



# A STOCHASTIC WEATHER GENERATOR FOR TEMPERATURE

Examples of use and future developments

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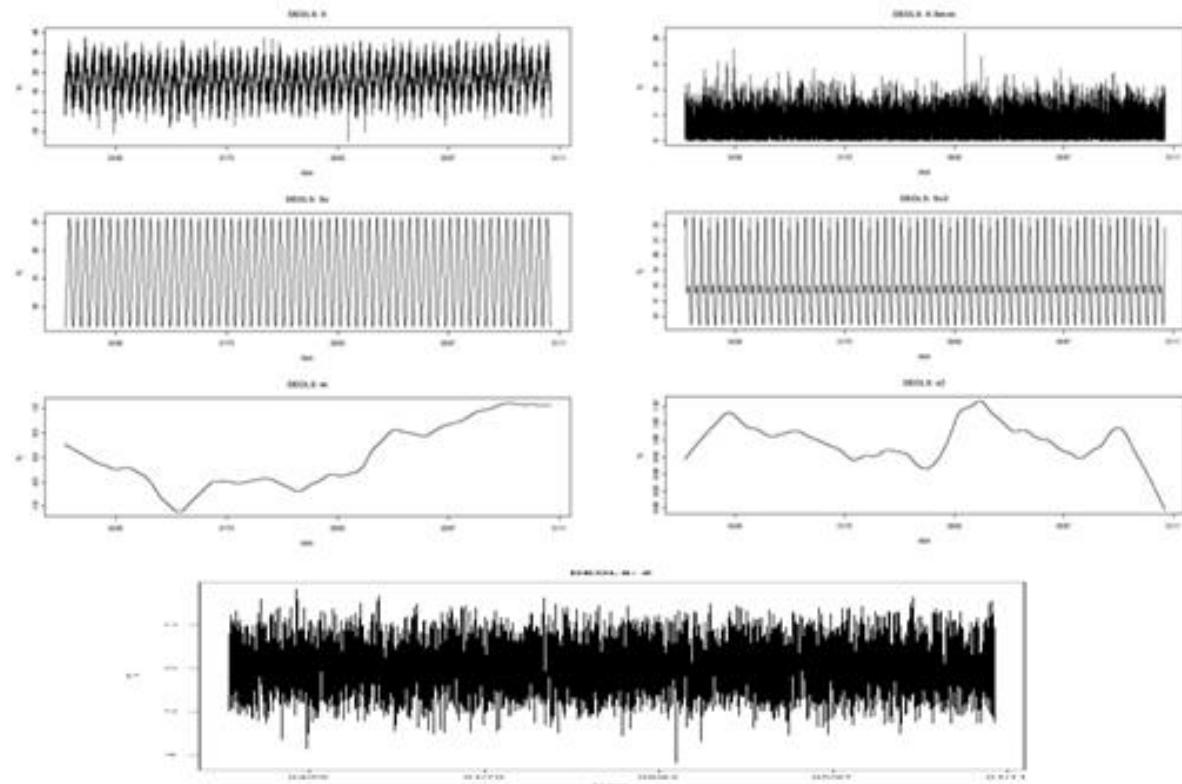


# Context

- Collaboration with the mathematics laboratory of Paris 11 Orsay University
- Temperature extremes and non stationarity
- Methodological developments
- Considerations for the development of a stochastic generator
  - Able to simulate extreme values
  - Allowing a (quick) simulation of a large number of equivalent time series

