

Preserving the space-time dependence structure of  
precipitation for hydro-meteorological forecasting:

A case study with analog-derived PQPF

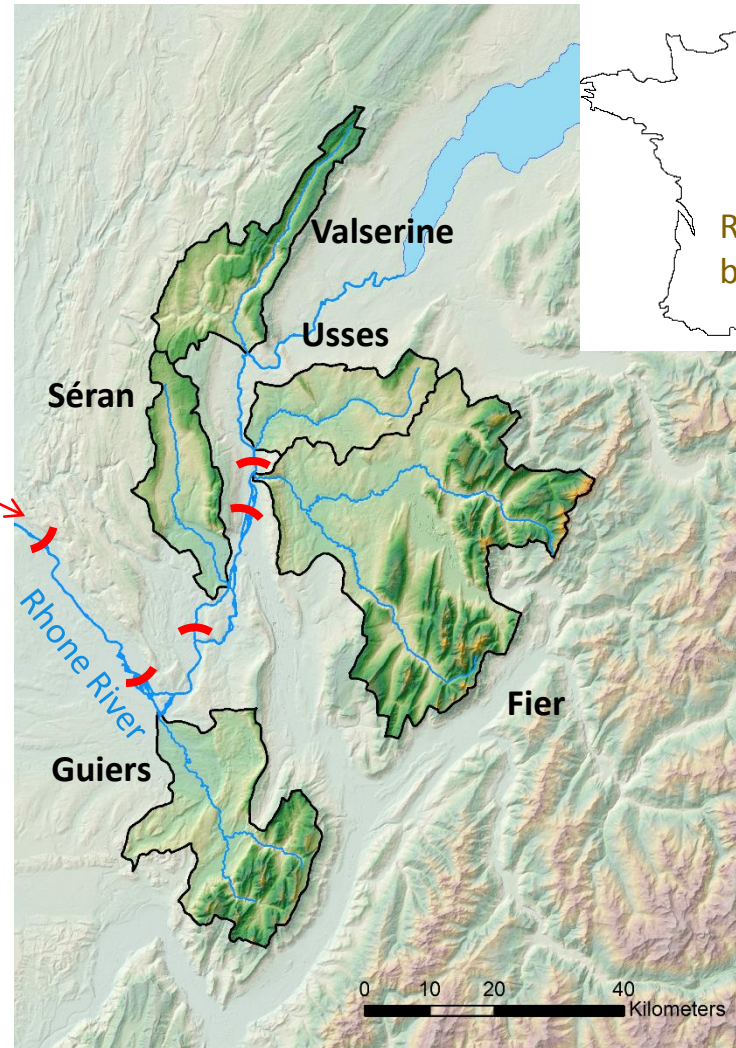
**Joseph Bellier** <sup>1\*</sup>, Isabella Zin <sup>1</sup>, Guillaume Bontron <sup>2</sup> and Stanislas Siblot <sup>2</sup>

<sup>1</sup> LTHE, Université Grenoble Alpes, Grenoble, FRANCE

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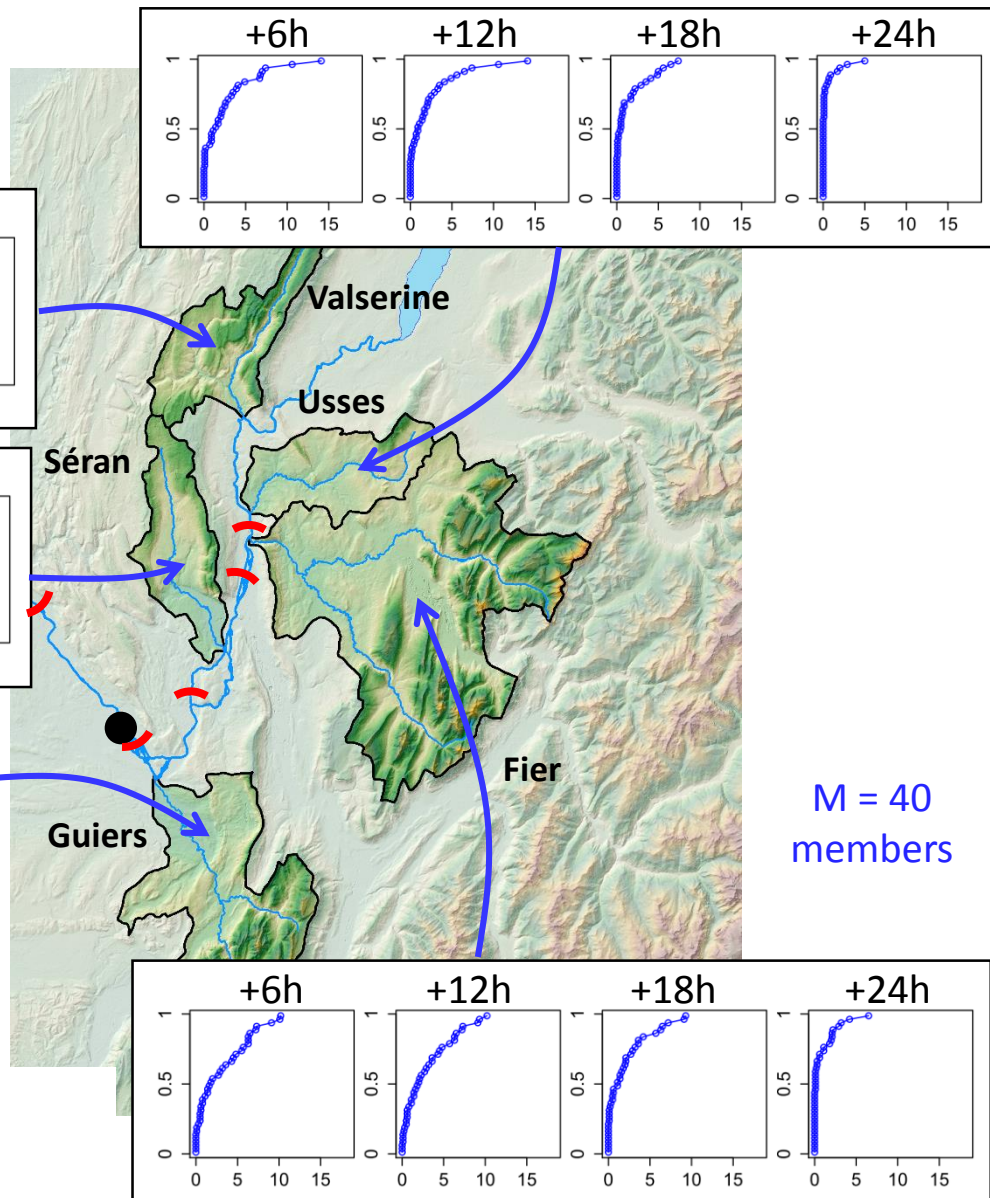
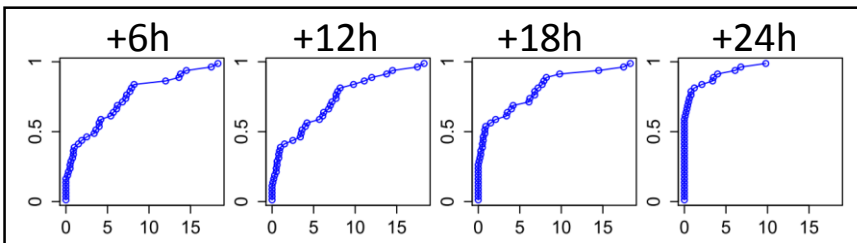
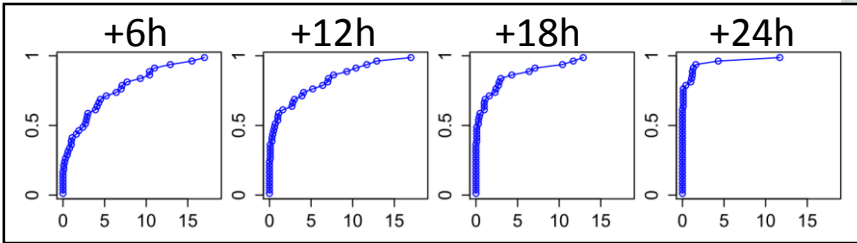
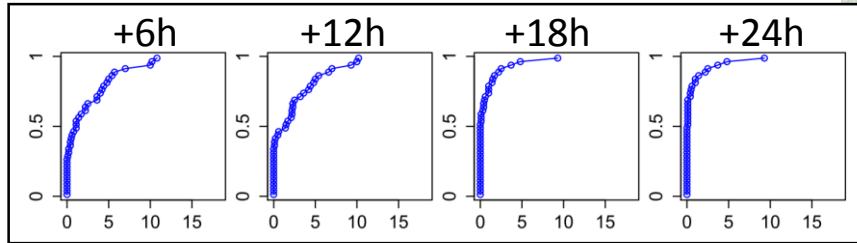
\* Corresponding author: [joseph.bellier@univ-grenoble-alpes.fr](mailto:joseph.bellier@univ-grenoble-alpes.fr)

