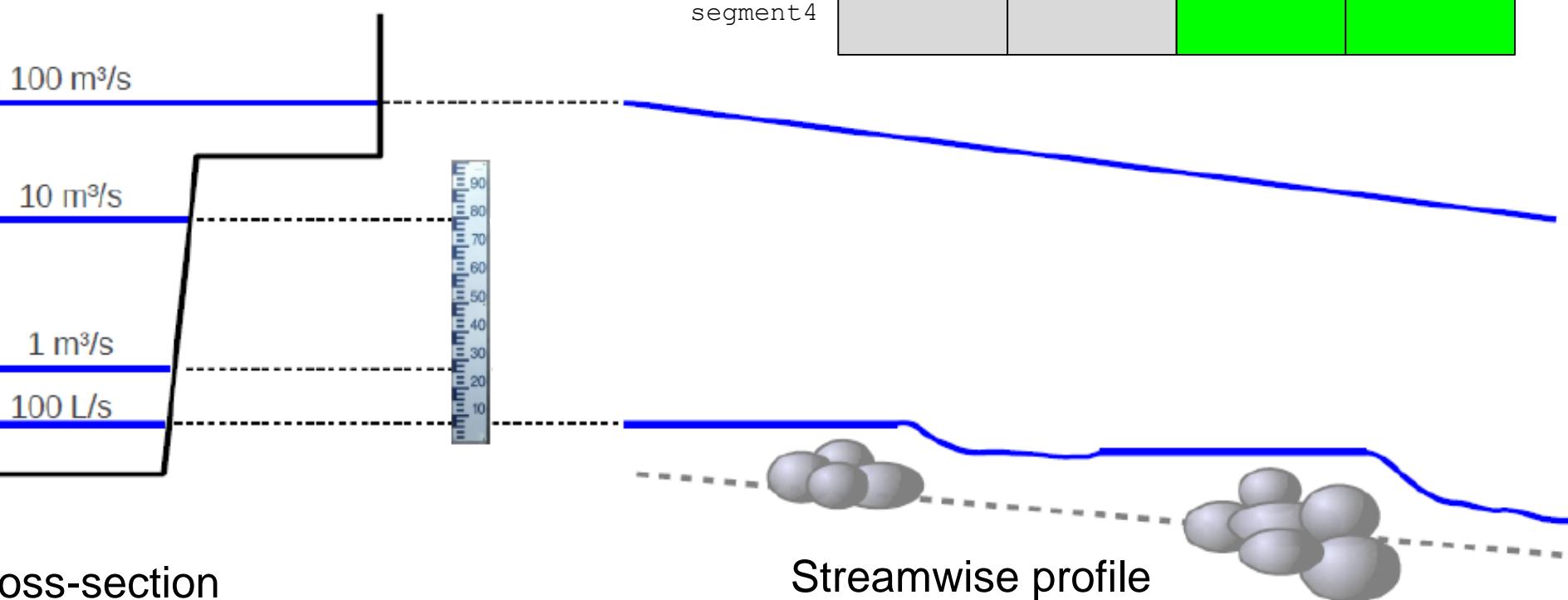
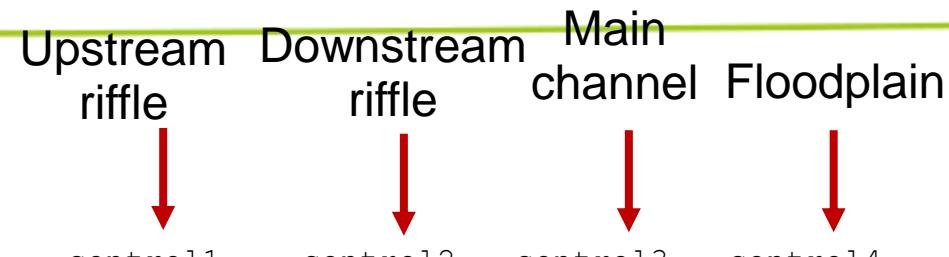
Depending on the water level, differing controls may appear or disappear. Several controls may add up.



Hydraulic analysis

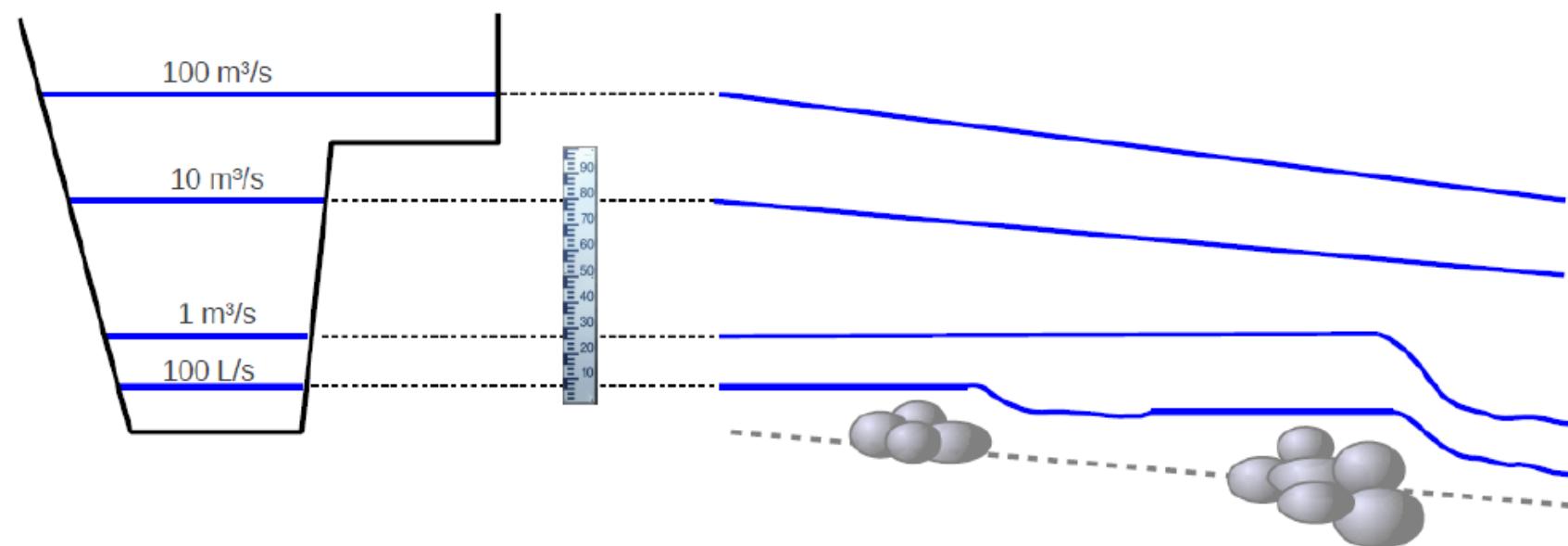
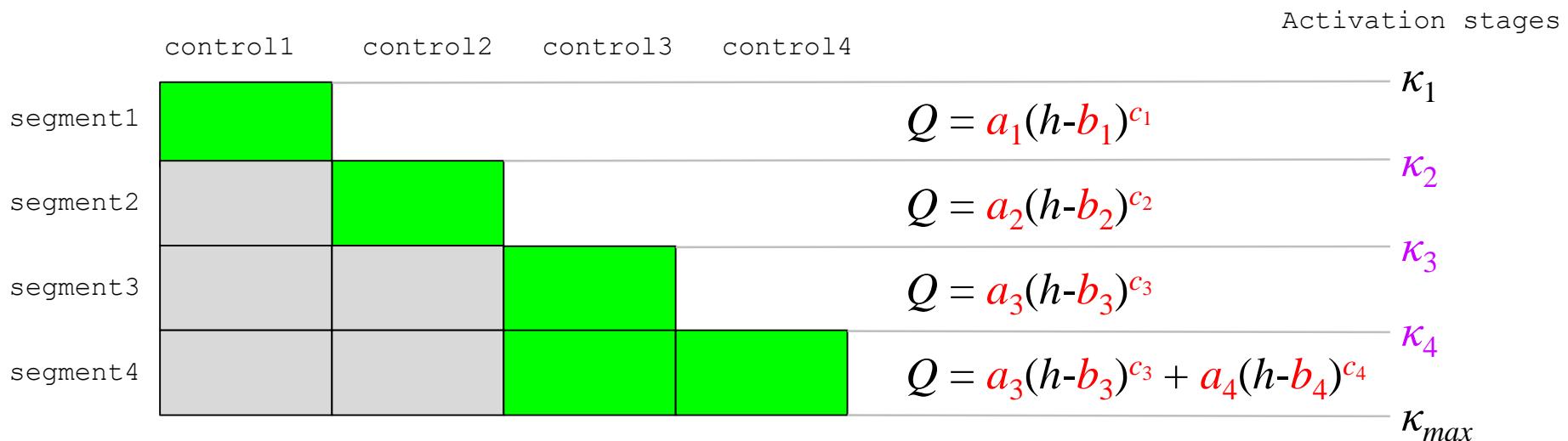
The main controls are identified or assumed.

The succession of controls over rating curve segments is represented by a matrix.



Rating curve equation

Each control can be modelled as: $Q = a(h-b)^c$

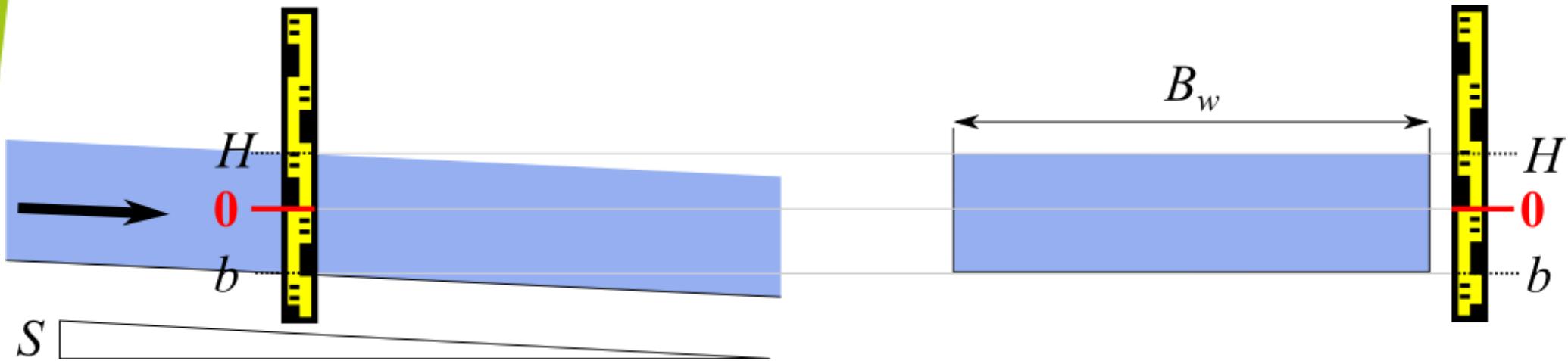


The BaRatin method for rating curves

- ✓ Introduction
- ✓ Hydraulic principles behind the rating curve
- ✓ Rating curve estimation
- ✓ Going further

Standard controls in BaRatinAGE

- Wide, rectangular channel (fairly uniform flow)

