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Catchments exports and monitoring

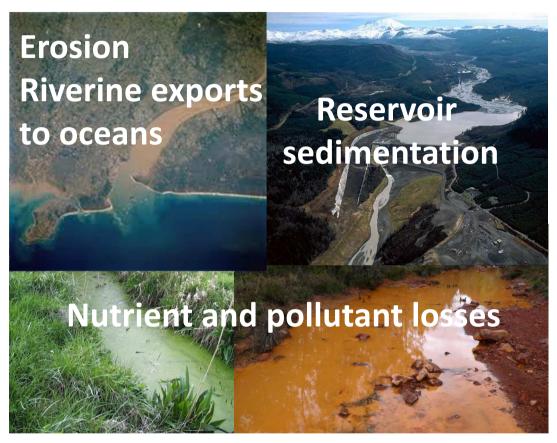
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Why estimating riverine exports ?

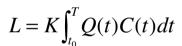


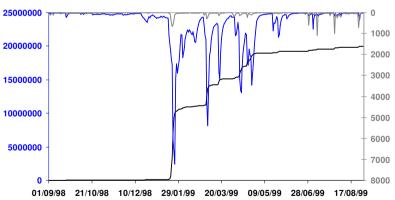


Why is challenging ?

Because a large proportion of annual load of most elements are exported during short time period

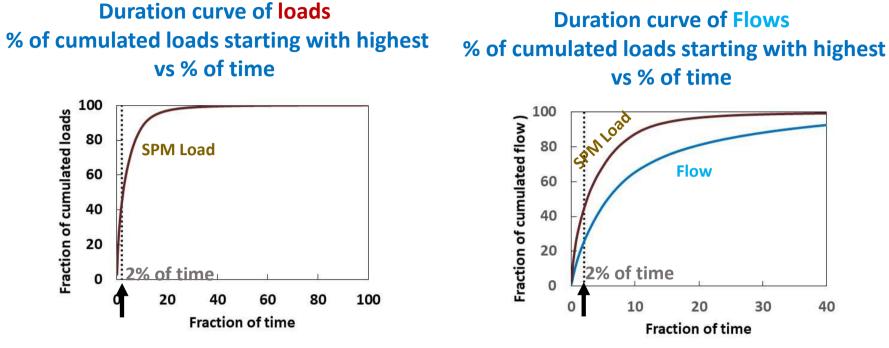
To better optimize sampling periods we need to better characterize load flashiness during these hot moments (high water periods)





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How to characterize load and flow flashiness?



M₂ : % of cumulated loads in 2% of time Good indicator to characterize load flashiness

W2%: % of cumulated flows in 2% of time Good indicator to characterize flow flashiness

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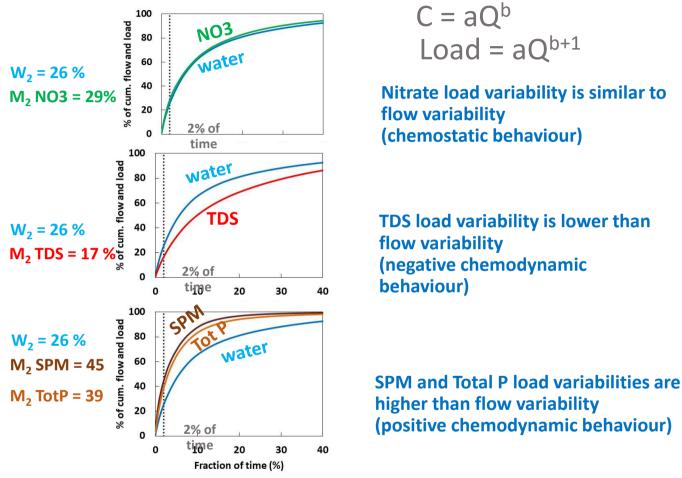
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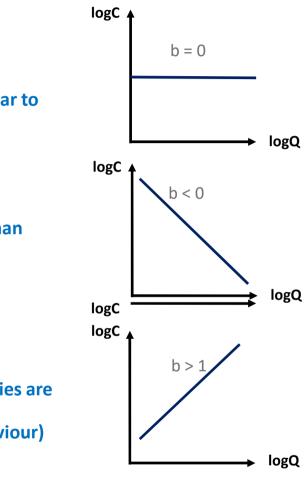