 hydropower dams



**Case study:**  
5 catchments of the Rhone River:

Required **precipitation forecasts** for a +24h streamflow forecast :



5 catchments }  
4 lead times } **20 univariate forecasts**  
**1 multivariate forecast**

## Problem:

Because of  $\left\{ \begin{array}{l} \text{post-processing} \\ \text{scale mismatch,} \\ \dots \end{array} \right.$   $\Rightarrow$  Wrong (or any) **space-time dependence structure**

How can we reconstruct the space-time coherence?

How can we evaluate it?

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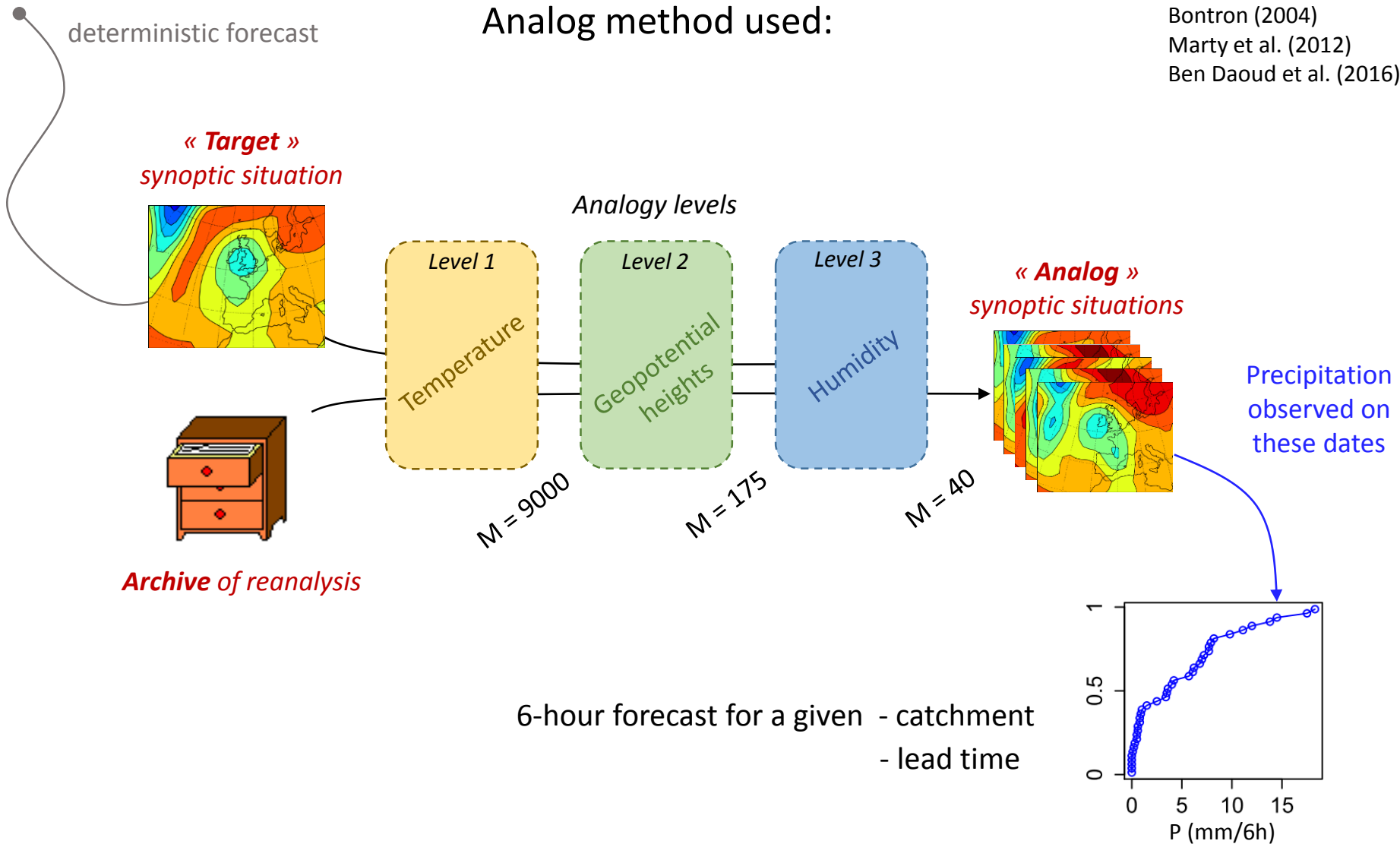
How can we evaluate it?

## Case study :

Precipitation forecasts from an analog method

**Papers:**  
Obled et al. (2002)  
Bontron (2004)  
Marty et al. (2012)  
Ben Daoud et al. (2016)

## Analog method used:



## Problem:

Because of  $\left\{ \begin{array}{l} \text{scale mismatch,} \\ \text{post-processing,} \\ \dots \end{array} \right. \Rightarrow$  Wrong (or not all) **space-time dependence structure**

## How can we reconstruct the space-time coherence?

### I. Reordering methods:

- Existing methods
- Adaptation of the Schaake shuffle using analogs

## How can we evaluate it?

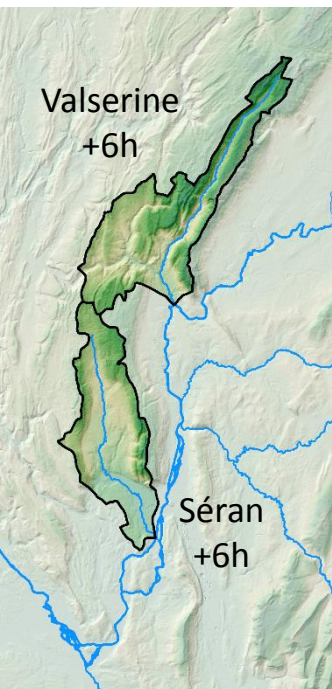
### II. 3-step verification:

- Correlations check
- Multivariate verification score
- Verification on streamflow

## Case study :

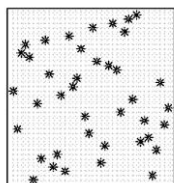
Precipitation forecasts from an analog method

Exemple in 2D:

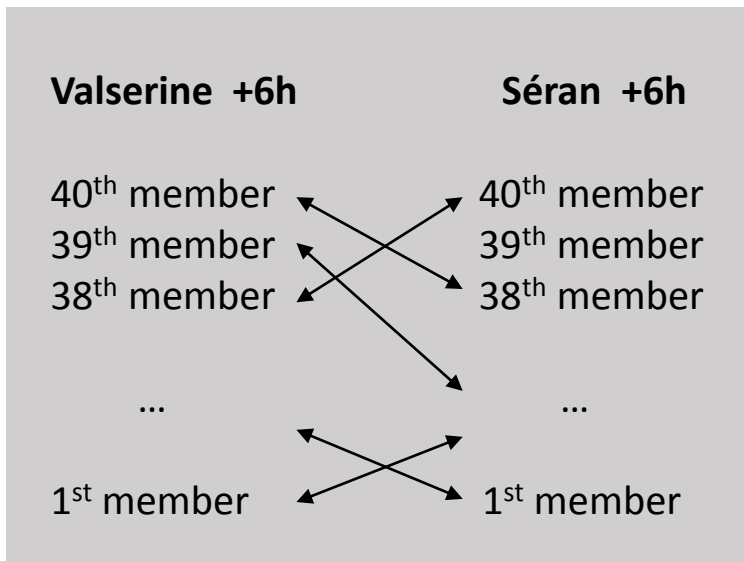


## What is a reordering?

**Ranks :**

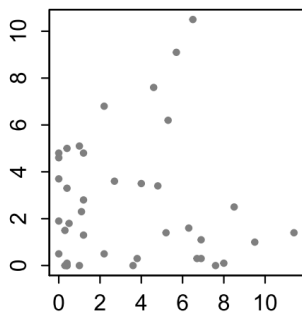


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**Univariate distributions**

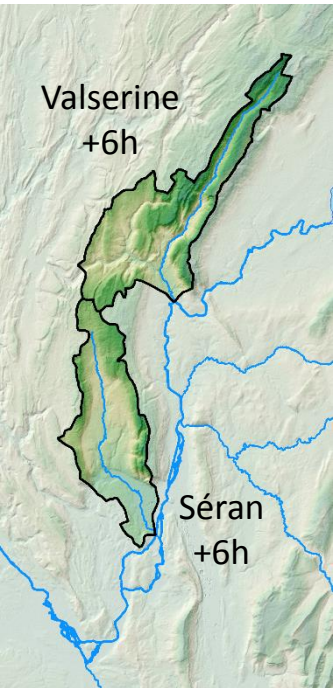
Valsérine +6h  
Sérán +6h



**Multivariate forecast**



Exemple in 2D:

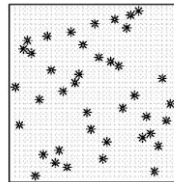


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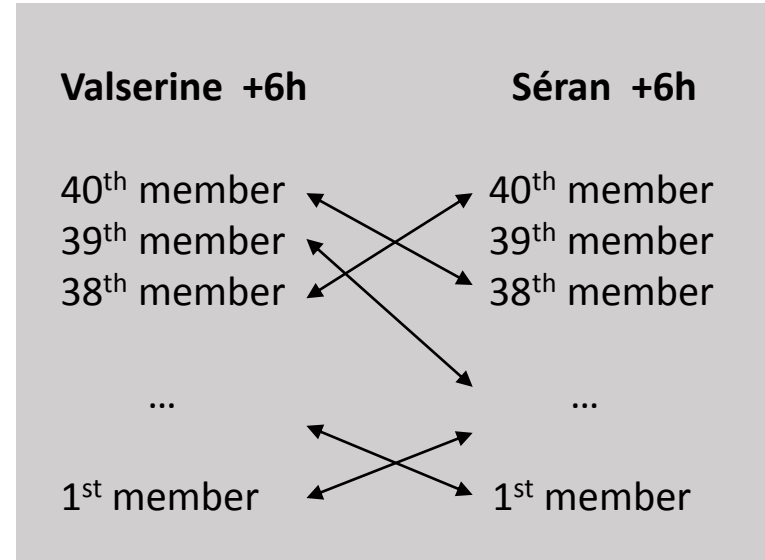
**Sample:**



**Ranks :**

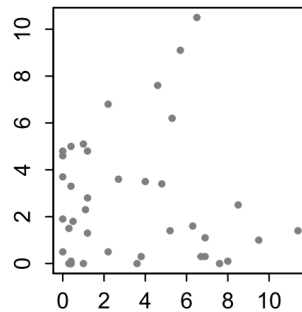


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**Univariate distributions**

Valsérine +6h  
Sérán +6h



**Multivariate forecast**

# I. Reordering methods

Exemple in 2D:



Clark et al. (2004)

Schefzik et al. (2013)

Random

Schaake shuffle

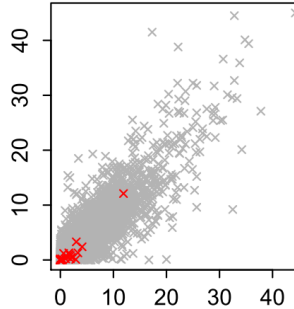
Ensemble copula coupling (ECC)

Sample:

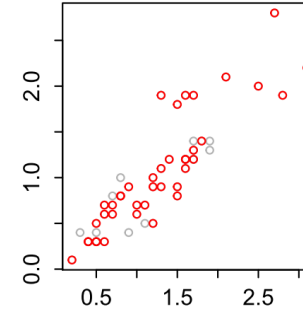
No sample



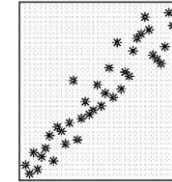
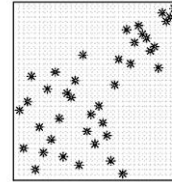
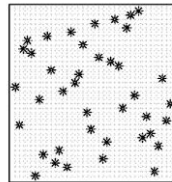
Sample from climatology



Sample from a raw ensemble forecast

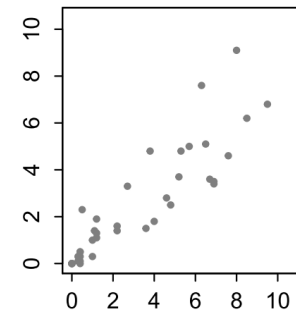
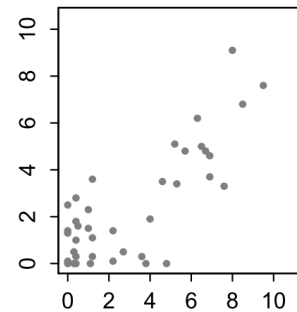
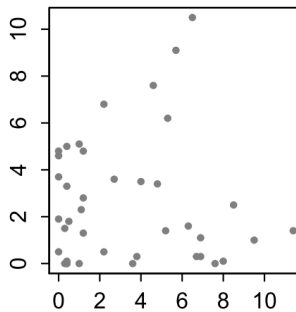


Ranks :