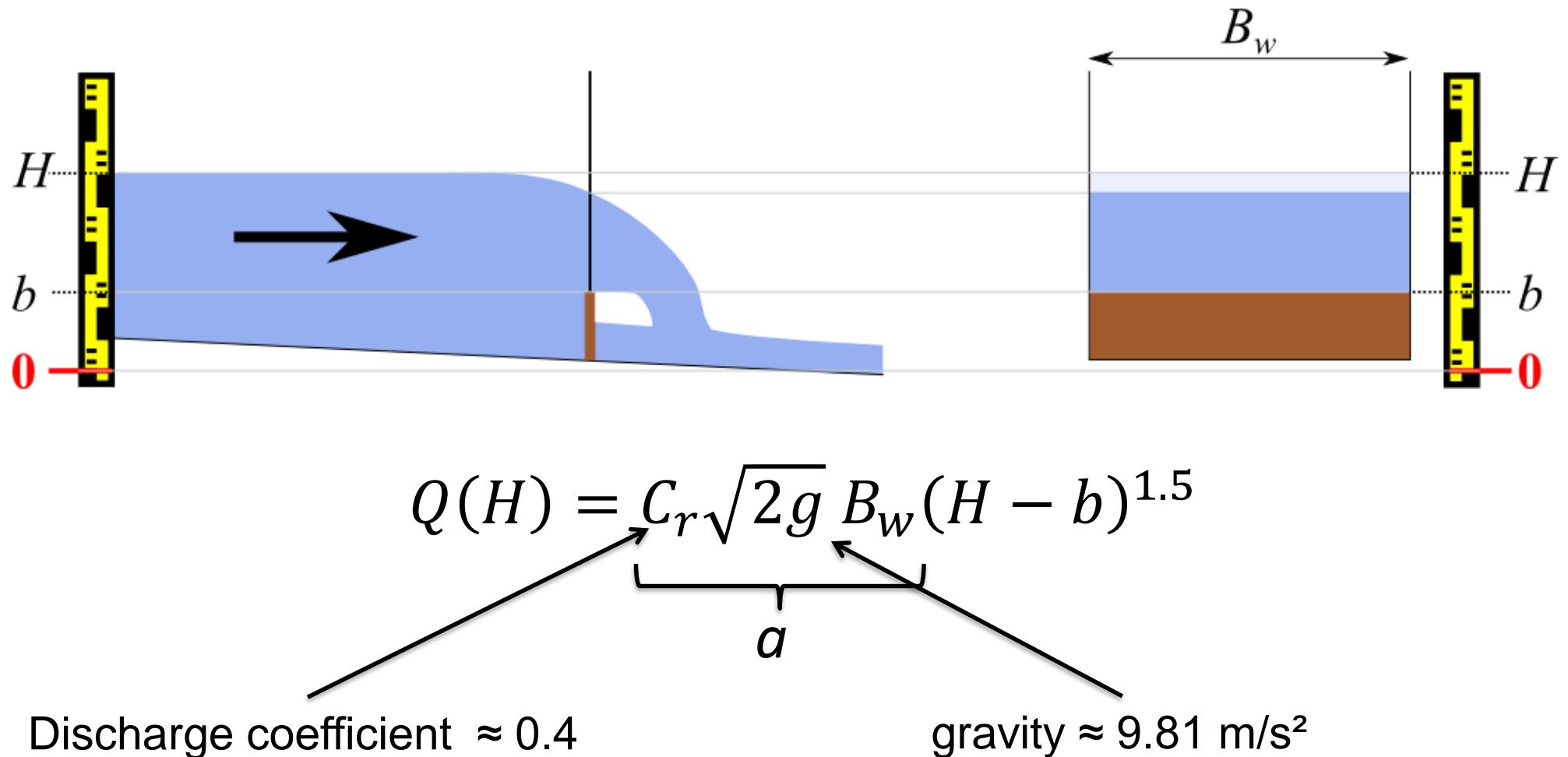


$$Q(H) = K_S \sqrt{S} B_w (H - b)^{5/3}$$

Strickler coefficient

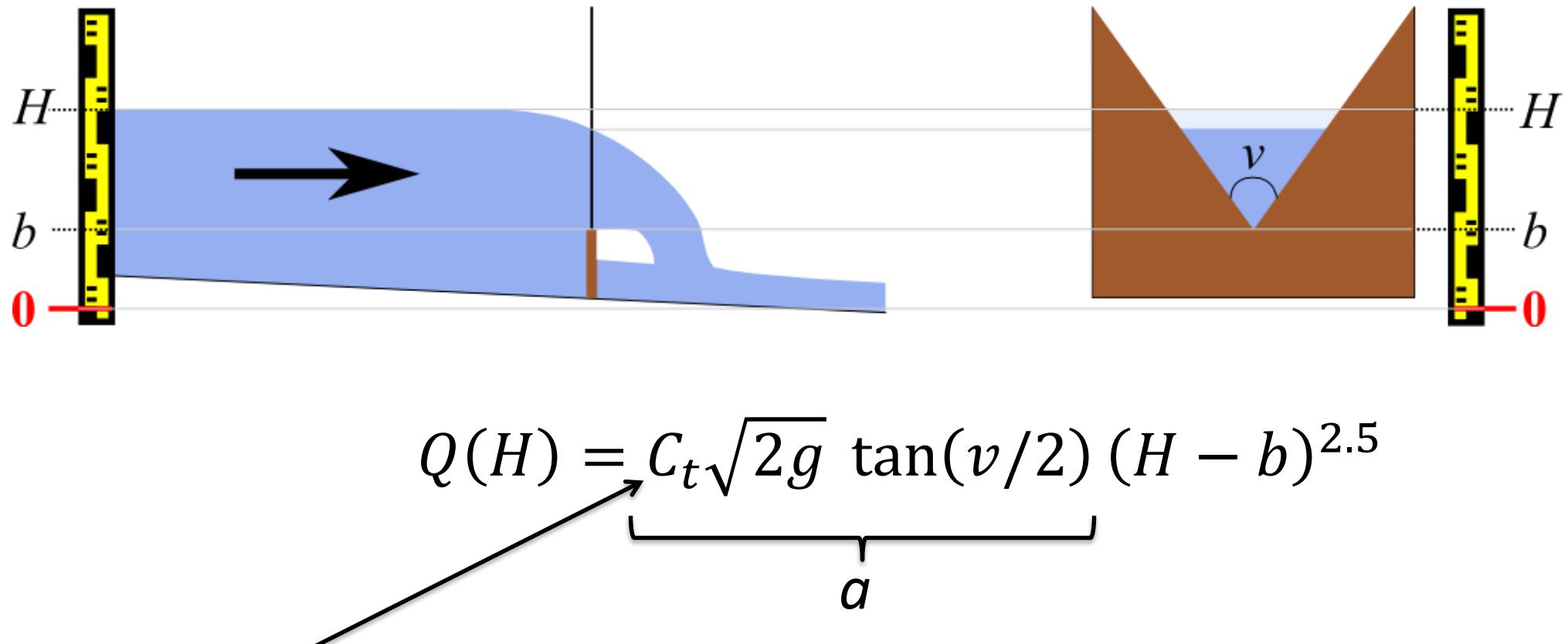
Standard controls in BaRatinAGE

- Rectangular weir / natural riffle



Standard controls in BaRatinAGE

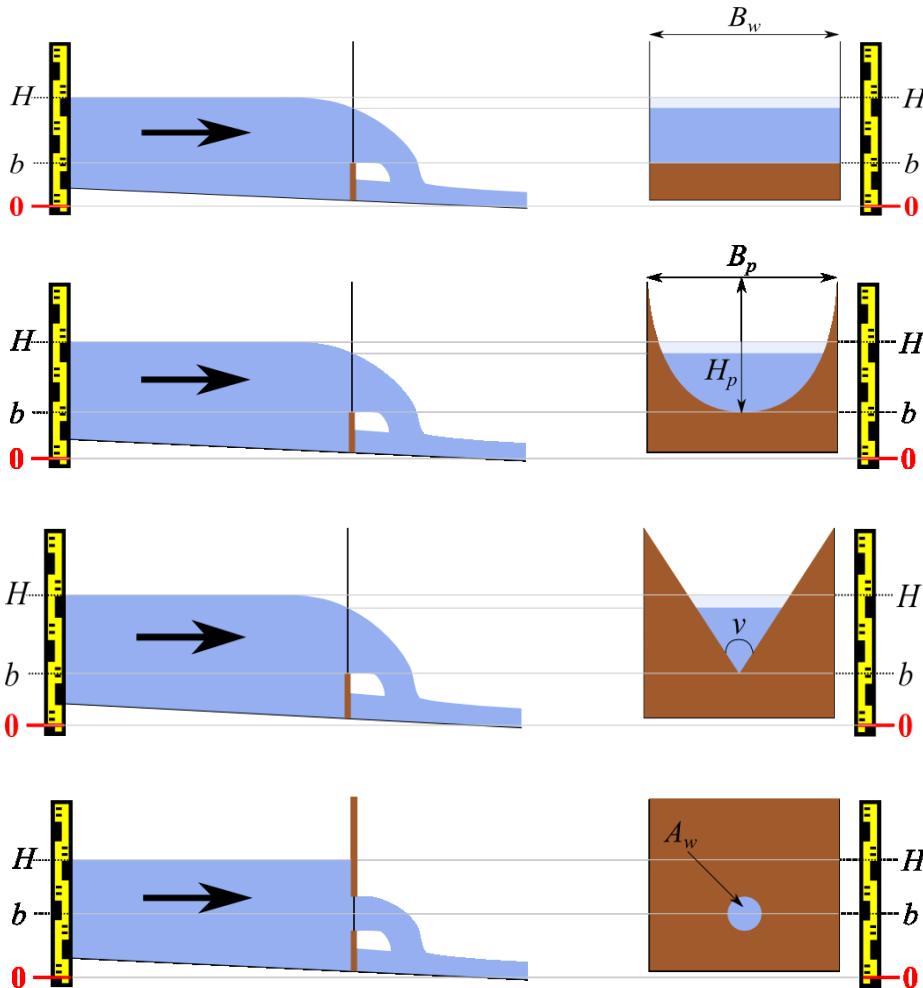
- Triangular weir / V-notch



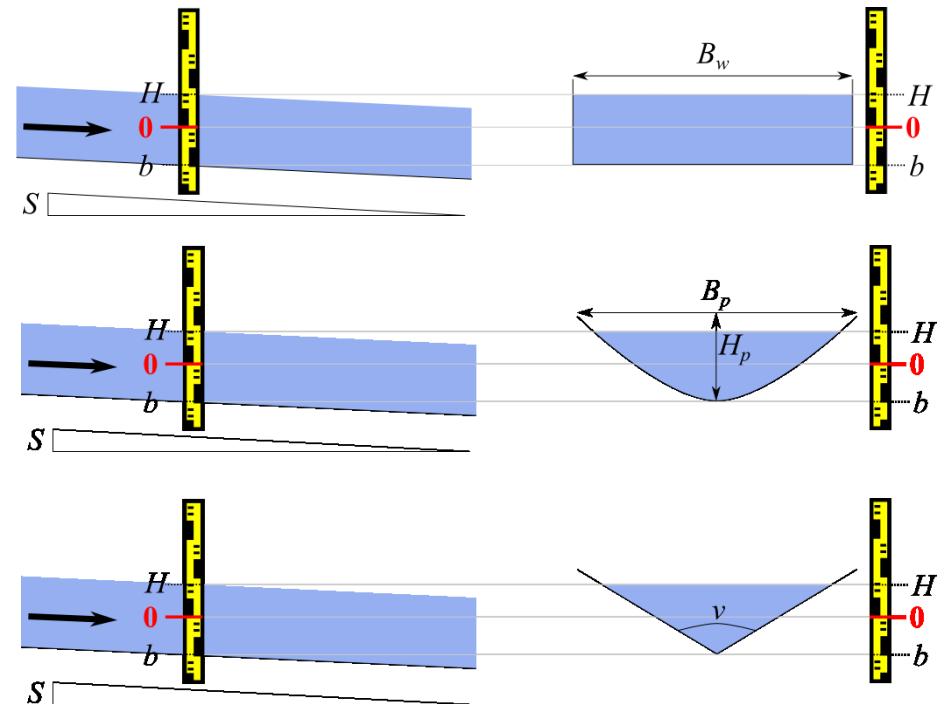
Discharge coefficient ≈ 0.31

Standard controls in BaRatinAGE

Section controls



Channel controls

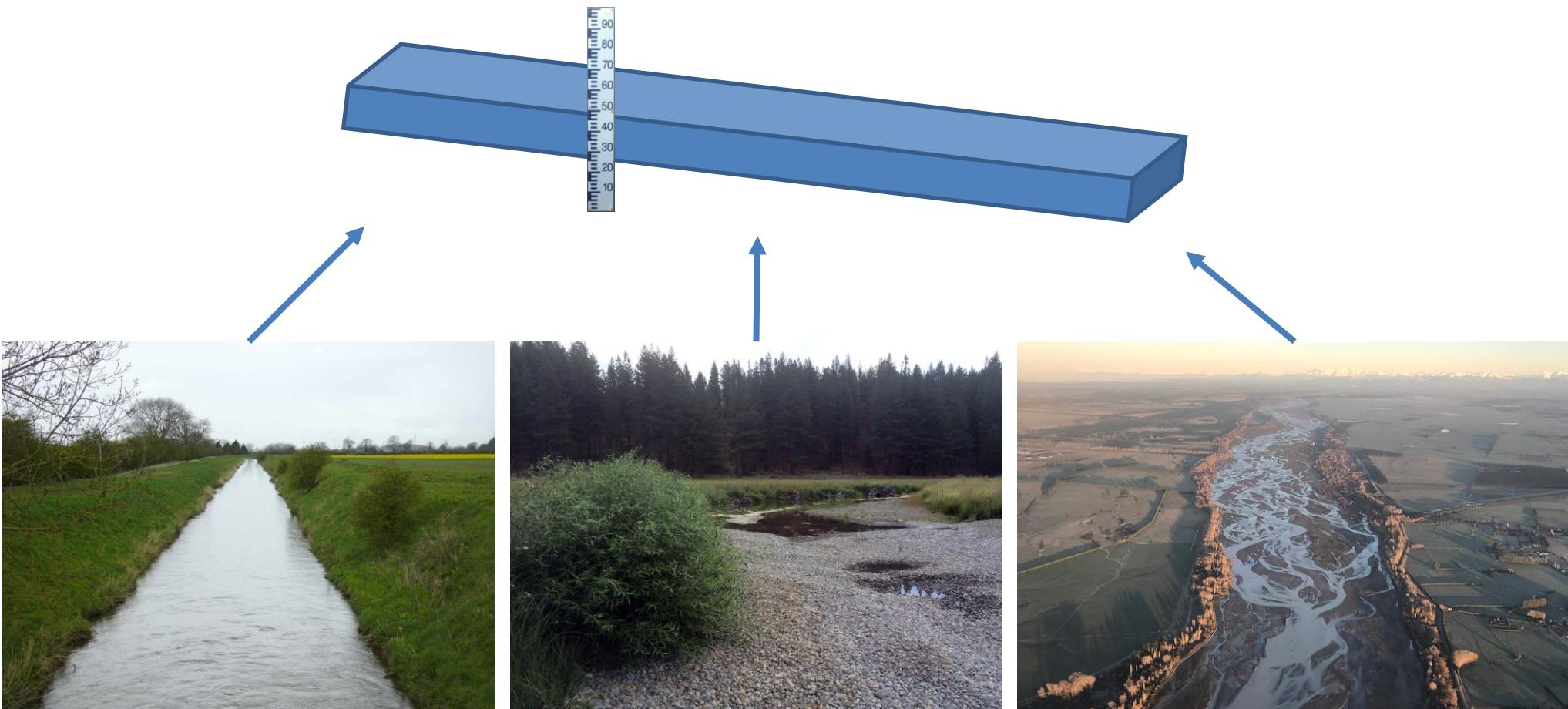


Unknown controls

$$Q = a(h - b)^c$$

Approximation of natural controls

- *Attention! The geometry of a channel control is an average of the section that extends downstream and upstream of the gauge*