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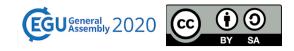


Load flashiness is linked to flow flashiness and type of elements

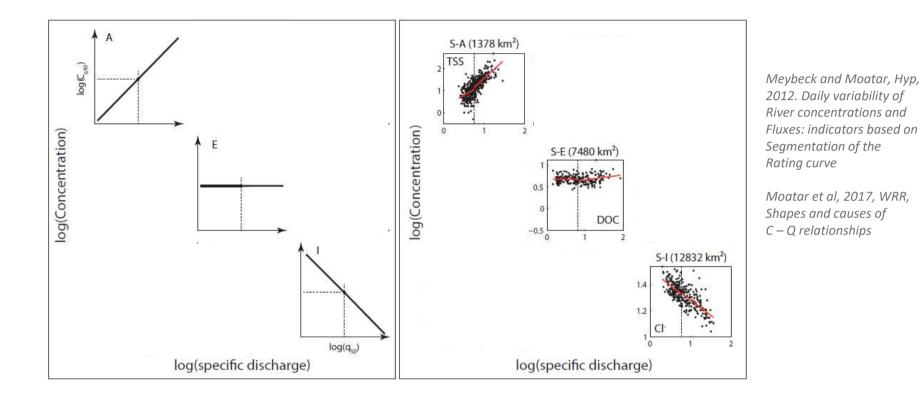




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Segmented C – Q relationships

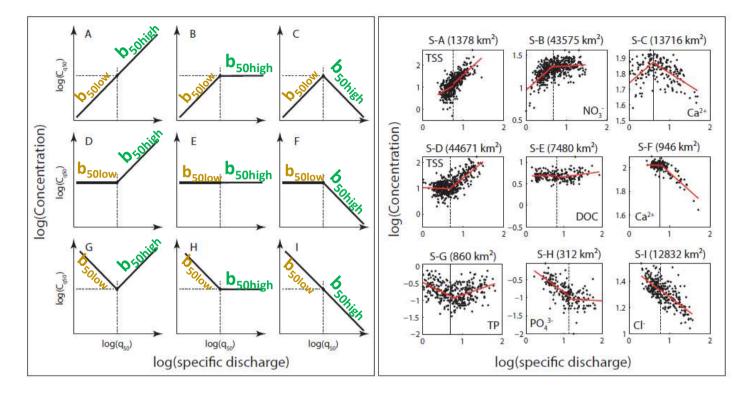


• Non-linear relationship on 60 % of catchment – element combination

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Segmented C – Q relationships : 9 types



Meybeck and Moatar, Hyp, 2012. Daily variability of River concentrations and Fluxes: indicators based on Segmentation of the Rating curve

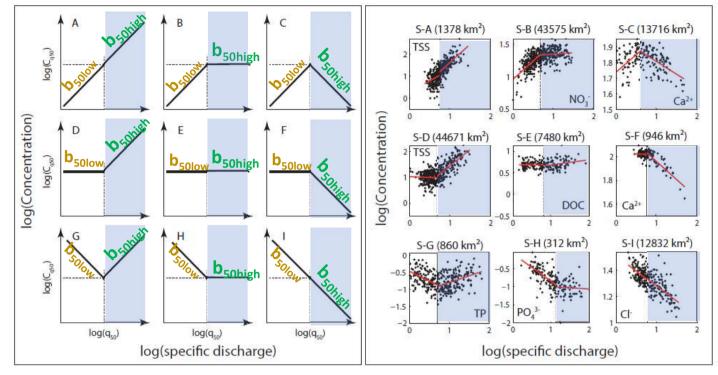
Moatar et al, 2017, WRR, Shapes and causes of C - Q relationships