Univariate distributions

Valsérine +6h  
Sérán +6h



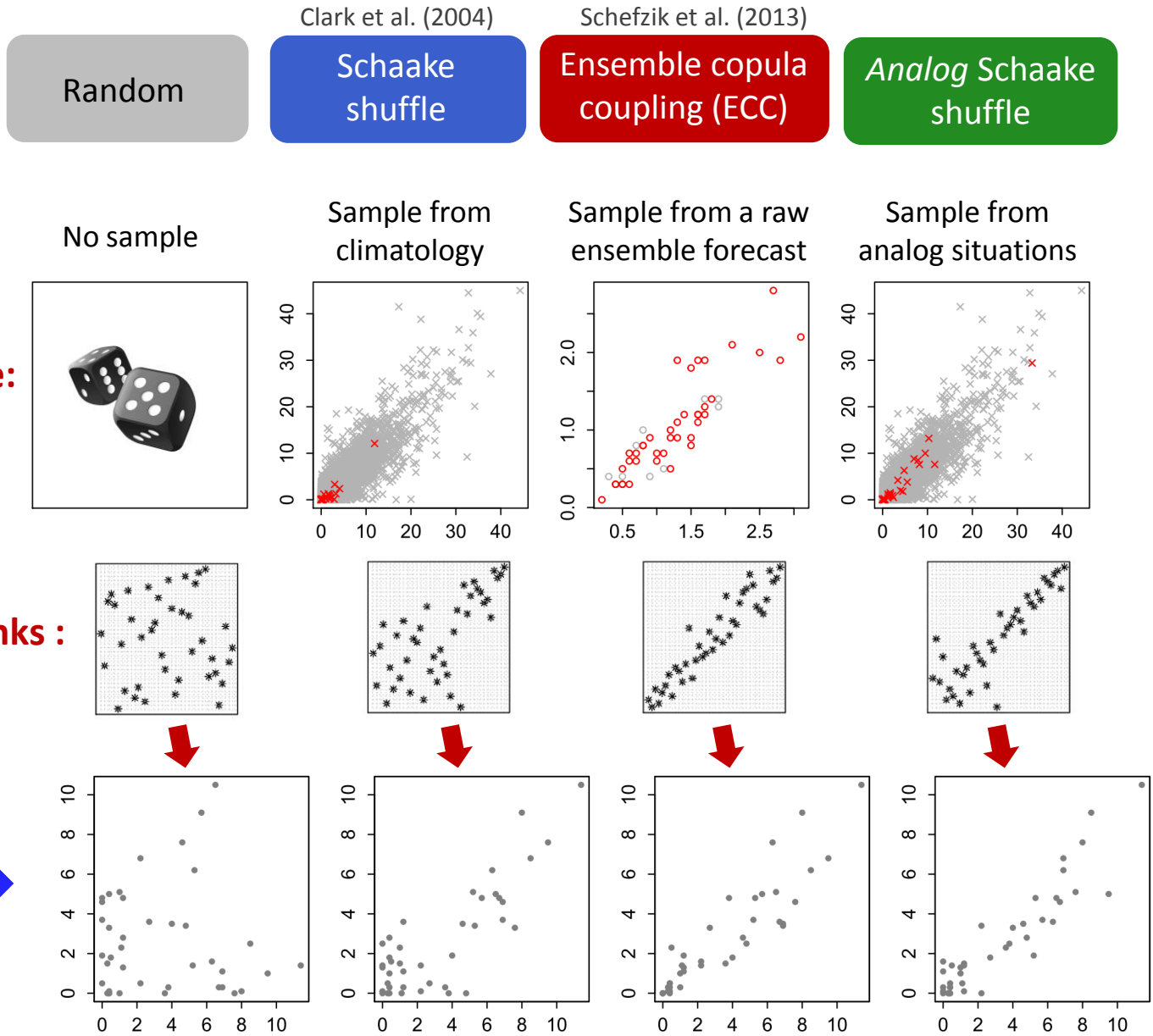
# I. Reordering methods

Exemple in 2D:



**Univariate distributions**

Valsérine +6h  
Sérán +6h



## How can we evaluate these methods?

Verification period: 2008-2014  
1 forecast/day (00 UTC)



## 1. Check of space-time climatological correlations:

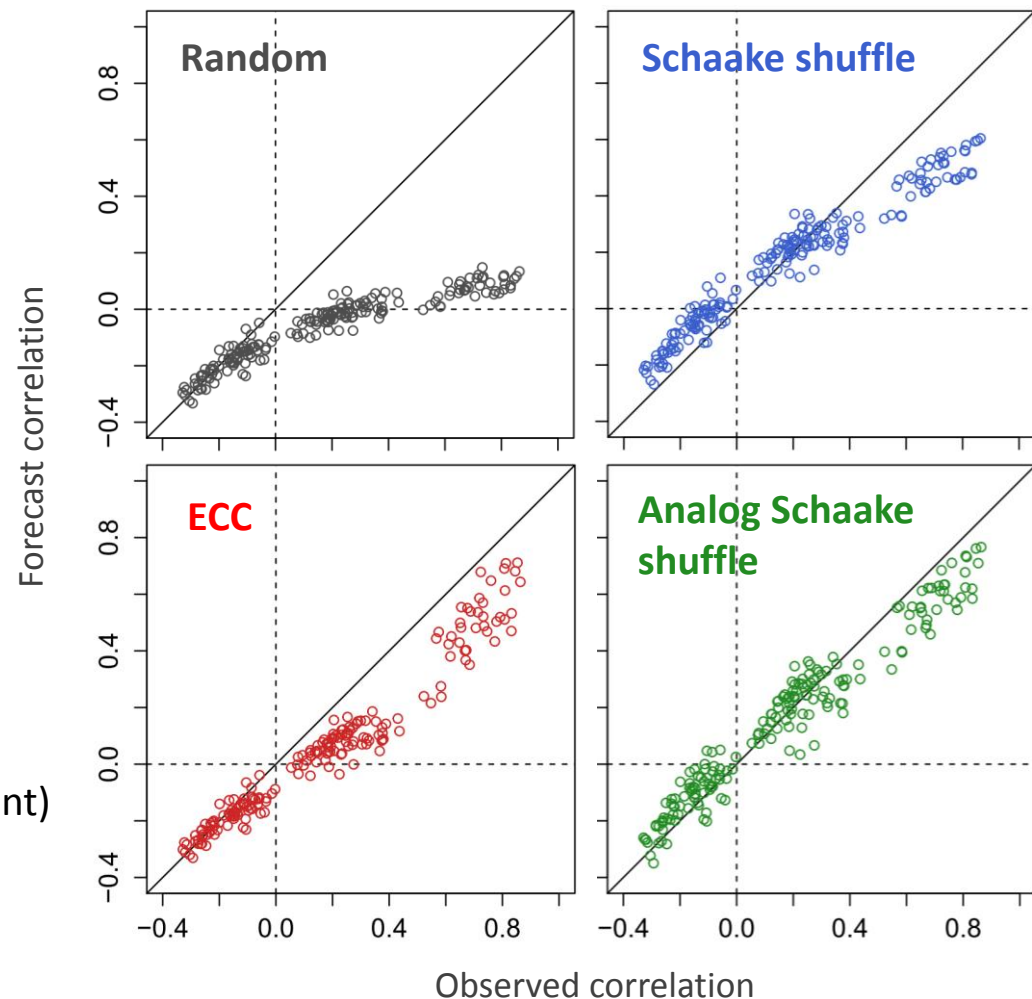
Points on the diagonal = **Good**

Multivariate forecast of dimension 20  
(5 catchments  $\times$  4 lead times)

$\Rightarrow$  190 possible pairs of dimension  
 $\Rightarrow$  190 points

X-coordinate = Observed correlation  
Y-coordinate = Forecast correlation

(using Spearman's rank correlation coefficient)

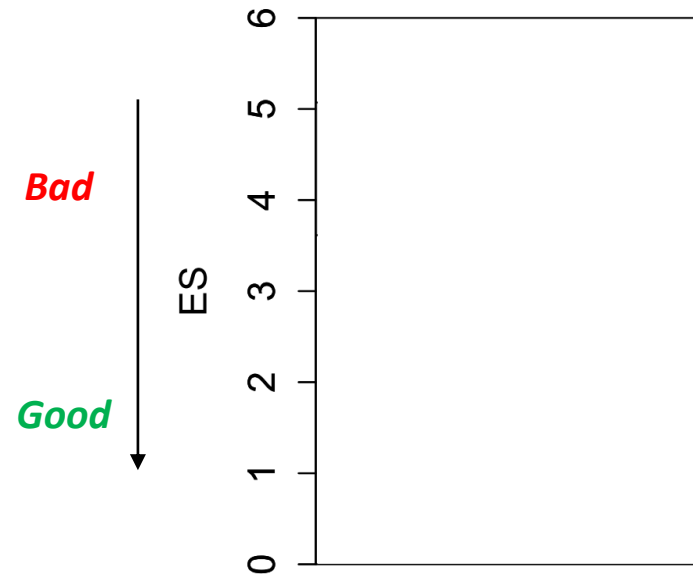


## 2. Use of a multivariate verification score:

- **Energy Score (ES):** extension of the CRPS in dimensions  $> 1$ .