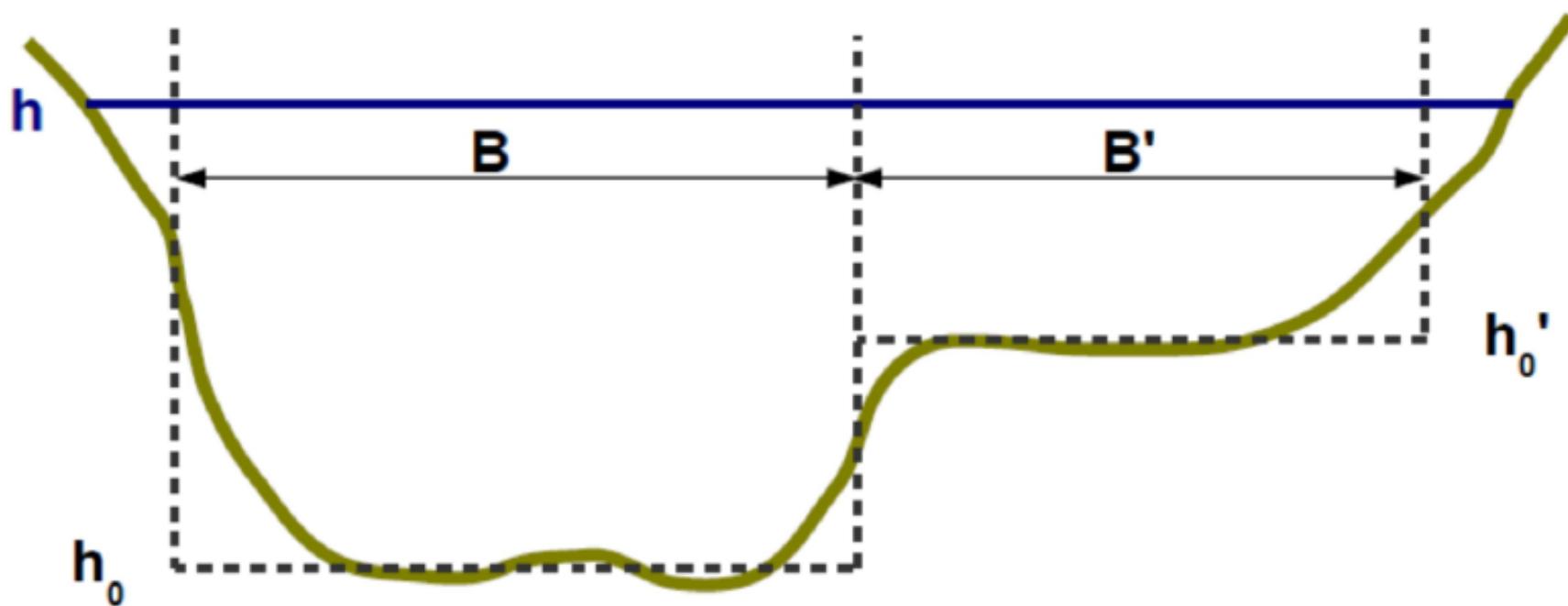
River Derwent, UK

Upper Truckee River, USA

Waimakariri, New Zealand

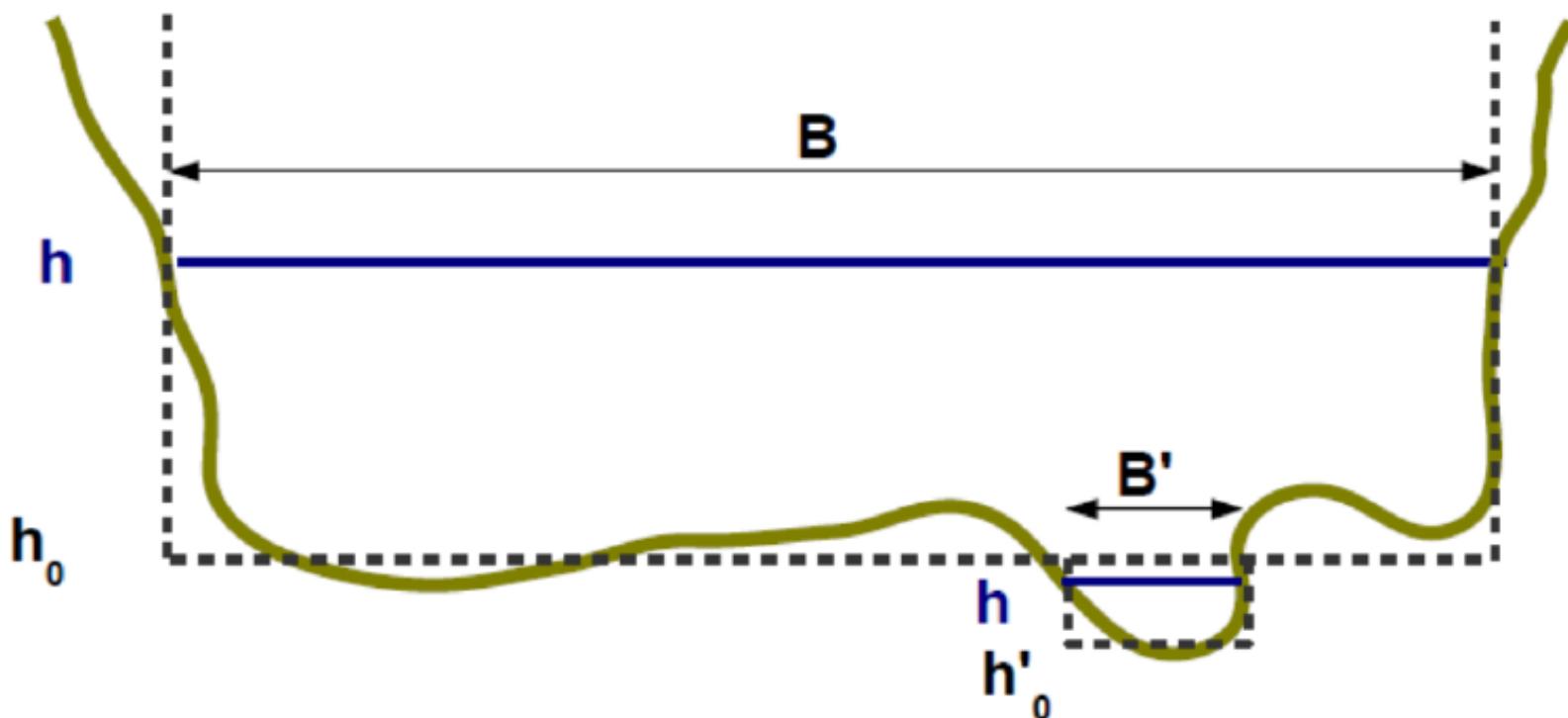
Approximation of natural controls

- Approximation of a compound channel using 2 rectangular channels



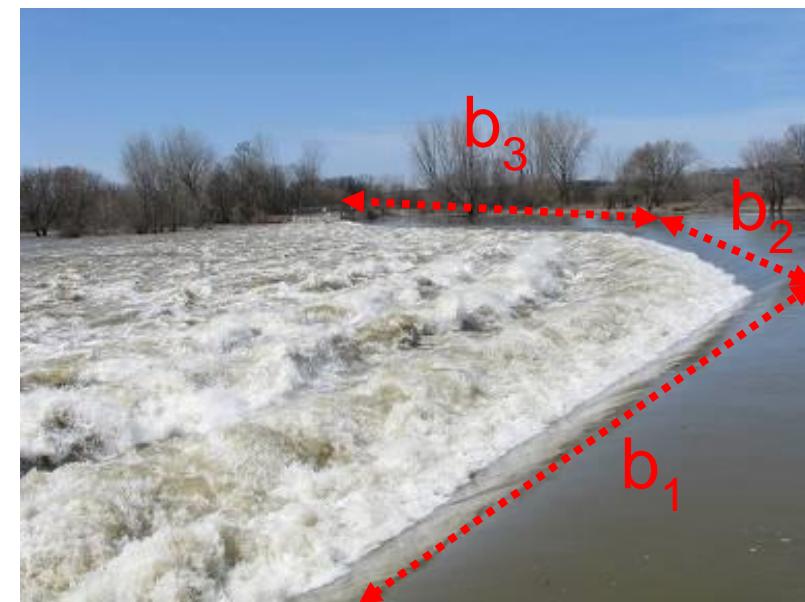
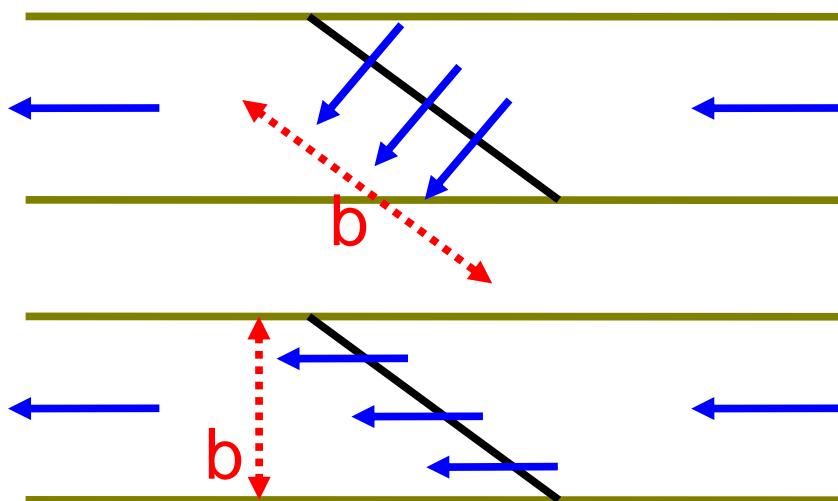
Approximation of natural controls

- Approximation of a complex critical section (natural riffle) by two nested rectangular weirs



Approximation of natural controls

- Attention! The overflow width of the weir is counted perpendicular to the direction of flow



Bayesian inference of hydraulic parameters

2

Bayesian inference

Example of a weir:

$$Q(h) = a(h - b)^c$$



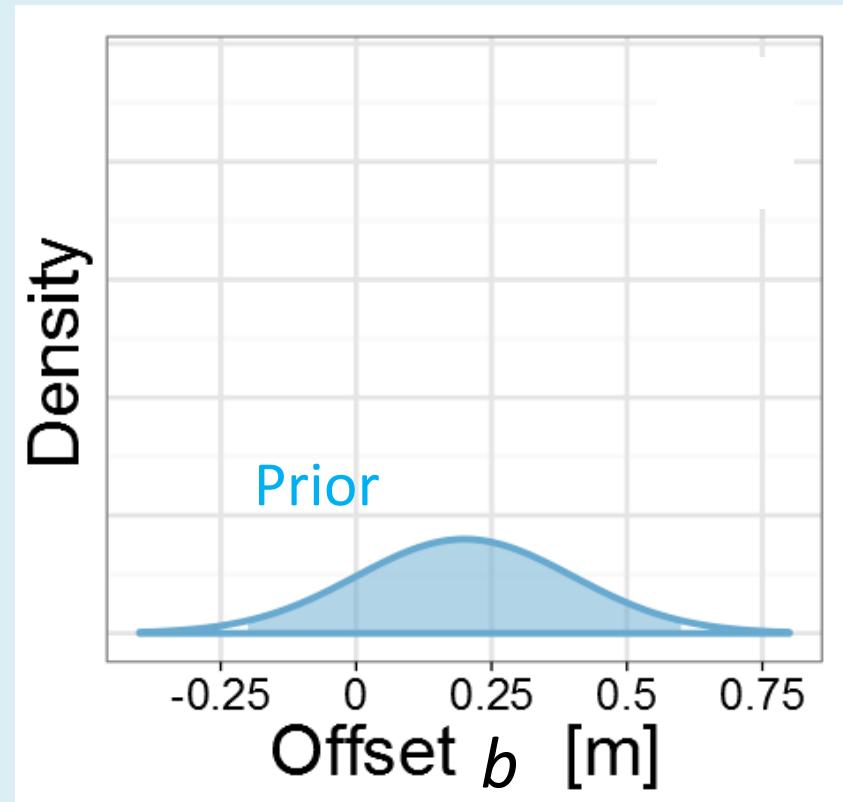
The Altier River at La Goulette,
France (EDF-DTG)

Bayesian inference of hydraulic parameters

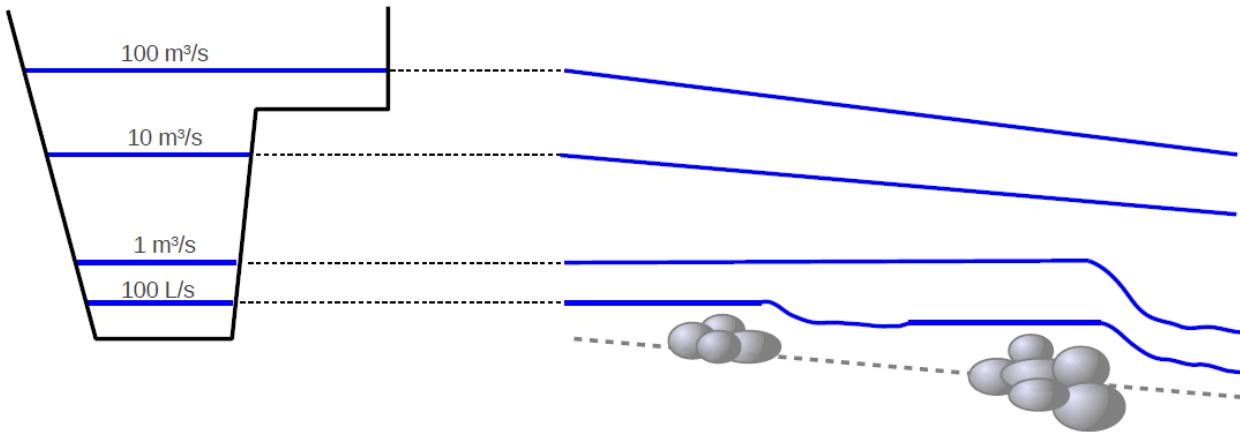
2

Bayesian inference

Prior knowledge: $b = 0.2 \text{ m} \pm 0.4 \text{ m}$



And now what?...



$$Q(h) = \begin{cases} 0 & \text{si } h < k_1 \\ a_1(h - b_1)^{c_1} & \text{si } k_1 \leq h < k_2 \\ a_2(h - b_2)^{c_2} & \text{si } k_2 \leq h < k_3 \\ a_3(h - b_3)^{c_3} & \text{si } k_3 \leq h < k_4 \\ a_3(h - b_3)^{c_3} + a_4(h - b_4)^{c_4} & \text{si } k_4 \leq h \end{cases}$$

To be estimated:
3 parameters
per control

Once we have the equation of the rating curve...