- 9 potential C Q relationships and 2 or 3 dominant possibilities for each element
- **b**_{50low} biogeochemical processes during low flow period
- b_{50high} export pattern during high flow period



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For solute and sediment export regimes : importance of b_{50high}



Meybeck and Moatar, Hyp, 2012. Daily variability of River concentrations and Fluxes: indicators based on Segmentation of the Rating curve

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> Moatar et al, 2017, WRR, Shapes and causes of C - Q relationships

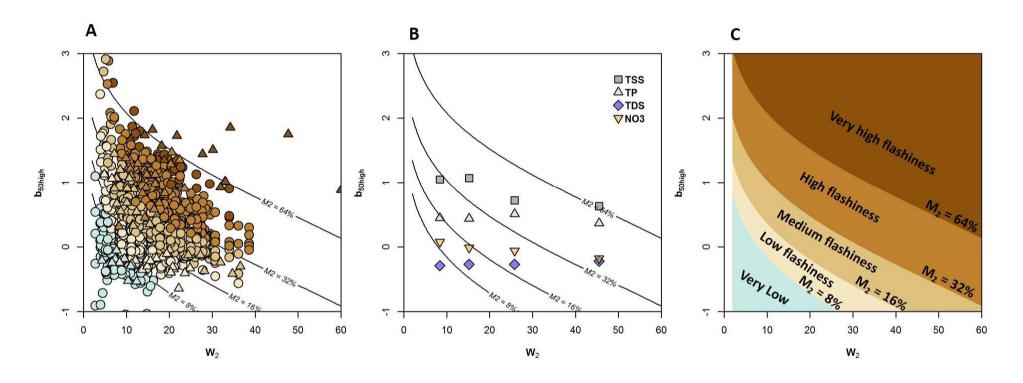
- 9 potential C Q relationships and 2 or 3 dominant possibilities for each element
- b_{50low} biogeochemical processes
- b_{50high} export pattern during high water (60 to 99% of annual load transported for Q > Q50)
- Therefore we use W₂ and b_{50high} to estimate M₂

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Load Flashiness (M2) Evaluation With Flow Flashiness (W2) and Export Pattern (b50high)

Probit(M2) = Probit(W2) + 0.79 b50high





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Moatar, F., et al. (2020). Stream Solutes and Particulates Export Regimes: A New Framework to Optimize Their Monitoring. Frontiers in Ecology and Evolution, 7. https://doi.org/10.3389/fevo.2019.00516

EGU 2020 Session 2.3.2. Data-driven analysis of water quality time series

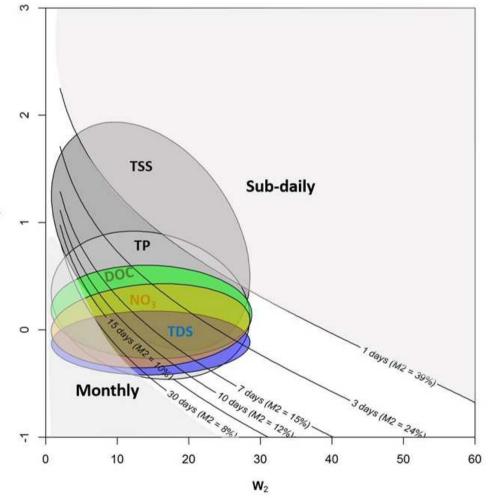


Optimizing Sampling Frequency for Reducing Load Calculation Uncertainty

Using the nomograph published by Moatar et al, HyP, 2013, for discharge-weighted concentration method And the relationship between M2, W2 and b50high

The optimal temporal monitoring frequency of the studied constituents decreases in the following order:

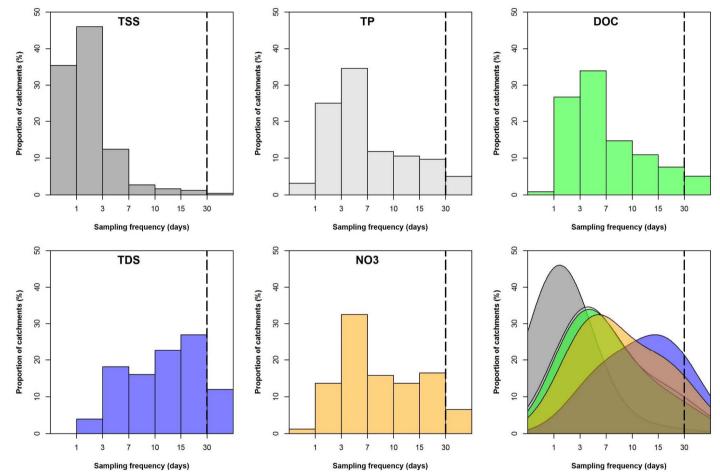




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Distribution of optimal sampling intervals required for 475 French stations for each parameter. The vertical dashed lines represent the current sampling frequency, i.e., monthly



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Catchment caracteristics can be used in a first approach to set up water quality monitoring design

TABLE 4 | Relationships between W₂, b_{50high}, and M₂ indicators and catchment characteristics: total percentage of variance explained by final regression models following a backward selection approach of explanatory variables (third column) and individual contributions (between 0 and 100%; next columns) of the selected variables according to hierarchical variation partitioning.

		Indicator/Constituent										
	Units	W ₂	b _{50high}					M ₂				
		-	TSS	TDS	TP	NO3	DOC	TSS	TDS	TP	NO3	DOC
% of explained variance	%	34.9	21.9	22.9	18.2	10.9	12.1	30.9	36.5	37.9	28.3	33.1
Area	km ²	21.3	9.6		15.5			12.9	16.9	7	10.8	19
Stream network density	km/km ²	25.4		11.7			20.6	24.9	15.7	16.6	39.1	19.7
Wetlands	%	7.3	19				20.9	1.4			6.8	13.5
Crystalline rocks	%		44.1	8	30.9	4.8	41.3	15.3	2.9		3.9	4.8
Low carbonate rocks	%	11.6		6.8		4		27.8	8.4	14.4		13.7
ligh carbonate rocks	%	5.1		12.7		5.6			4.5	2		
Riparian vegetation	%						4.6		5.8			
Forest	%	10.2			12.8	20.5		4.2			8.2	10.1
Extensive agriculture	%	2.6				26.8	3.7	2.8		3.2	3.3	3.3
ntensive agriculture	%	13.6		18.5		27.1		8.9	31.1	33.9	14	15.9
Urban area	%		17.2	42.2				1.7	10.9	5		
Population density	ind/km ²	3				11.2						
Point source P	kg/ha/yr				22.8					9.8		
Soil P	g/kg									8.1		
Point source N	kg/ha/yr										7.8	
N surplus	kgN/kg										6.2	
Erosion risk	%		10.2		17.9		8.9		3.8			

Blue and red cells indicate positive and negative relationships, respectively. Bold numbers highlight variables with the greatest influence (cumulated contribution >50%).

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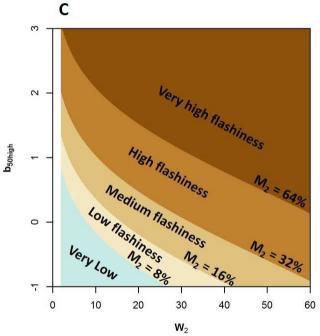
> Conclusion

Solute and sediment export regimes are quantified by the load flashiness (M_2), which can be easily determined from flow flashiness (W_2) and b_{50high} (from high frequency data or long-term low frequency)

The load flashiness diagram can be used to:

- classify elements, catchments,
- quantify human impacts and mitigations,
- optimize sampling strategies
- calculate uncertainties of long-term monitoring surveys

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