(Gneiting and Raftery, 2007)

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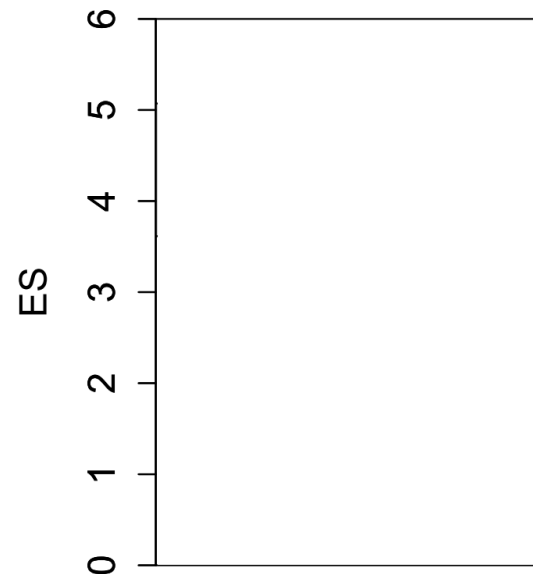
⇒ For each multivariate forecast of **40 members**  
in **dimension 20** (5 catchments  $\times$  4 lead times):

$(40!)^{20-1}$  possible reordering

→  $(40!)^{20-1}$  different ES

*Bad*

*Good*



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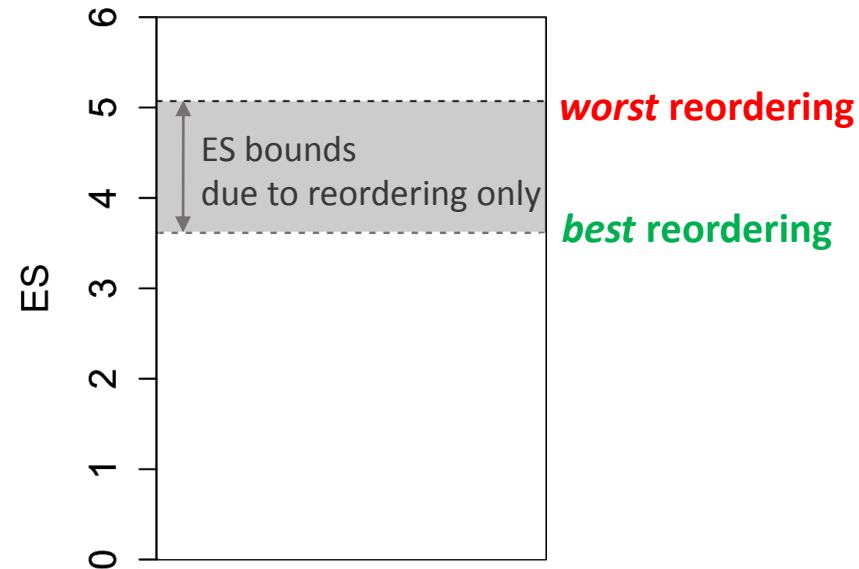
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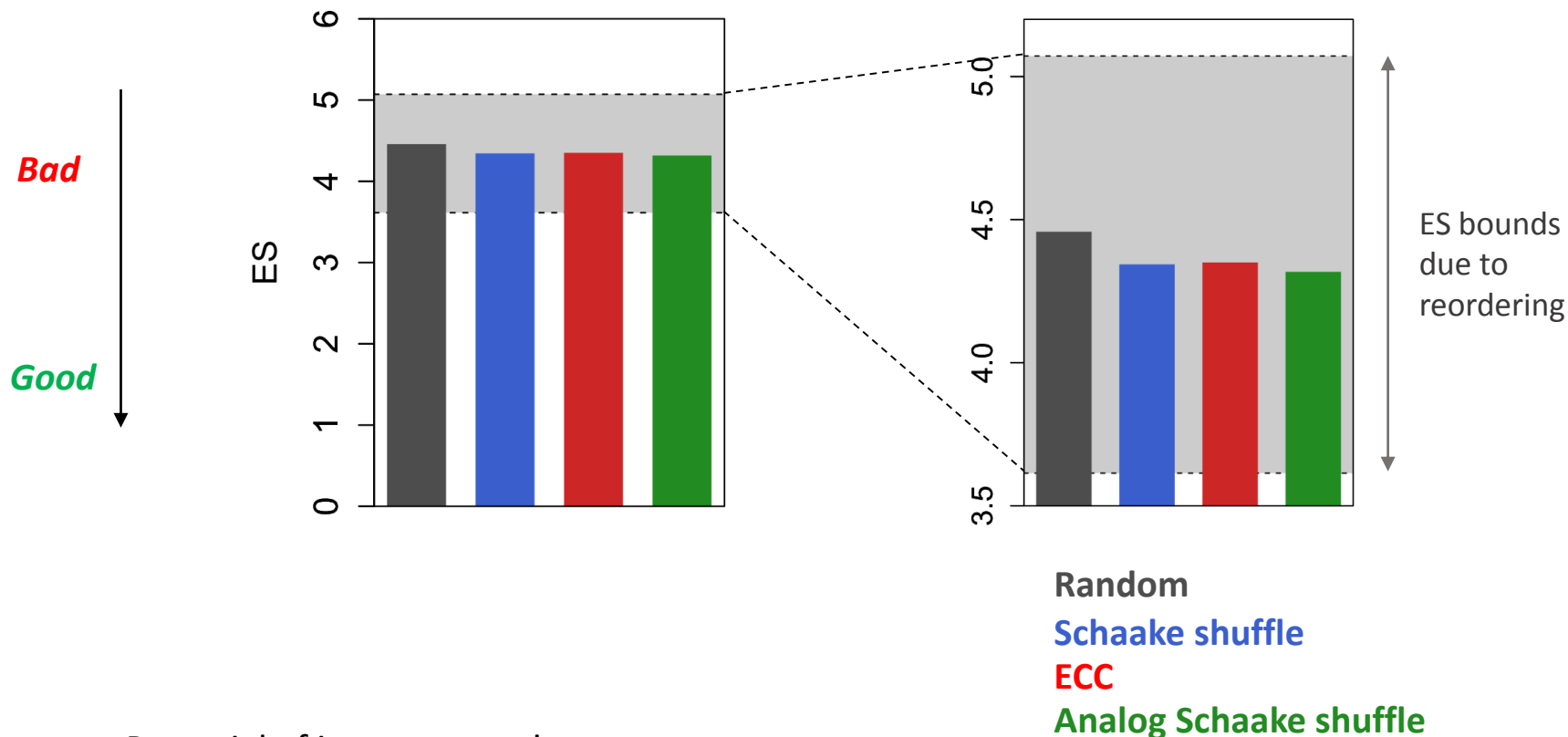
*Bad*

*Good*



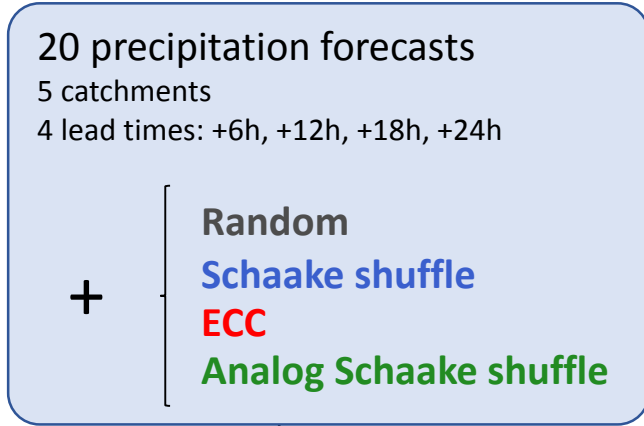


Mean ES over all forecasts:

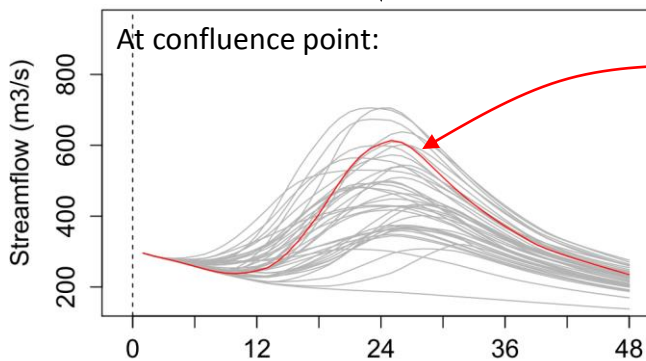


⇒ Potential of improvement!