... we need to estimate parameters k_i, a_i, c_i (b_i are deduced by continuity)

The magics of Bayesian inference



The "posterior" distribution of the parameters of the rating curve can be computed using Bayes theorem:

Reverend
Thomas Bayes
(1702-1761)

$$p(\theta|y) = \frac{p(y|\theta)p(\theta)}{p(y)}$$

Posterior distribution

Likelihood

Prior distribution

Normalization constant

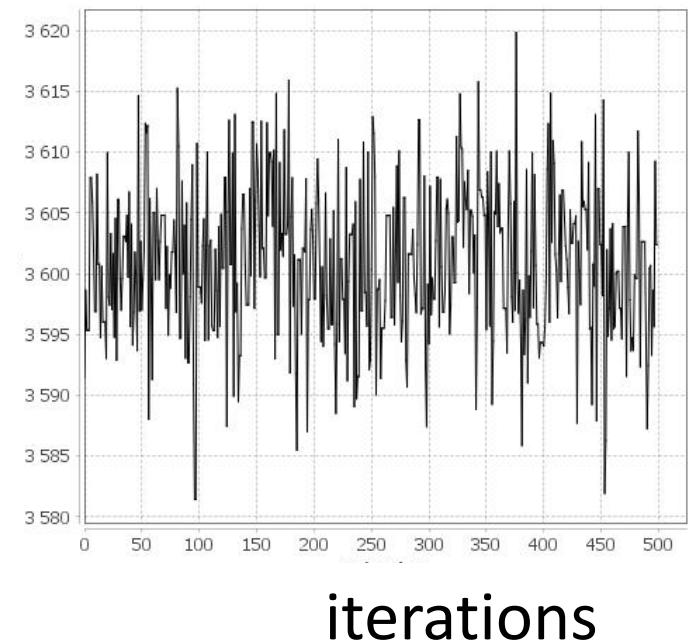
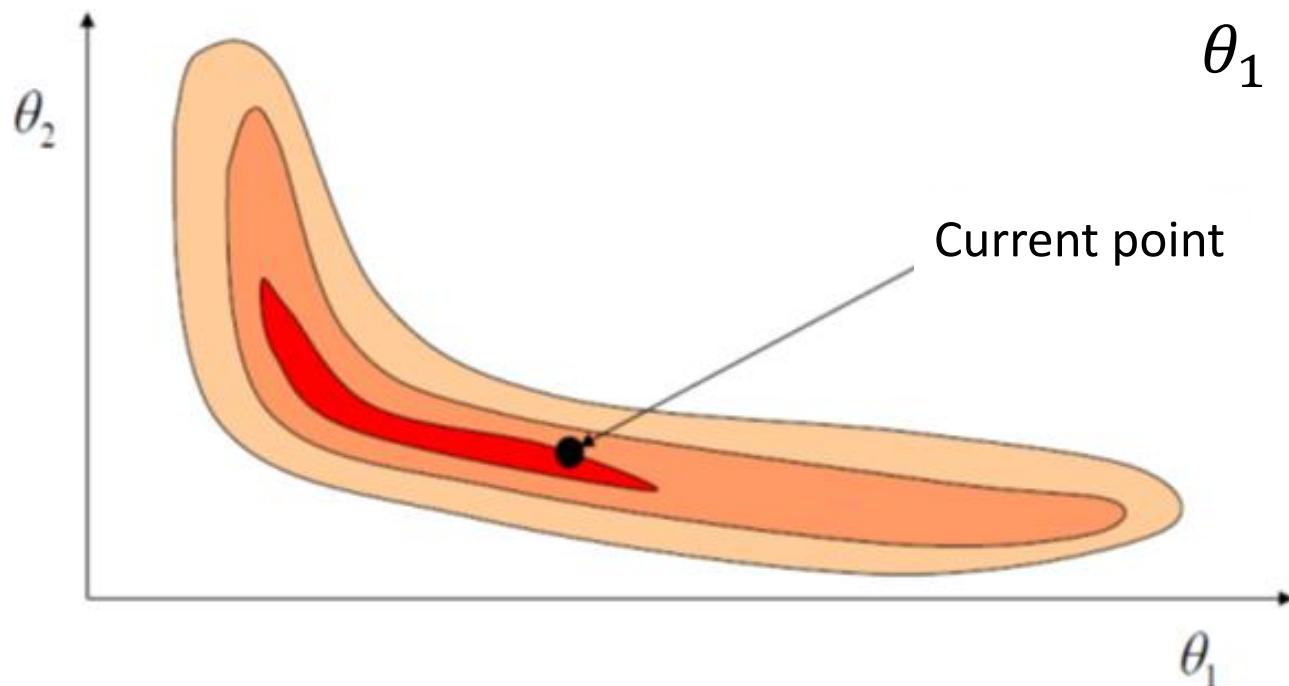
y : observations
(stage-discharge pairs: the gaugings)

θ : parameters of the rating curve

The magics of Bayesian inference



The posterior distribution is sampled by the Markov Chains Monte Carlo method (MCMC, Metropolis algorithm).



Bayesian inference of hydraulic parameters

Example of a weir:

$$Q(h) = a(h - b)^c$$

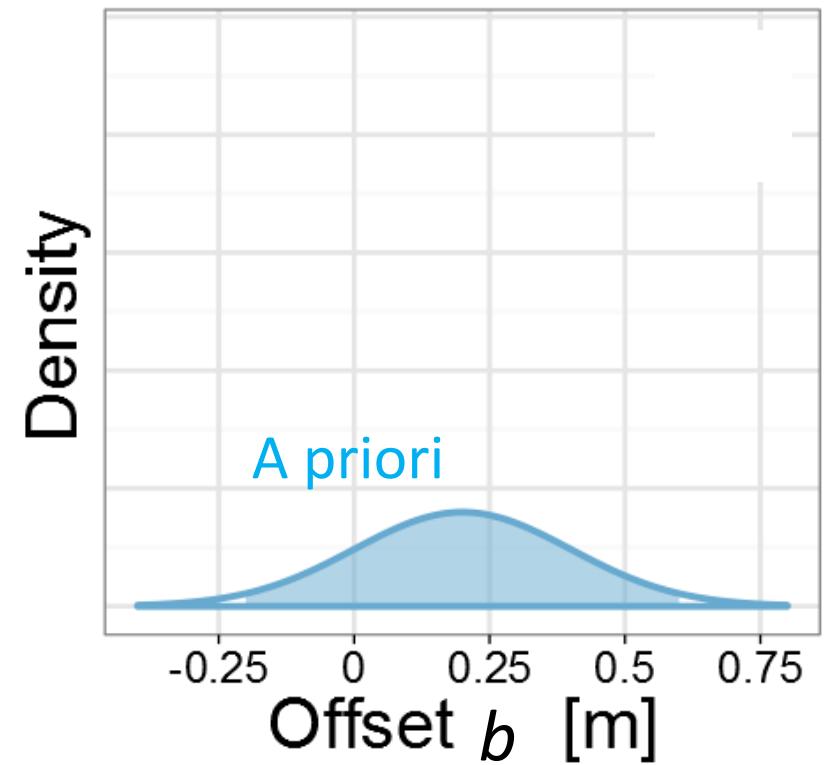


Altier River at Goulette, France (EDF-DTG)

Bayesian inference of hydraulic parameters

Prior knowledge:

$$b = 0.2 \text{ m} \pm 0.4 \text{ m}$$



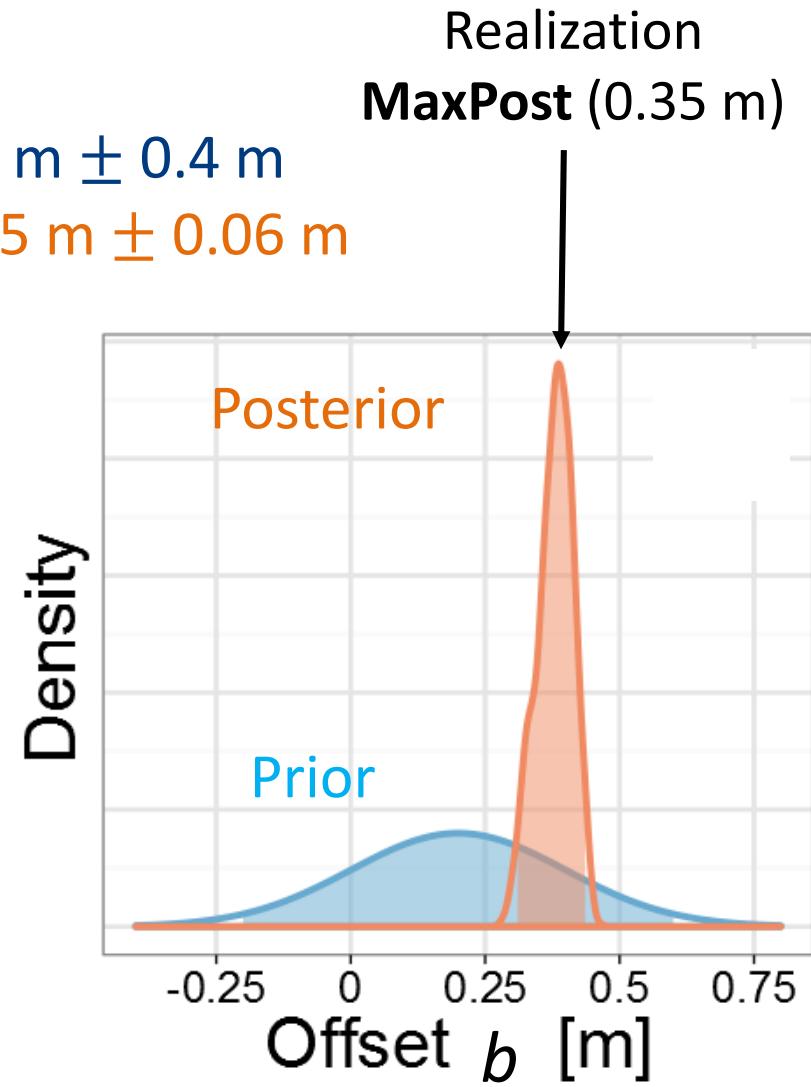
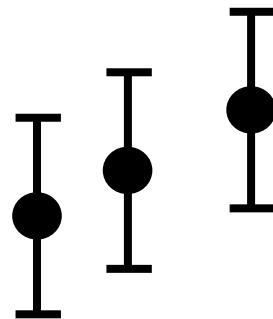
Bayesian inference of hydraulic parameters

Prior knowledge:
A posteriori :



$$b = 0.2 \text{ m} \pm 0.4 \text{ m}$$
$$b = 0.35 \text{ m} \pm 0.06 \text{ m}$$

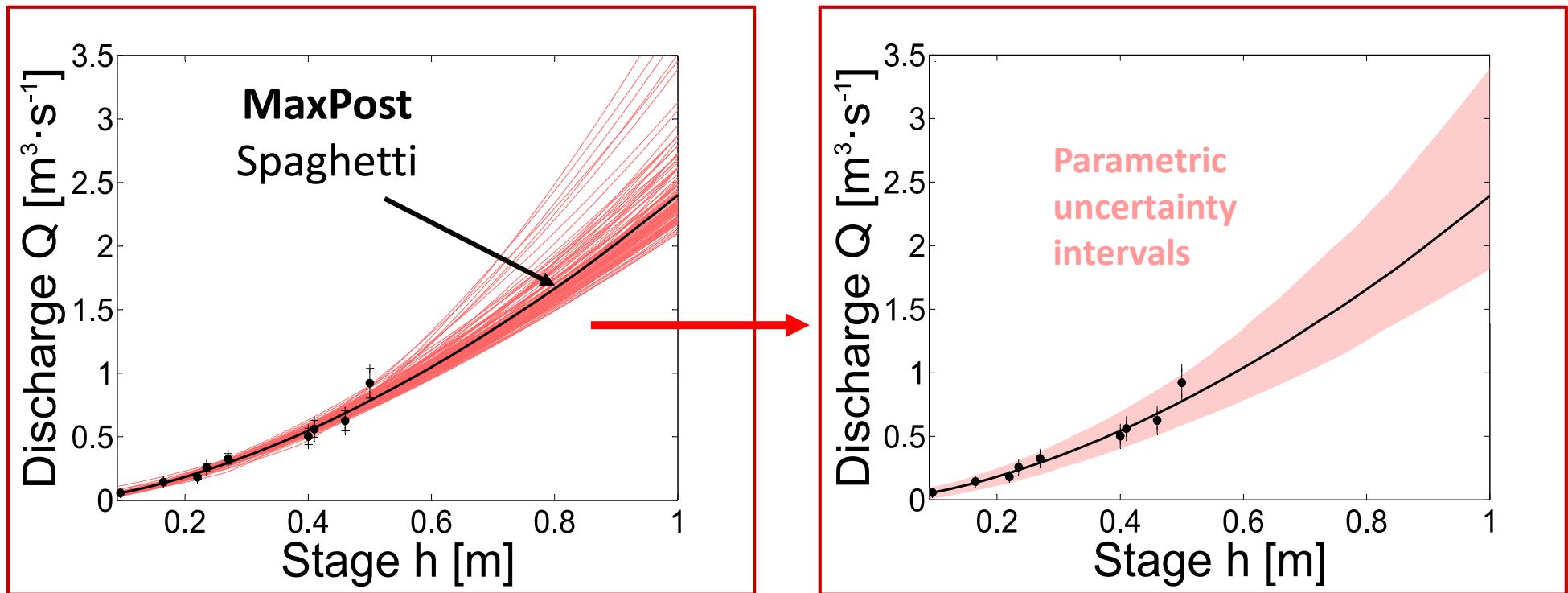
Observations (gaugings):



The "spaghetti" approach



Posterior distribution is sampled using MCMC techniques



+ Structural/Remnant uncertainty:

What is lacking to explain the scatter of the gaugings around the rating curve

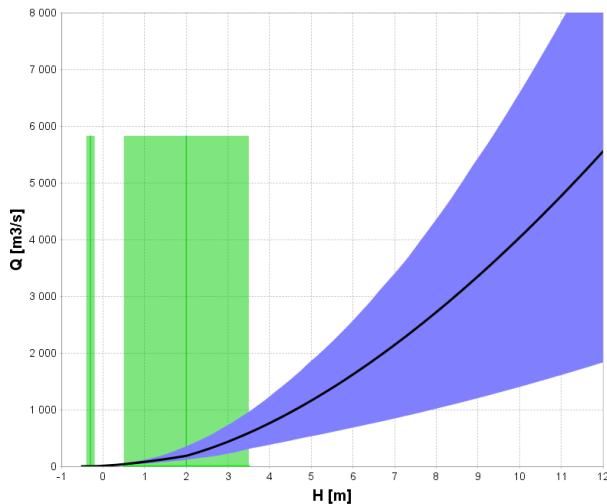
Bayesian estimation of the rating curve

The information contents of the measurements and of the hydraulic knowledge are combined.

