

# Verification of extremes using proper scoring rules and extreme value theory

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# Plan

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- 1 Extremes : difficult to forecast... and to verify
- 2 Weighted CRPS for extremes
- 3 Extreme Value Theory and CRPS distribution
- 4 A relevant case study

# Verification & extremes : a challenging issue

## ▶ **Verification habits**

- ▶ Set of observed events and associated forecasts
- ▶ Standard verification methods applied on the set

## ▶ **But for extremes**

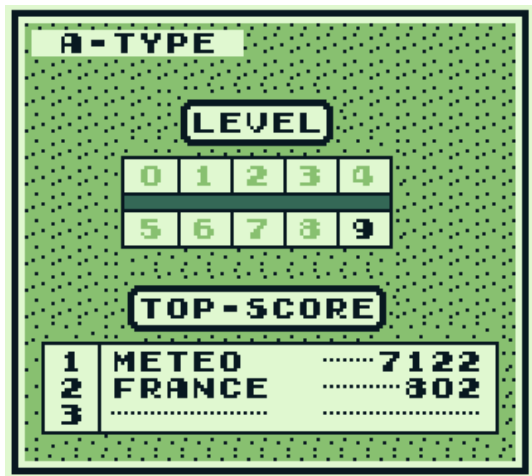
- ▶ Small number of observed events
  - ▶ Standard verification methods degenerate
  - ▶ Models (even ensemble forecasts) are usually quite bad
- ▶ Misguided inferences/assessments : *The forecaster's dilemma*  
(see Sebastian's talk)

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# Proper scoring rules



# Proper scoring rules

- ▶  $Y$  : observation with CDF  $G$  (unknown...)
- ▶  $X$  forecast with CDF  $F$
- ▶  $s(.,.)$  function of  $\mathcal{F} \times \mathbb{R}$  in  $\mathbb{R}$

$s$  is a proper scoring rule (Murphy 1968 ; Gneiting 2007)

$$\mathbb{E}_Y(s(G, Y)) \leq \mathbb{E}_Y(s(F, Y)) \quad (1)$$

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## The CRPS...

- ▶ A widely used proper score : the CRPS (Murphy 1969 ; Gneiting and Raftery 2007 ; Naveau et al. 2015 ; Taillardat et al. 2016)

$$\begin{aligned}
 CRPS(F, y) &= \int_{-\infty}^{\infty} (F(x) - \mathbf{1}\{x \geq y\})^2 dx \\
 &= \mathbb{E}_F |X - y| - \frac{1}{2} \mathbb{E}_F |X - X'| \\
 &= y + 2 [\bar{F}(y) \mathbb{E}_F(X - y | X > y) - \mathbb{E}_F(XF(X))] \\
 &= \mathbb{E}_F |X - y| + \mathbb{E}_F(X) - 2\mathbb{E}_F(XF(X))
 \end{aligned}$$



## ... And its weighted derivation

- ▶ A weighted score : the wCRPS (Gneiting and Ranjan 2012)

$$\begin{aligned}
 wCRPS(F, y) &= \int_{-\infty}^{\infty} w(x)(F(x) - \mathbf{1}\{x \geq y\})^2 dx \\
 &= \mathbb{E}_F |W(X) - W(y)| - \frac{1}{2} \mathbb{E}_F |W(X) - W(X')| \\
 &= W(y) + 2 \left[ \bar{F}(y) \mathbb{E}_F (W(X) - W(y) | X > y) - \mathbb{E}_F (W(X)F(X)) \right] \\
 &= \mathbb{E}_F |W(X) - W(y)| + \mathbb{E}_F (W(X)) - 2 \mathbb{E}_F (W(X)F(X))
 \end{aligned}$$

where  $W = \int w$  and  $\int wf < \infty$

- ▶ The weight function cannot depend on the observation : it leads to improper scores.

## (Weighted) CRPS embarrassing properties

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$$w_q(x) = \log(x) \mathbf{1}\{x \geq q\}$$

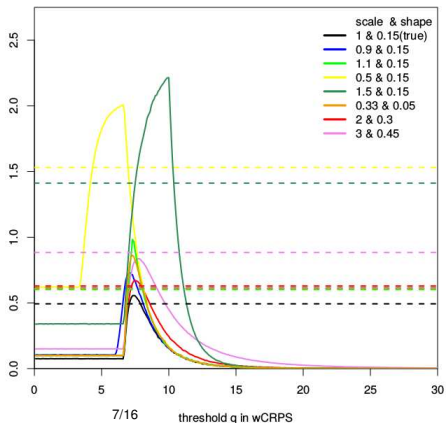
This weight function is closely linked to the Hill's tail-index estimator.

# (Weighted) CRPS embarrassing properties

$$w_q(x) = \log(x) \mathbf{1}\{x \geq q\}$$

This weight function appears suitable for extremes but...

wCRPS (plain) and CRPS(dotted) for GPDs



## (Weighted) CRPS embarrassing properties II

- ▶ Tail equivalence

$$\lim_{x \rightarrow \infty} \frac{\bar{F}(x)}{\bar{G}(x)} = c \in (0, \infty)$$

- ▶ For any given  $\epsilon > 0$ , it is always possible to construct a CDF  $F$  that is **not tail equivalent** to  $G$  and such that

$$|\mathbb{E}_Y(wCRPS(G, Y)) - \mathbb{E}_Y(wCRPS(F, Y))| \leq \epsilon$$

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## Paradigm of verification for extremes ?