### 3. Verification on streamflow

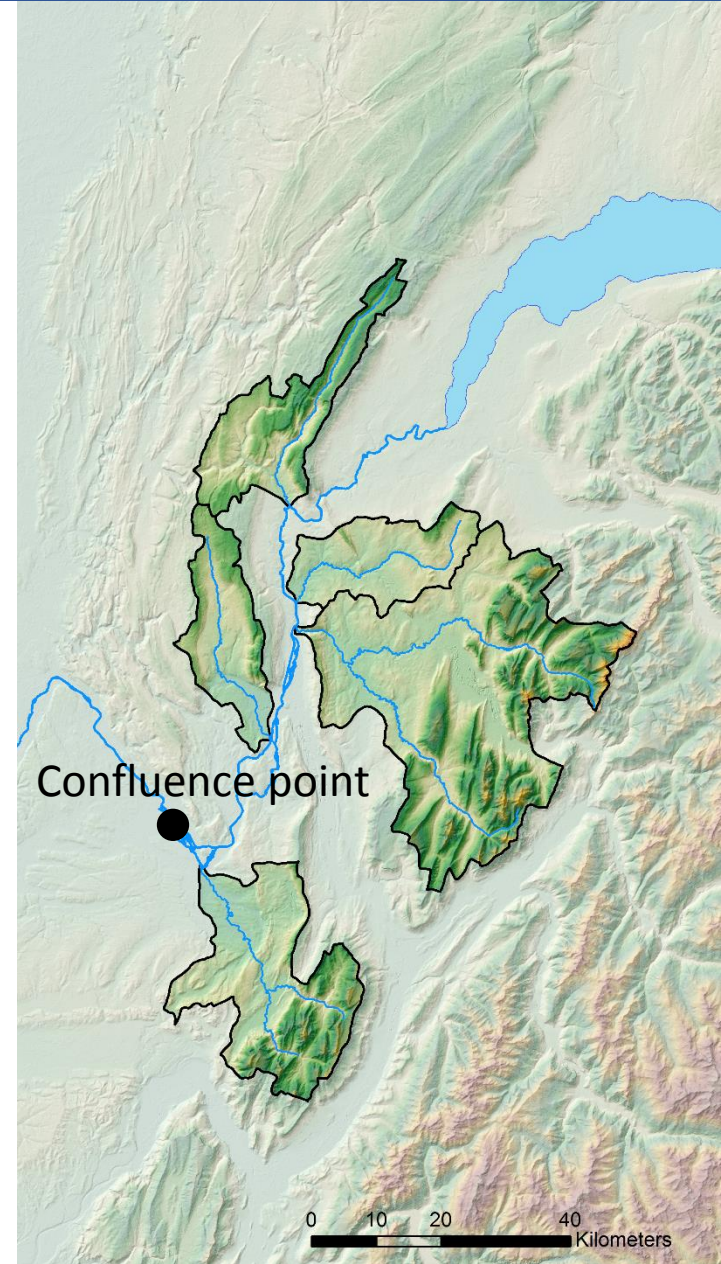


5 lumped hydrological models + propagation model



Simulated streamflow  
 (with observed precipitation)

⇒ Cancels out hydrologic uncertainty



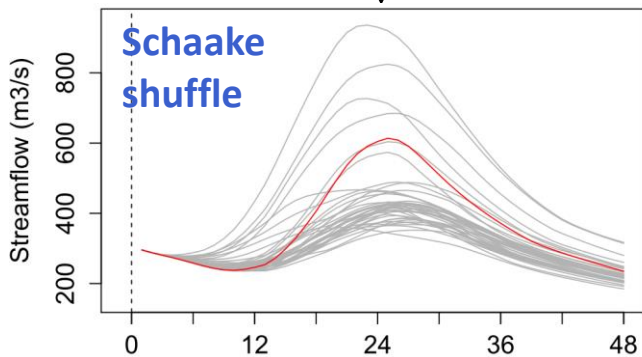
### 3. Verification on streamflow

20 precipitation forecasts  
5 catchments  
4 lead times: +6h, +12h, +18h, +24h

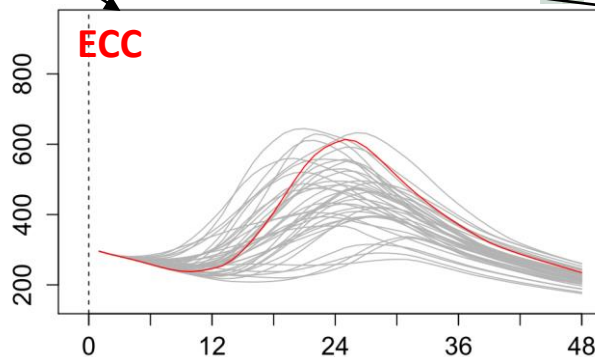
+

Random  
Schaake shuffle  
ECC  
Analog Schaake shuffle

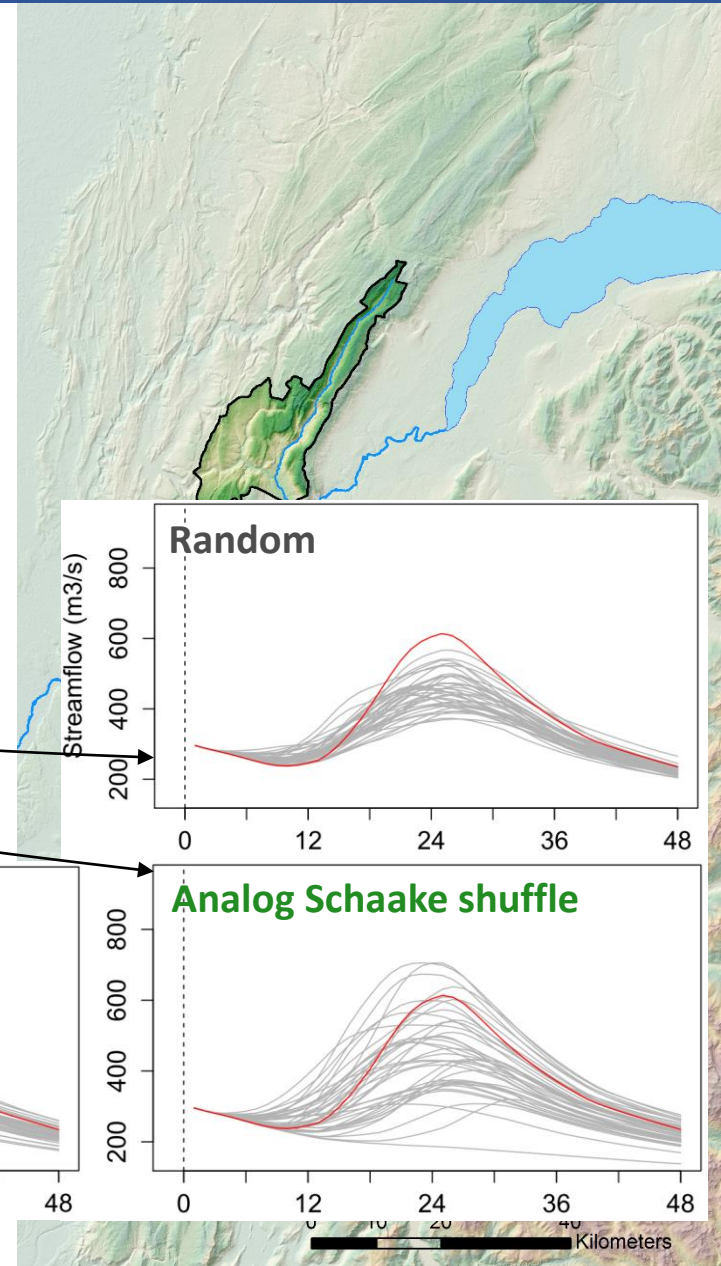
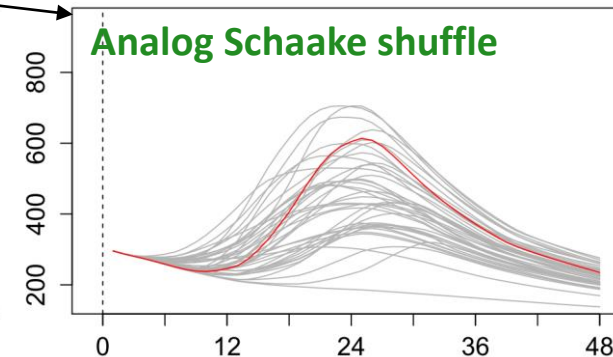
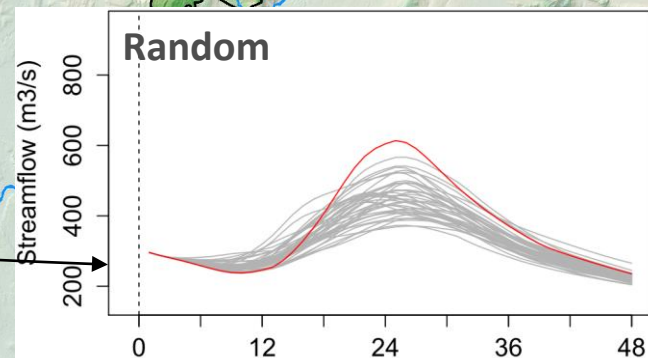
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08/06/2016