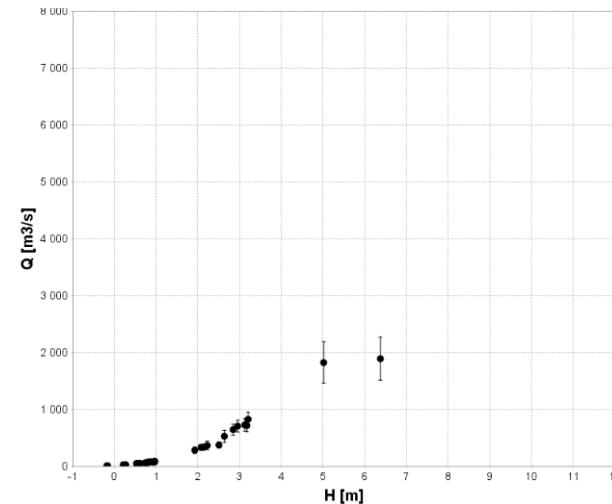
Possible interpretations:

1. The hydraulic estimation is refined using the measurements
2. The rating is adjusted to the measurements under hydraulic constraints

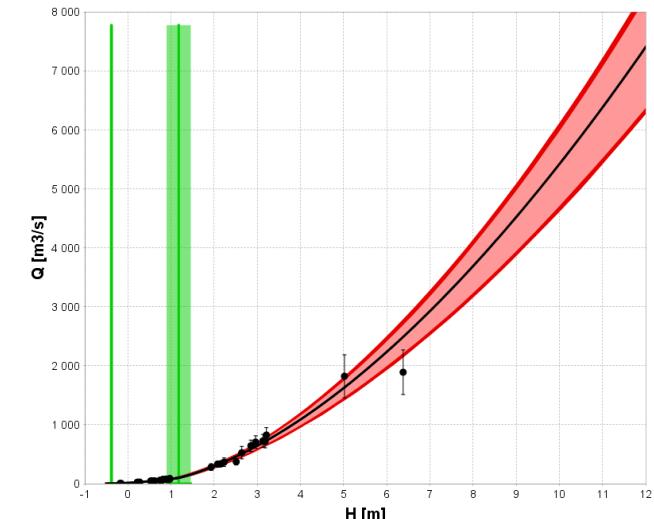
Prior rating



Measurements (uncertain)



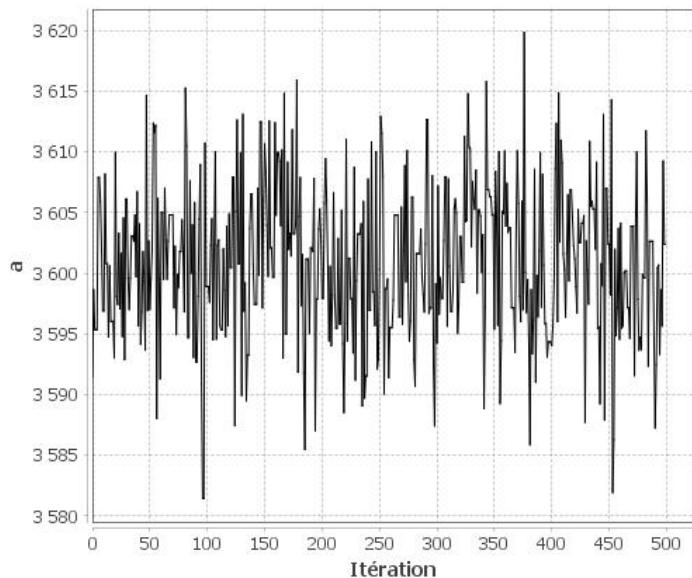
Posterior rating



Example: the Ardèche at Sauze

Practically: Check MCMC realizations (traces)

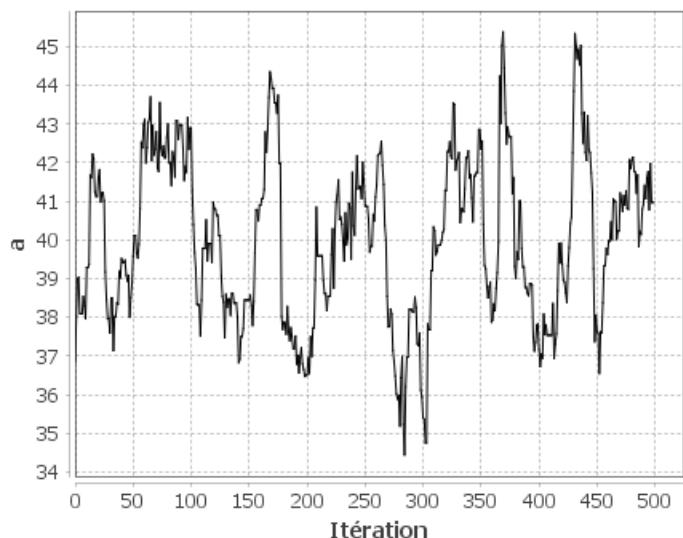
a - Contrôle 1



Good!

('stationary' random walk)

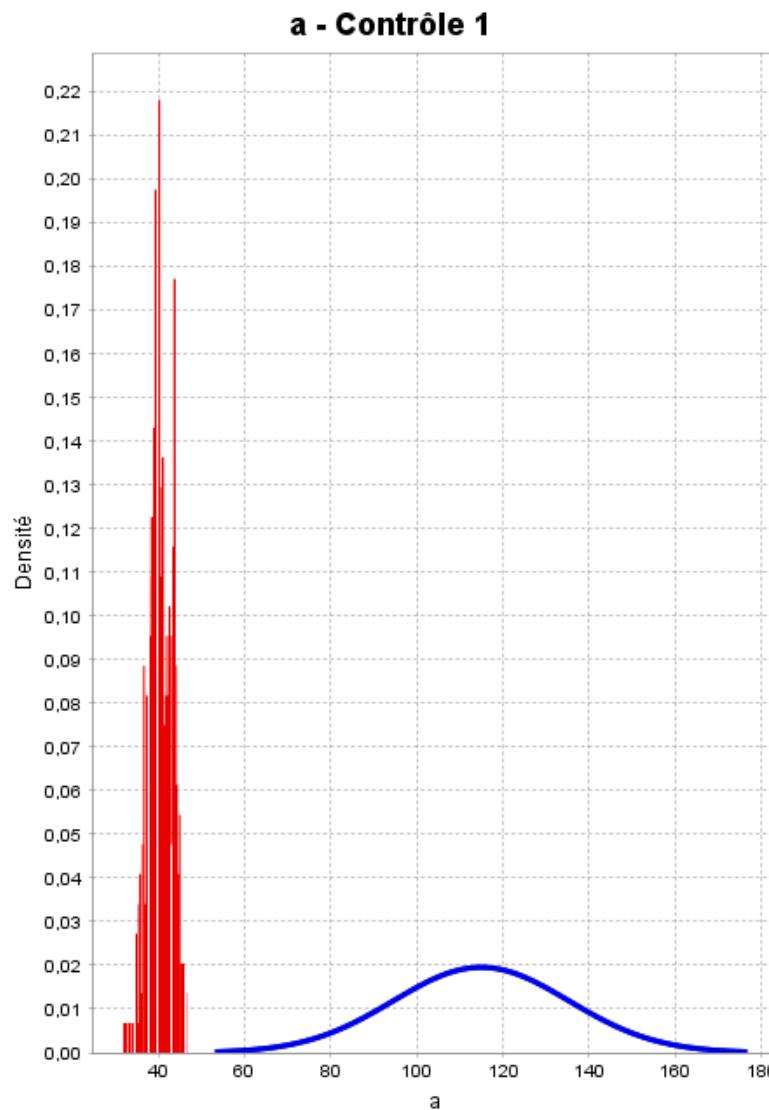
a - Contrôle 1



Not so good...
(clear trends)

Practically:

Check there is no conflit between prior and posterior



Not
good!

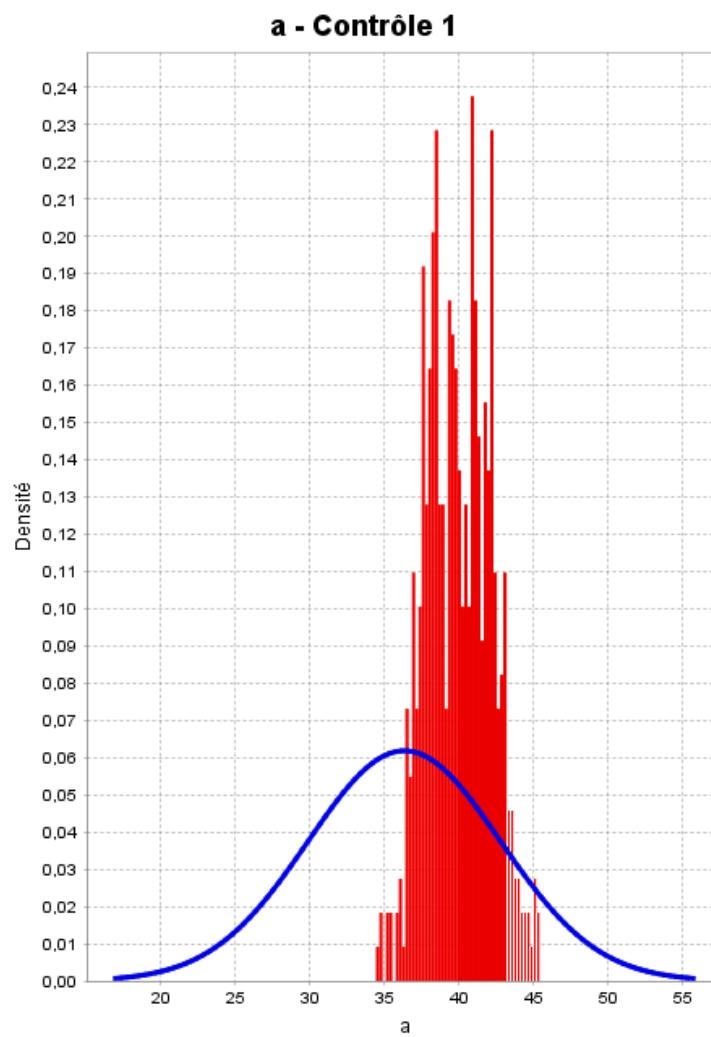
Practically:

Check there is no conflict between prior and posterior



Practically:

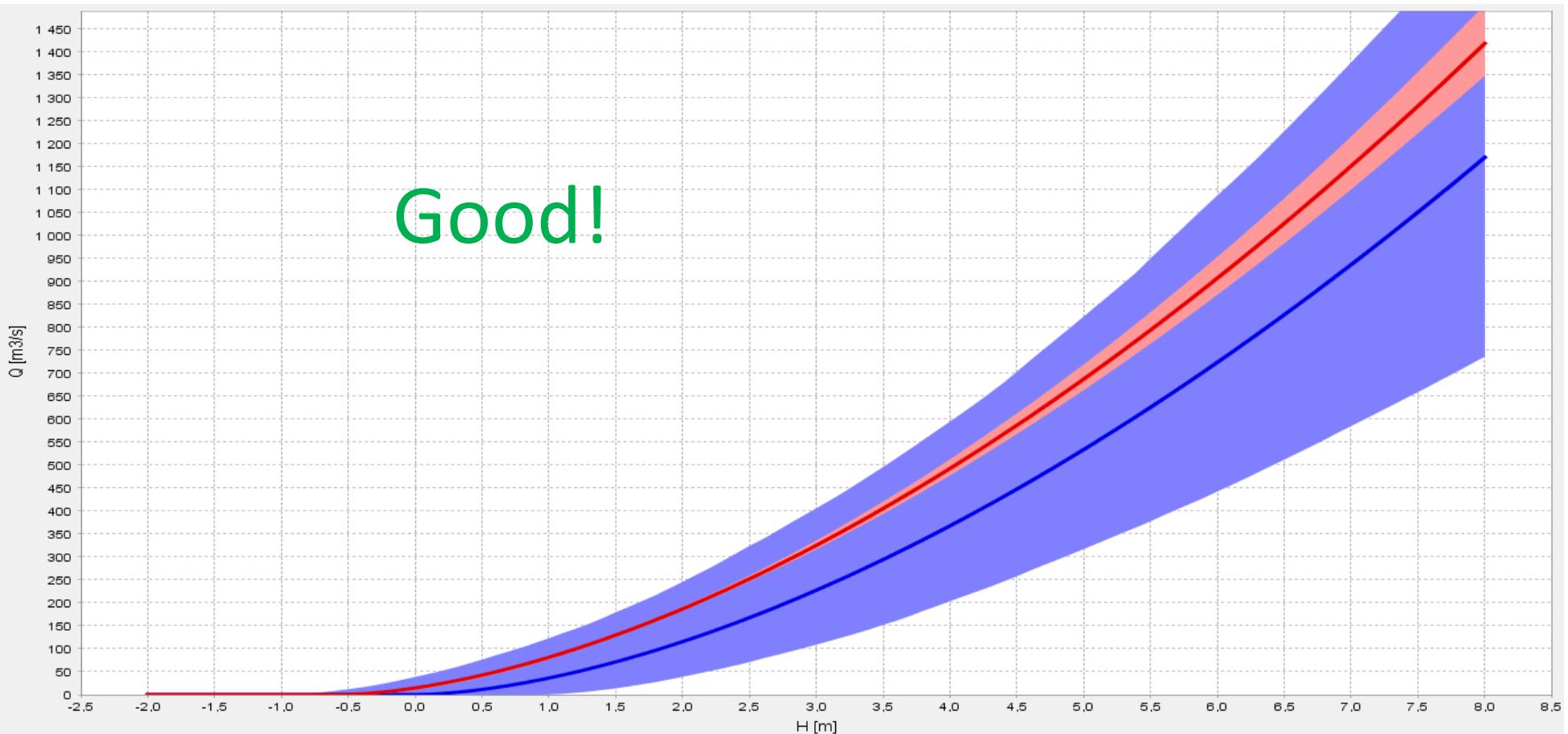
Check there is no conflit between prior and posterior



Good!

Practically:

Check there is no conflict between prior and posterior



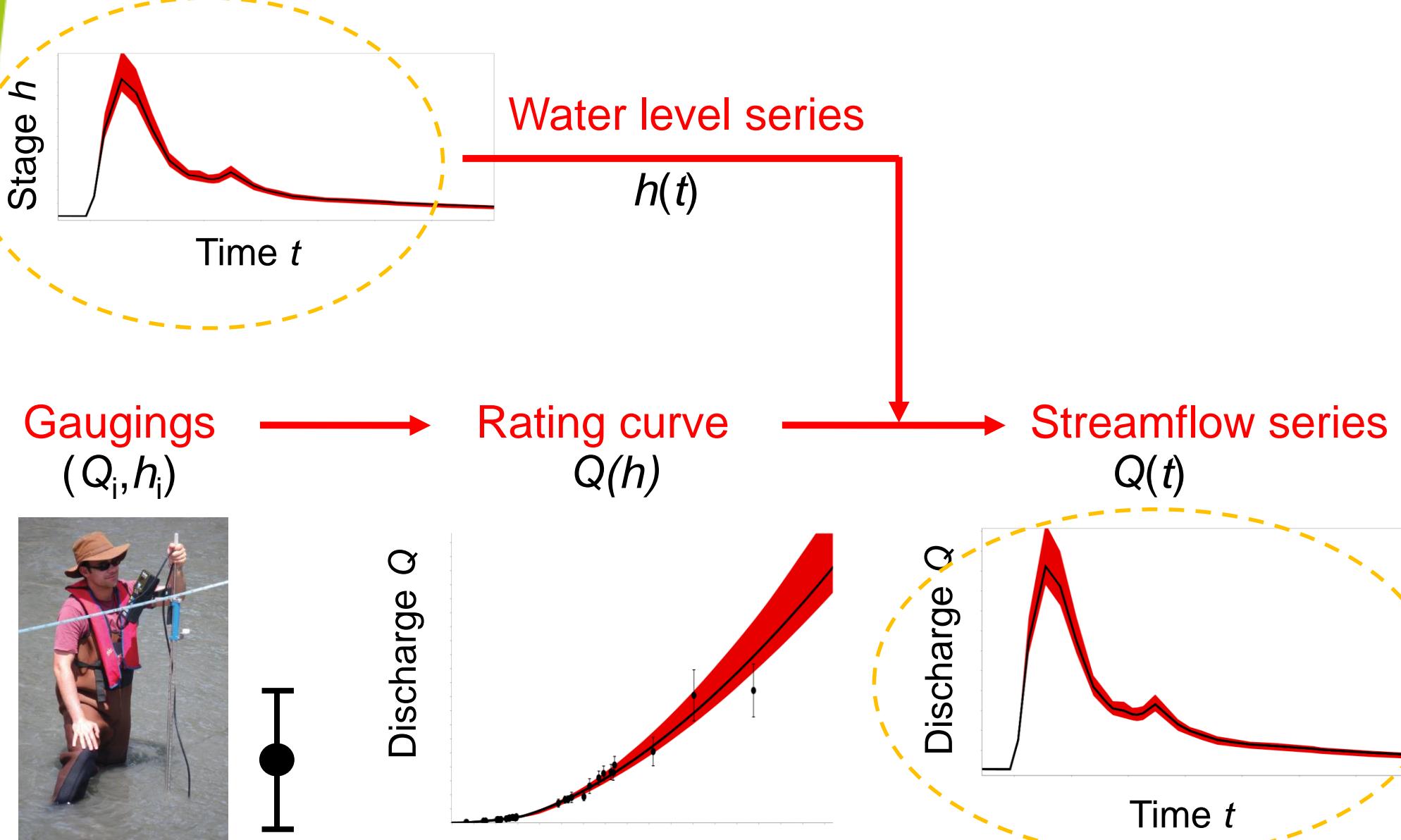
Practically:

Check there is no conflict between prior and posterior

If there appears to be a conflict:

- ✓ Check that the calculations went well (convergence of MCMC iterations)
- ✓ Check the values of the priors (do not set them using the results or the gaugings used in the estimation!)
- ✓ Review your assumptions on hydraulic controls, test other hydraulic configurations
- ✓ Check the gaugings and their uncertainties

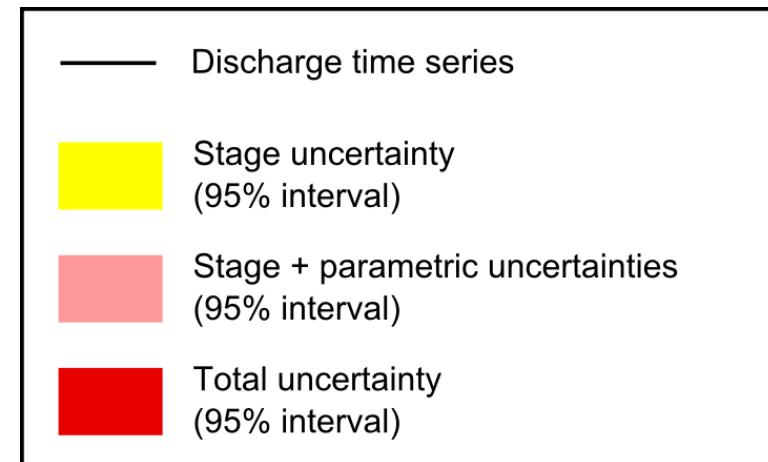
Establishing probabilistic streamflow data



Uncertainty budgets

Uncertainty budgets help to rank sources of error and improve the measurement process

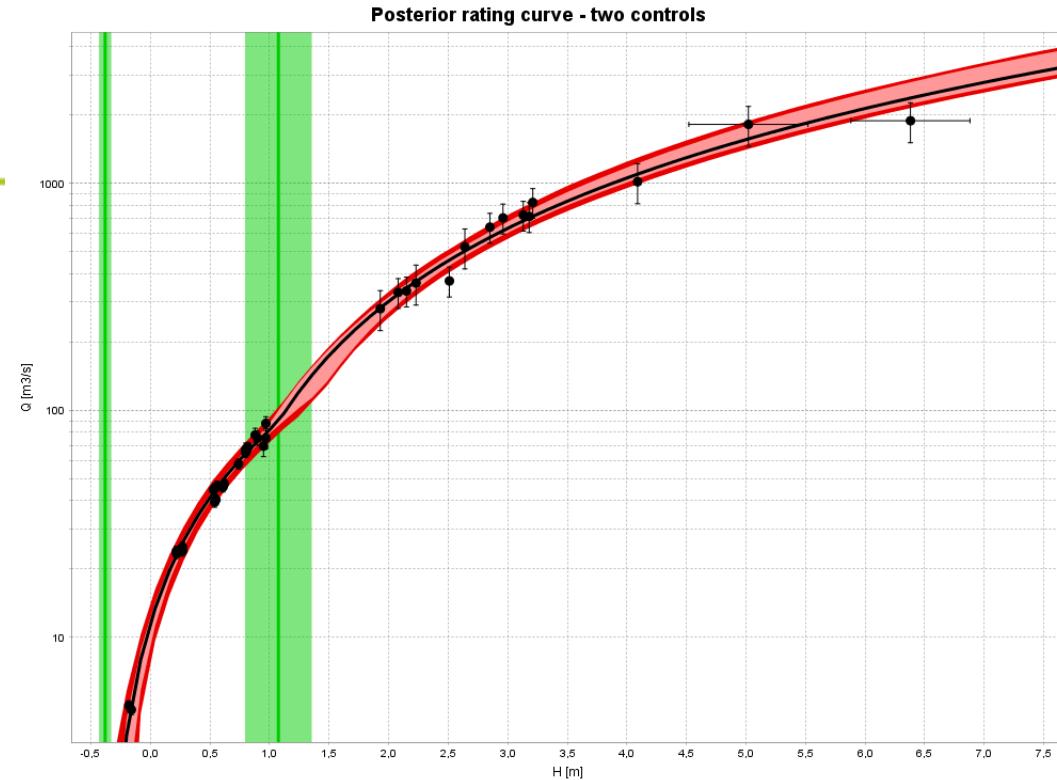
- ✓ Reflects the measurement uncertainties of the stage records
- ✓ Reflects the (limited) information contents of the priors and observations (the gaugings)
- ✓ Combines parametric and structural uncertainties
- ✓ Structural uncertainty reflects the limitations of the rating curve model for describing the real hydraulic conditions of the site (complex controls, shifts, hysteresis, variable backwater...)



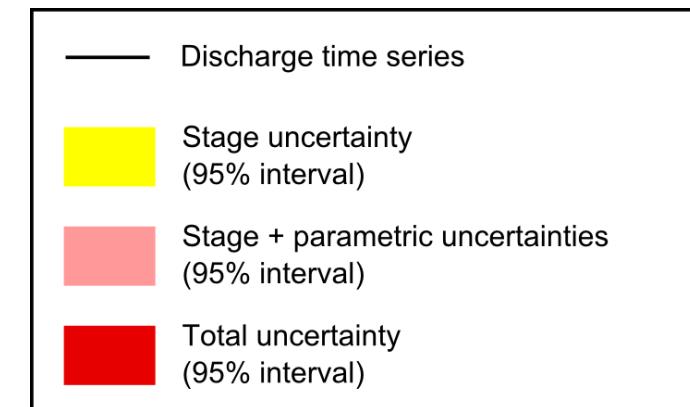
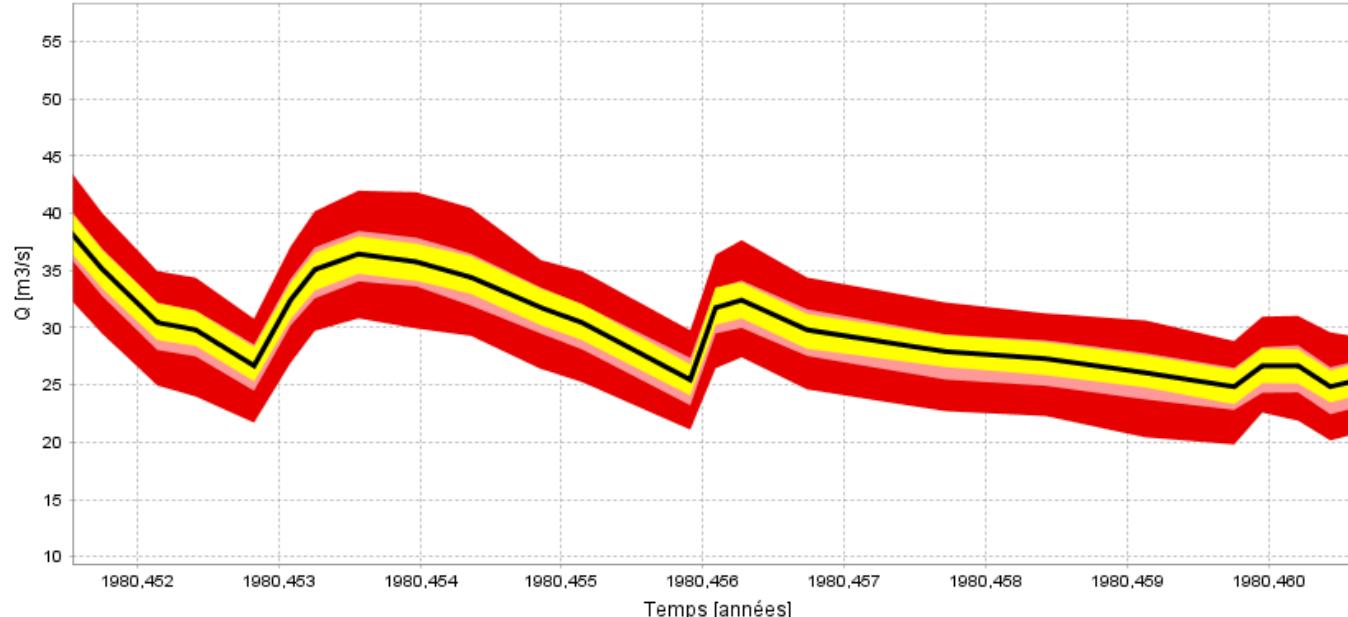
Uncertainty budgets