*“The paradigm of maximizing the sharpness of the predictive distributions subject to calibration” (Gneiting et al. 2006)*

*“Extreme events are often the result of some extreme atmospheric conditions and combinations : Most of the time just few members in the ensemble leads to such events. We could just look at the information brought by the forecast. But how ?”*

- ▶ **Consequence** : we do not care about reliability here ! (More in “detection” logic)
- ▶ An example : The ROC Curve
- ▶ **Different criterion** : Be skillful for extremes subject to a good overall performance.
- ▶ **Question** : How combining an extreme verification tool with the CRPS ?

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## How using extreme value theory with CRPS ?

The classical Continuous Ranked Probability Score (CRPS) can be written as :

$$CRPS(F, y) = \mathbb{E}|X - y| + \mathbb{E}(X) - 2\mathbb{E}(XF(X))$$

And for large  $y$  it is possible to show that :

$$CRPS(F, y) \approx y - 2\mathbb{E}(XF(X))$$

### Pickands-Balkema-De Haan Theorem (1974-1975)

If the observed value is viewed as a random draw  $Y$  with CDF  $G$ , the survival distribution of  $CRPS(F, Y)$  can be approximated by a GPD with parameters  $\sigma_G$  and  $\xi_G$  :

$$\mathbb{P}(CRPS(F, Y) > t + u \mid CRPS(F, Y) > u) \sim GP_t(\sigma_G, \xi_G)$$



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## And so what ?

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- ▶ Are we trapped ? Parameters are the same whatever the forecast
- ▶ **Crucial (and unrealistic) assumption here** :  $F$  and  $G$  are independent
- ▶ In practice, the convergence to these parameters is driven by the skill of ensembles for extreme events

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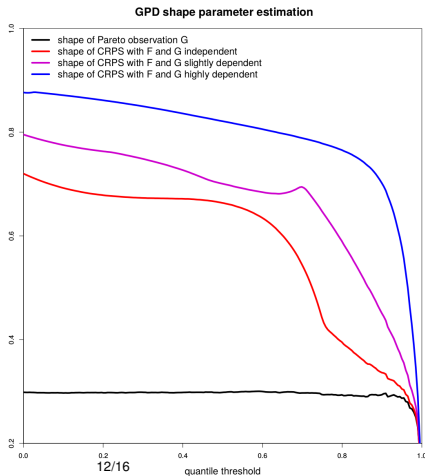
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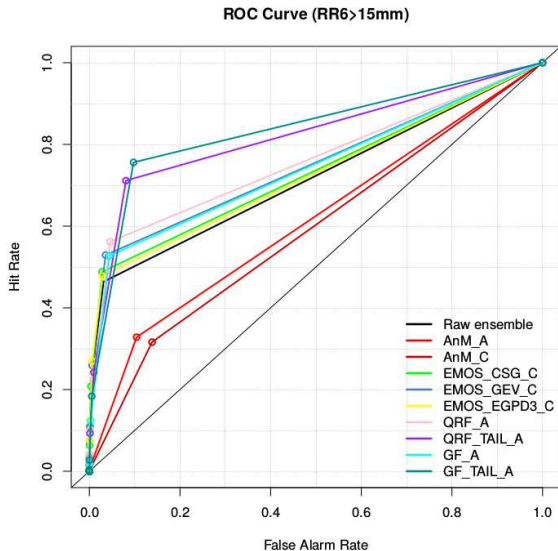


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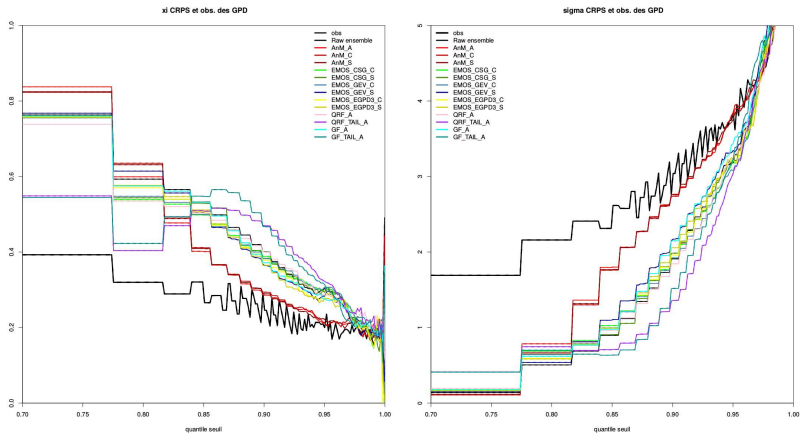
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# Post-processing of 6-h rainfall for extremes



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Estimations of GPD parameters are highly correlated

## Conclusions

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- ▶ Some properties of (w)CRPS are debated... And also used
- ▶ A different criterion for extreme verification is established  
**Be skillful for extremes subject to a good overall performance**
- ▶ A new way to verify ensemble (only ?) forecasts for extremes is shown
  
- ▶ This tool can be viewed as a summary of ROCs among thresholds.
- ▶ It seems to be consistent with simulations and real data



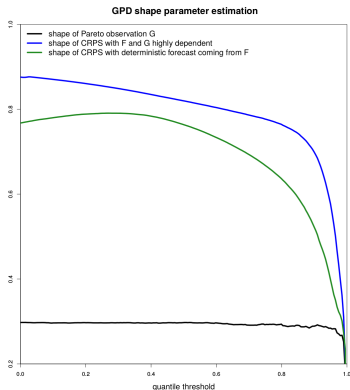
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## Références I

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