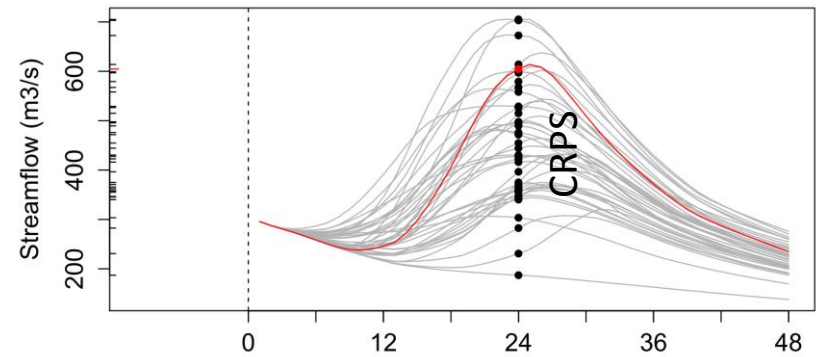


2016 HEPEX Workshop - Québec



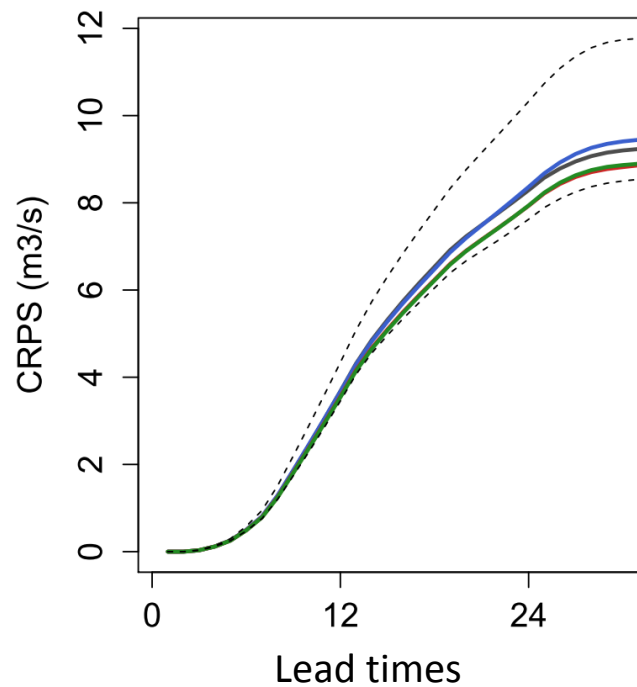
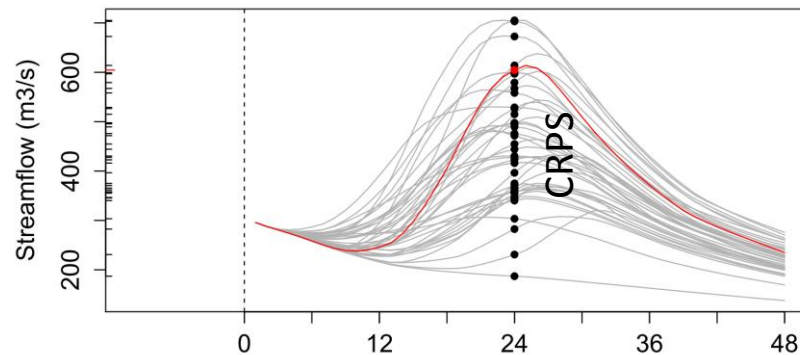
What is the variable of interest?

In a **hydropower production forecasting** context:



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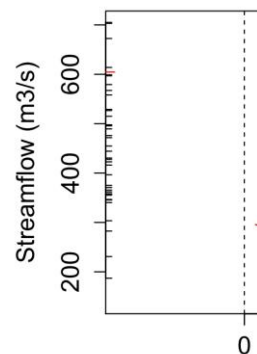
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Random  
 Schaake shuffle  
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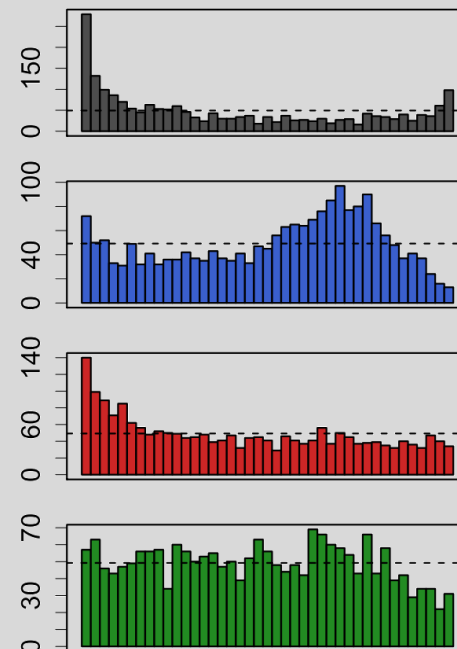
In a **hydropower production forecasting** context:



Lead time: +24h

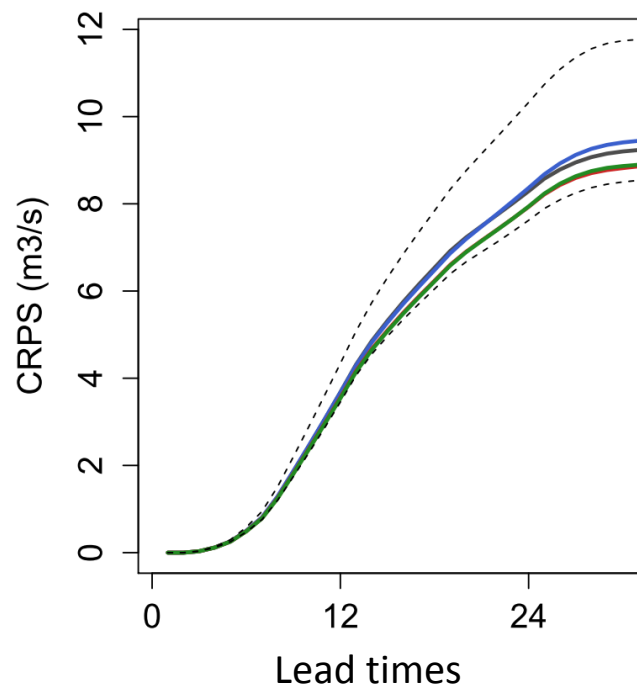
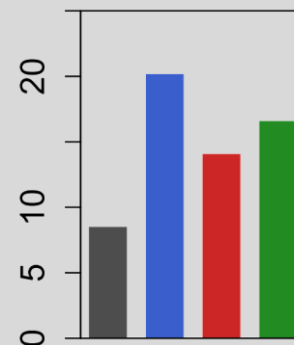
Calibration