Uncertainty budgets help to rank sources of error and improve the measurement process

Low flows



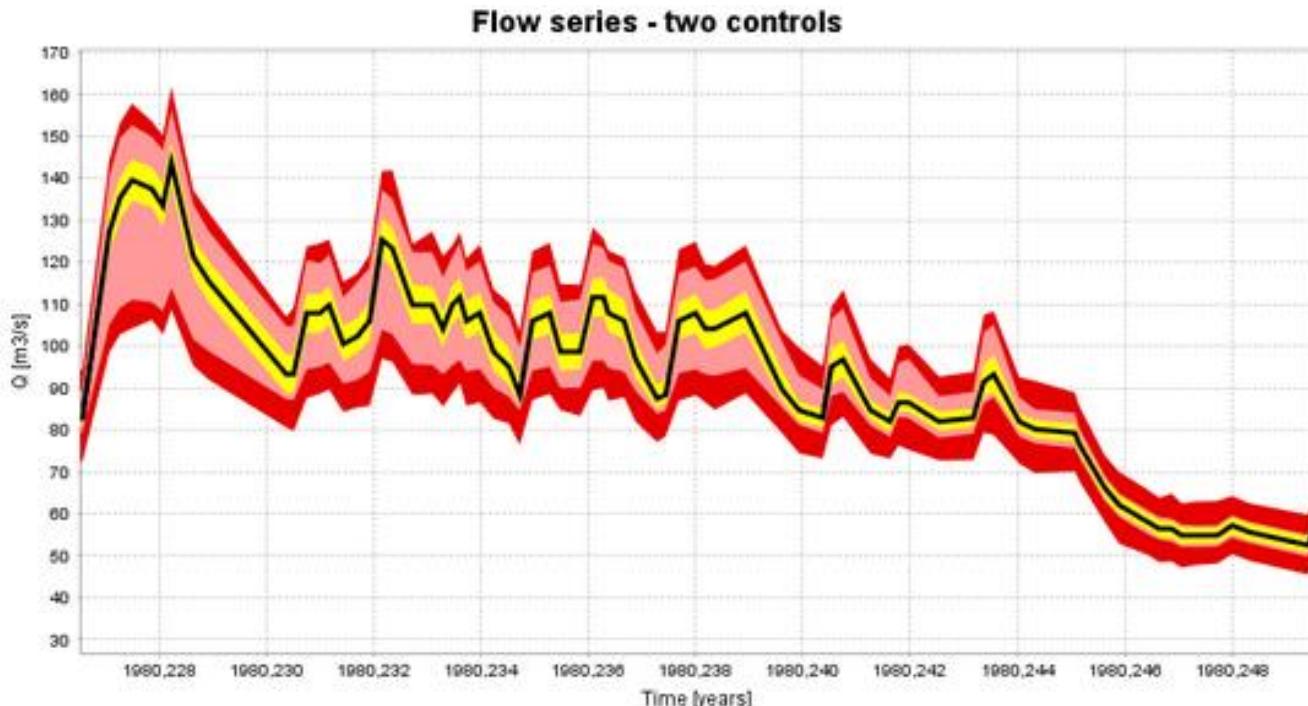
Hydrogramme - two controls



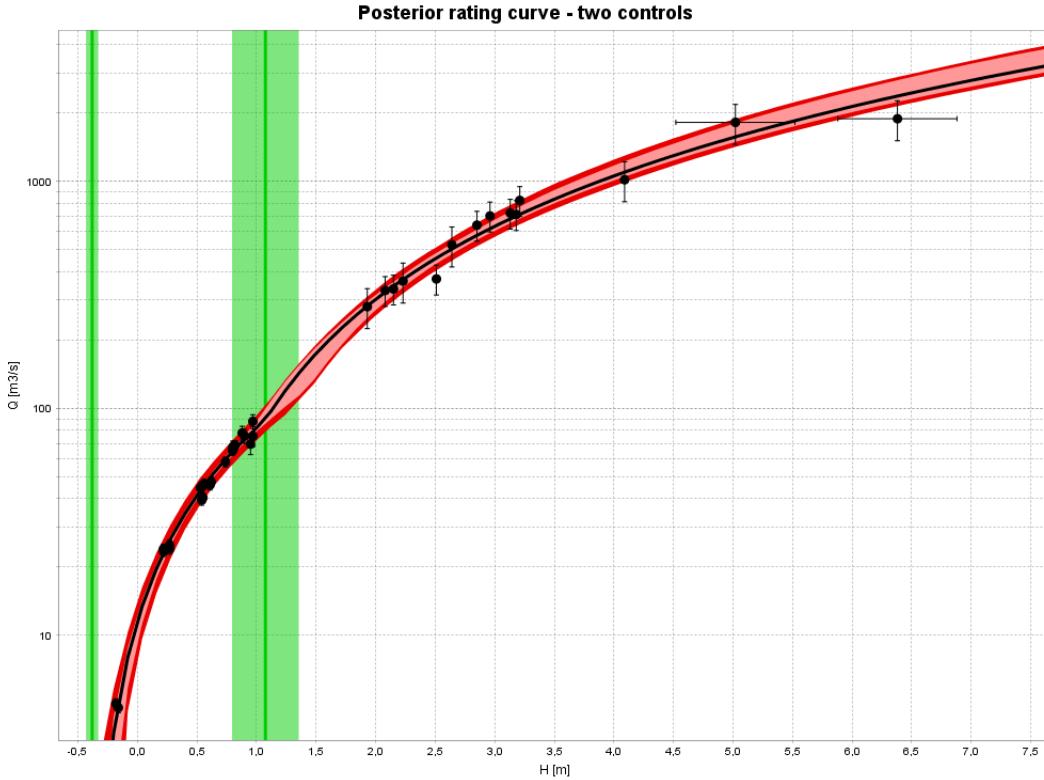
Uncertainty budgets

Uncertainty budgets help to rank sources of error and improve the measurement process

Intermediate flows



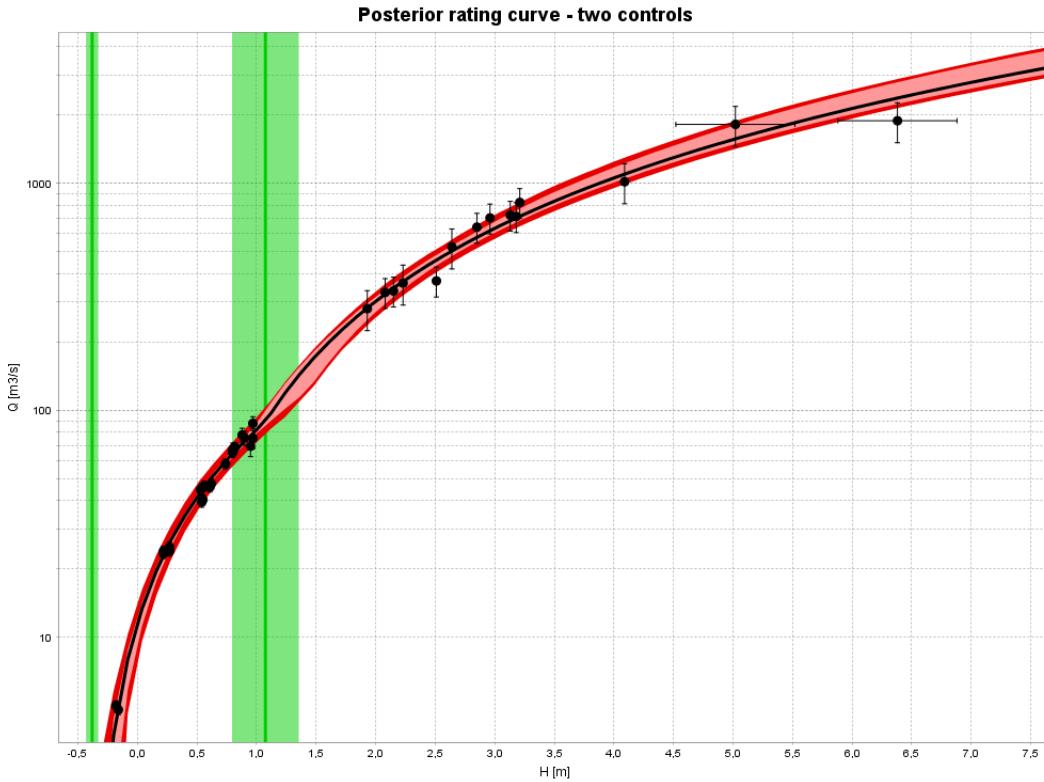
- Discharge time series
- Yellow: Stage uncertainty (95% interval)
- Pink: Stage + parametric uncertainties (95% interval)
- Red: Total uncertainty (95% interval)



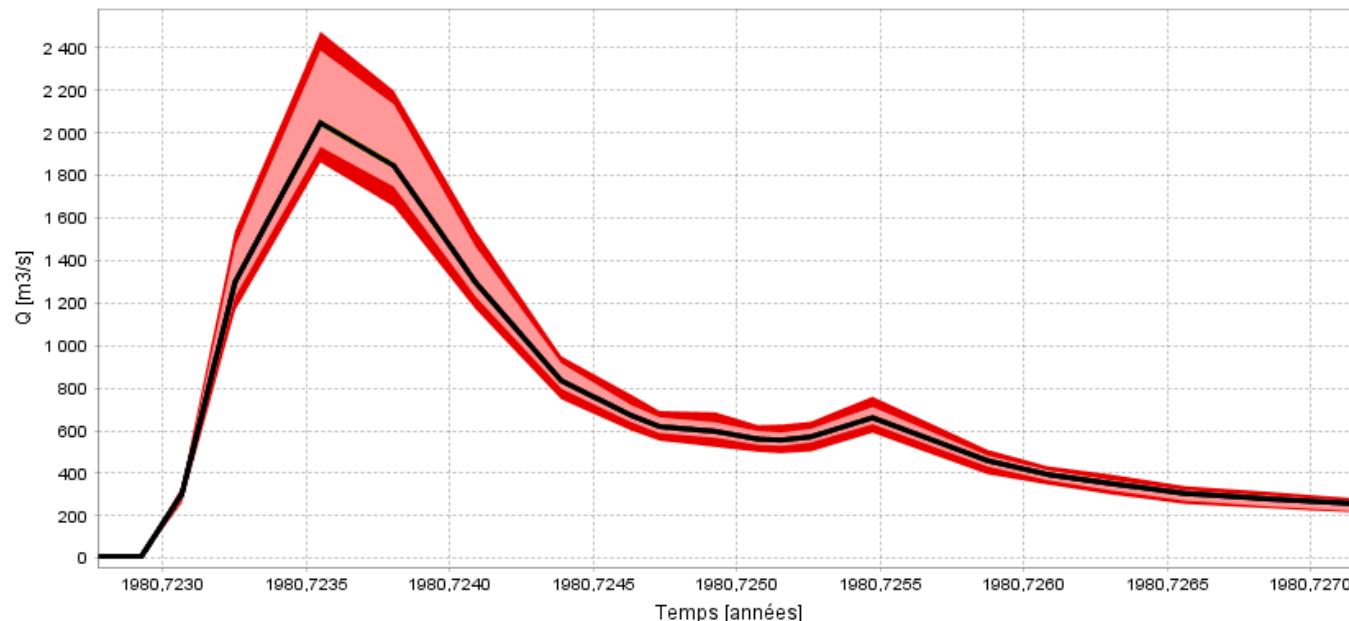
Uncertainty budgets

Uncertainty budgets help to rank sources of error and improve the measurement process