Rank histograms:



Sharpness

Standard deviation (m3/s)



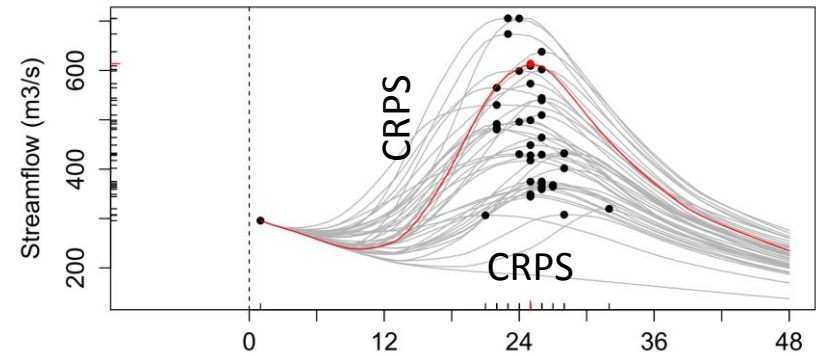
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*Bad*

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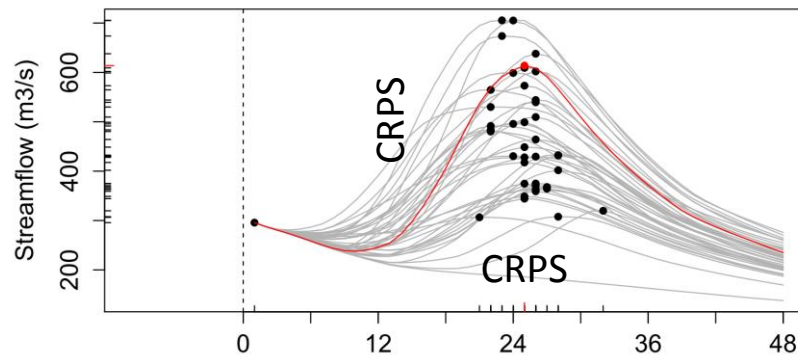
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In a **flood warning** context:

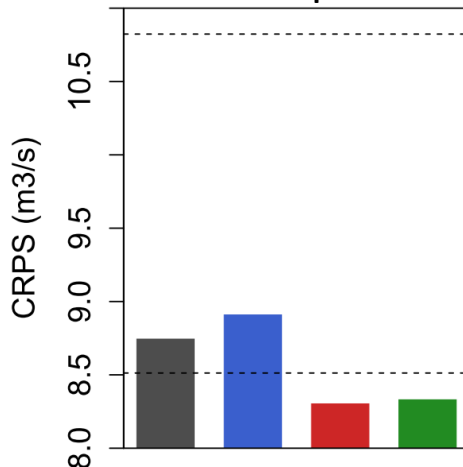


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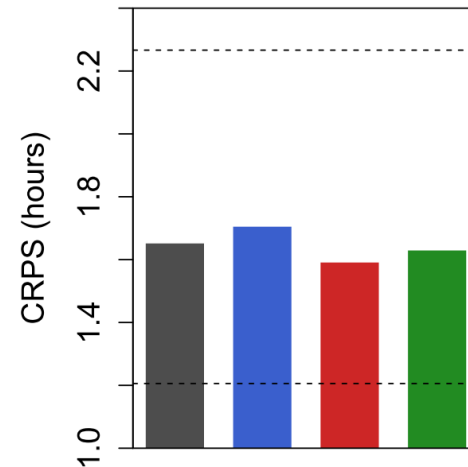
Peak amplitude



with **worst** reordering (according to ES)

with **best** reordering (according to ES)

Peak timing

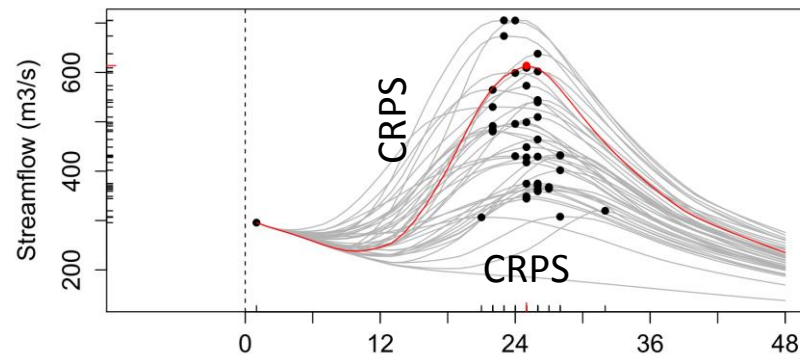


Random  
 Schaake shuffle  
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 Analog Schaake shuffle



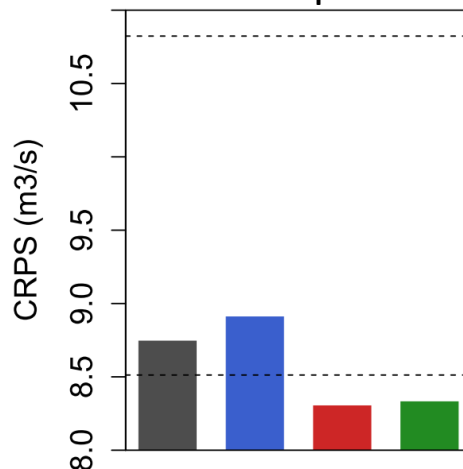
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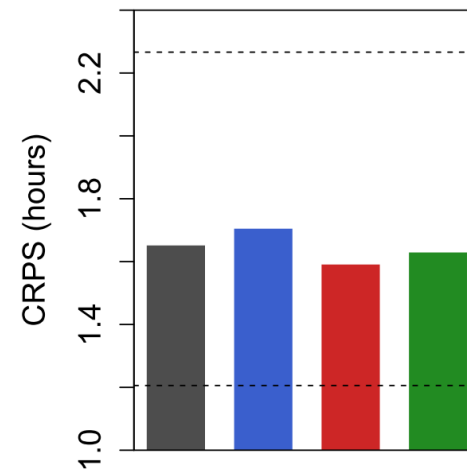
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with **worst**  
 reordering  
 (according to ES)

with **best**  
 reordering  
 (according to ES)

Peak timing



- ⇒ Is the ES the best metric to evaluate precipitation forecasts for hydrological purposes?
- ⇒ Best precipitation forecasts always lead to best streamflow forecasts?

## Conclusion

- Existing reordering methods

+ **Adaptation of the Schaake shuffle**  
Reordering depending on the atmospheric situations

- **3-step verification strategy:**

1. Correlation check

⇒ Necessary but not sufficient (climatology only)

2. Multivariate score

⇒ In case of no *a priori* about the use of forecasts

3. Verification on streamflow

⇒ Necessary since conclusions may be different!

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## Perspectives

⇒ Search numerically for **“best”** and **“worst” reordering** on **streamflow forecasts**.

→ To evaluate the potential of improvement

→ To conclude about the impact of reordering on streamflow forecasting

→ To better understand links between **Energy Score** and **Streamflow verification**

⇒ With other forecasts (including Temperature?)

Thank you for your attention

Example of a streamflow forecast:  
(12/30/2009 00:00)