High flows



Hydrogramme - two controls



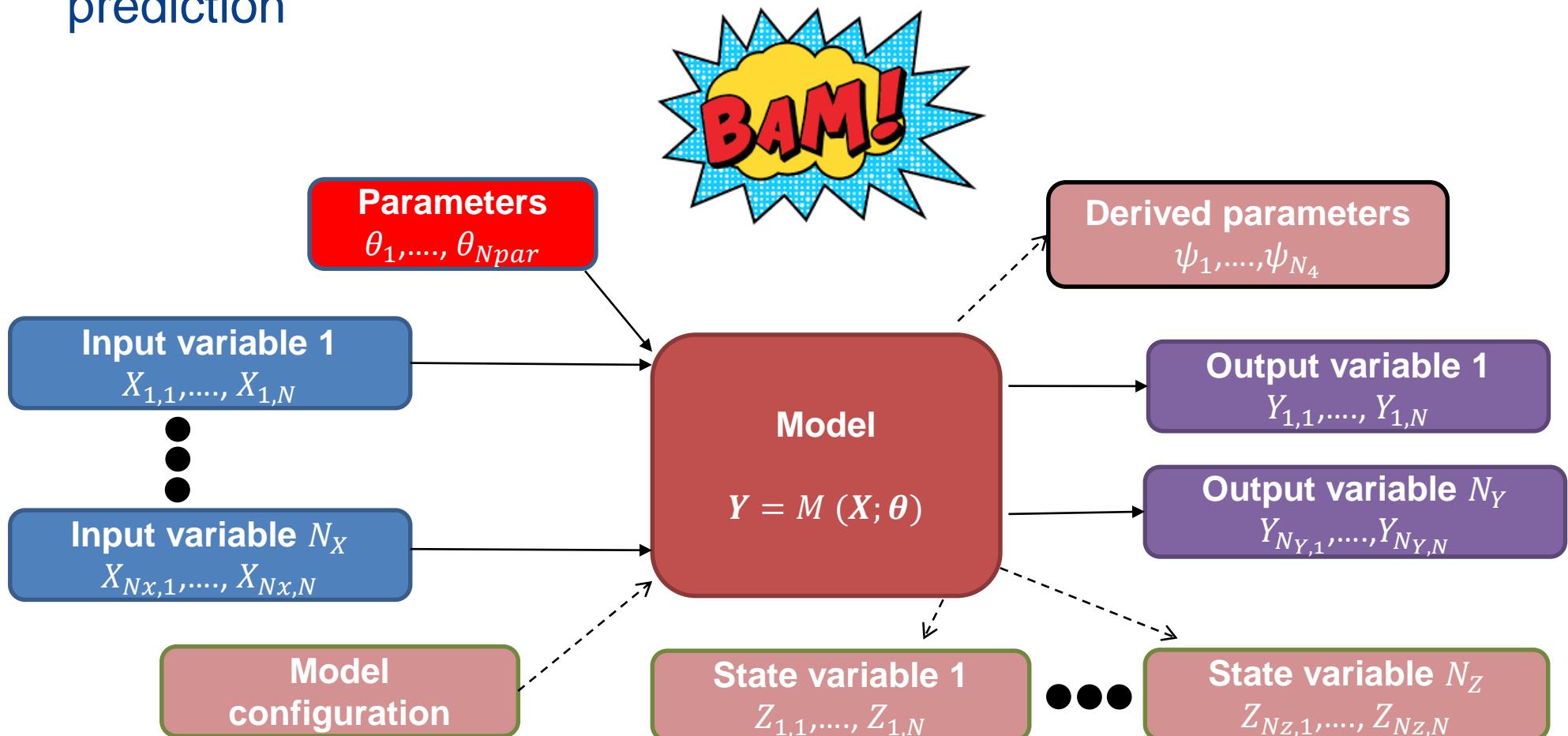
- Discharge time series
- Yellow box: Stage uncertainty (95% interval)
- Pink box: Stage + parametric uncertainties (95% interval)
- Red box: Total uncertainty (95% interval)

The BaRatin method for rating curves

- ✓ Introduction
- ✓ Hydraulic principles behind the rating curve
- ✓ Rating curve estimation
- ✓ Going further: more on BaRatin

Extension of BaRatin: *BaM!*

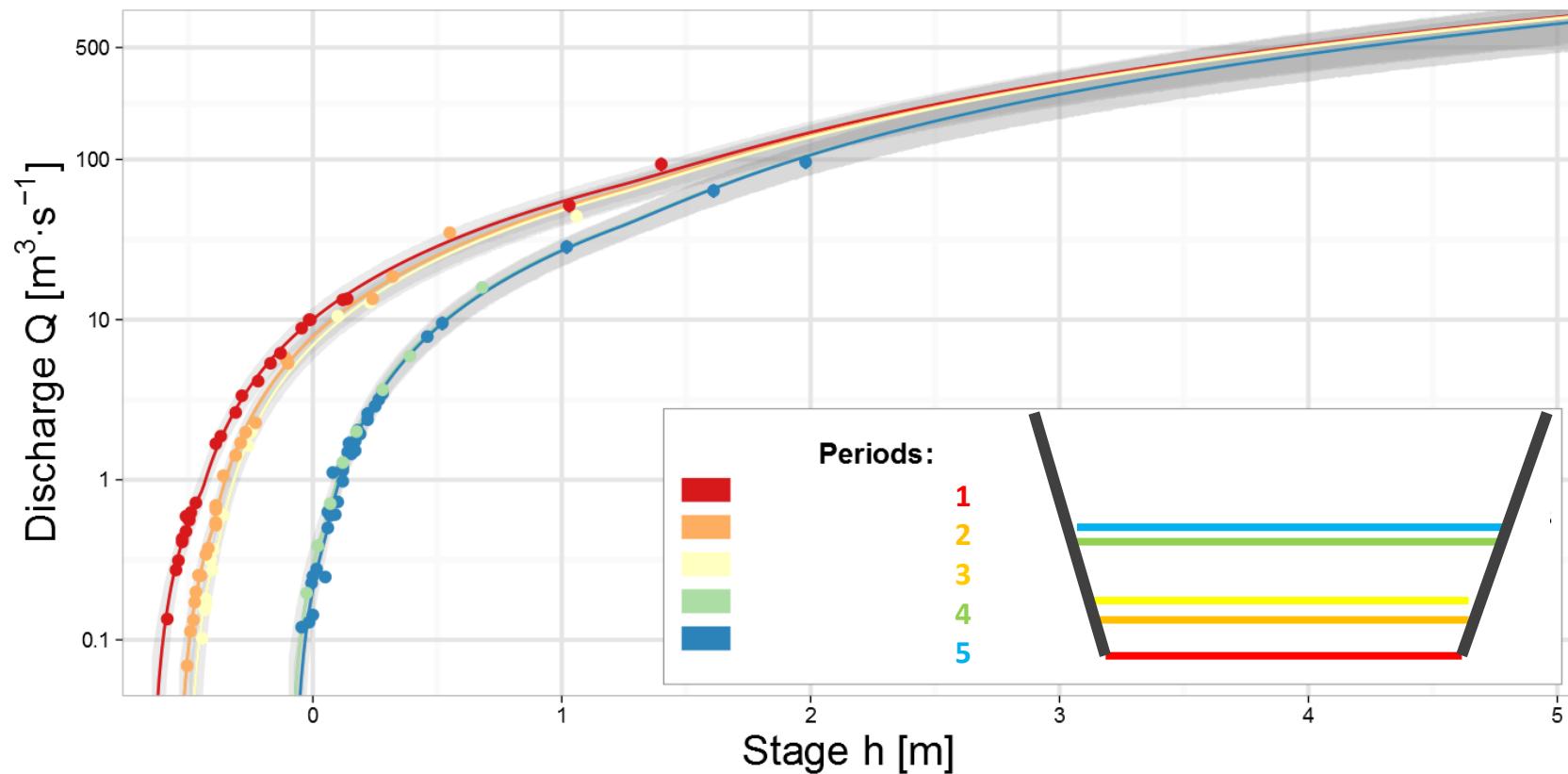
Calculation code for estimating any model and using it for prediction



Schematisation of *BaM!* (Renard, 2017)

Rating shifts due to bed evolution

BaRatin-SPD (stage-period-discharge): rating changes due to bed evolution at known times



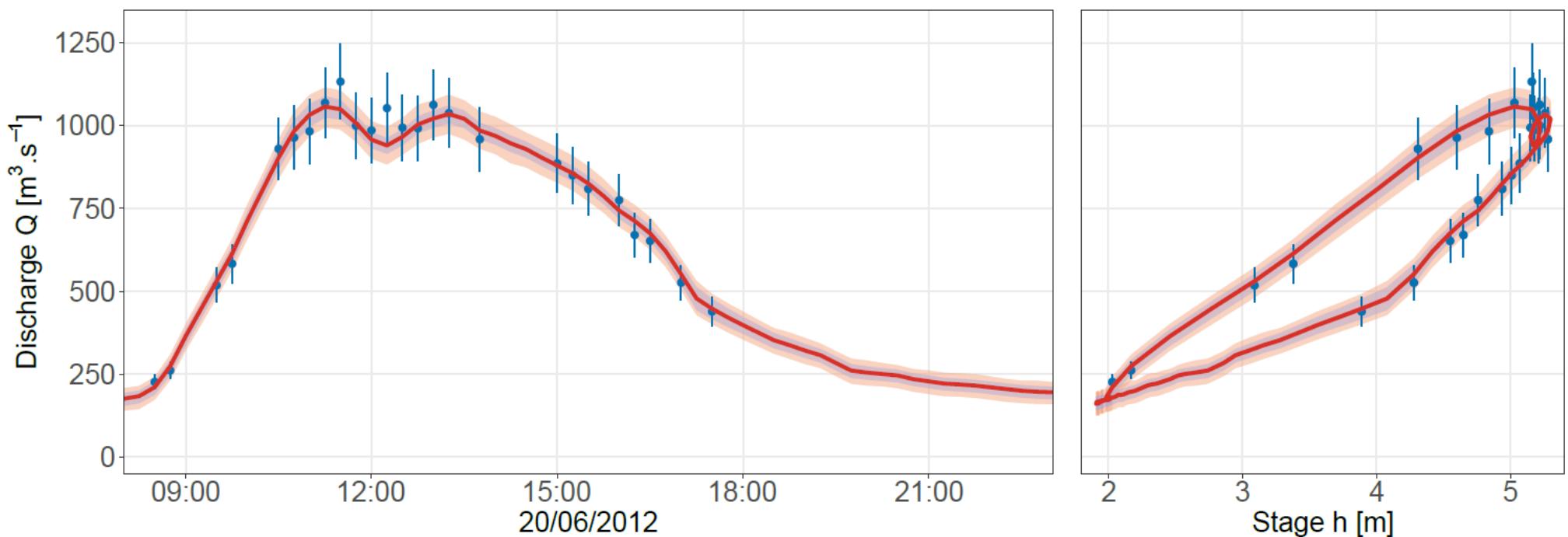
PhD of Valentin Mansanarez (2016)

Mansanarez et al. (2019)

$$Q(h, i_{\text{period}})$$

Complex rating curves

BaRatin-SGD (stage-gradient-discharge): hysteresis due to transient flow

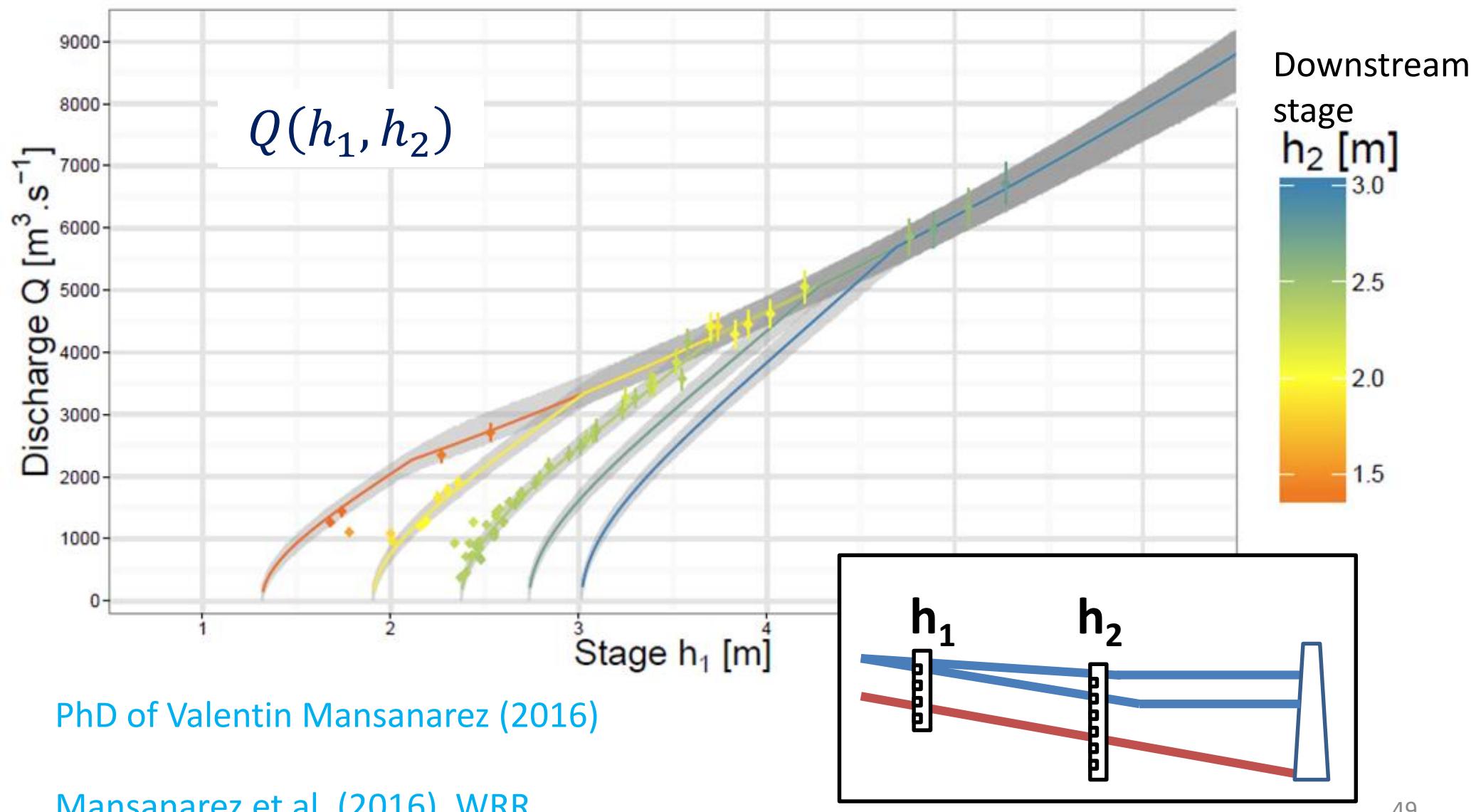


PhD of Valentin Mansanarez (2016)

$$Q \left(h, \frac{\partial h}{\partial t} \right)$$

Complex rating curves

BaRatin-SFD (stage-fall-discharge): variable backwater



Rating curves affected by aquatic vegetation

Modified channel control equation with vegetation roughness added

$$Q(h, t) = \frac{1}{\sqrt{n_b^2 + n_v^2}} B \sqrt{S_0} (h(t) - b)^c$$