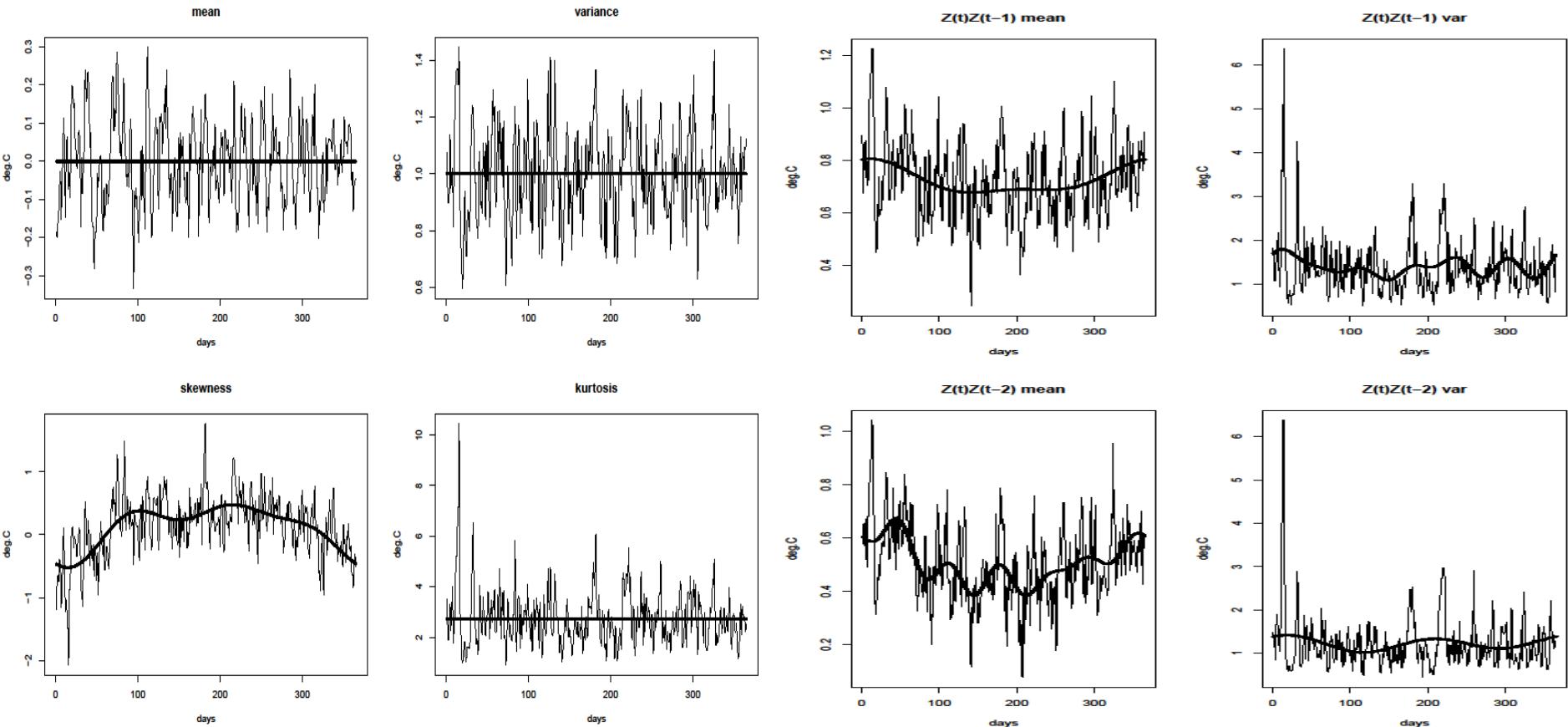
# Model principle

- Simulation of  $Z(t)$



$$Z(t) = \frac{X(t) - S_m(t) - m(t)}{S_v(t)s(t)}$$

# Characteristics of Z



Remaining seasonalities; no more trends

# The SFHAR(seasonal functional heteroscedastic autoregressive) model

➤ Extension: SFHAR model

$$Z(t) = \left[ \theta_{0,k} + \sum_{j=1}^{p_1} \left( \theta_{1,k}^j \cos \frac{2j\pi t}{365} + \theta_{2,k}^j \sin \frac{2j\pi t}{365} \right) \right] Z(t-1) + a(t, Z_{t-1}) \varepsilon_t$$

➤  $a^2(t, Z_{t-1})$

- Seasonal, depend on the state
- Zero out of the boundaries
- positive
- constraints  $C$  on the first derivative from the continuous-time diffusion process (see thesis of Hoang, 2010):

$$(a^2)'(r_1) = \frac{2b(r_1, t)}{1 - 1/\xi_1} \quad \text{et} \quad (a^2)'(r_2) = \frac{2b(r_2, t)}{1 - 1/\xi_2}$$

- Form of  $a$ :  
$$\begin{cases} \hat{a}^2(t, Z_{t-1}) = (\hat{r}_2 - t)(t - \hat{r}_1) \sum_{k=0}^5 \sum_{j=1}^{p_2} \left( \alpha_{1,k}^j \cos \frac{2j\pi t}{365} + \alpha_{2,k}^j \sin \frac{2j\pi t}{365} \right) Z_{t-1}^k \\ C(\hat{r}_1, t), C(\hat{r}_2, t) \\ \hat{a}^2(t) > 0 \quad \forall t \end{cases}$$

# References

- Dacunha-Castelle D., Hoang T.T.H., Parey S.: Modeling of air temperatures: preprocessing and trends, reduced stationary process, extremes, simulation, Journal de la Société Française de Statistique, 2013
- Parey S., Hoang T.T.H., Dacunha-Castelle D. (2014): Validation of a stochastic temperature generator focusing on extremes and an example of use for climate change, Climate Research, Vol 59, 61-75, doi: 10.3354/cr01201

## Examples of use

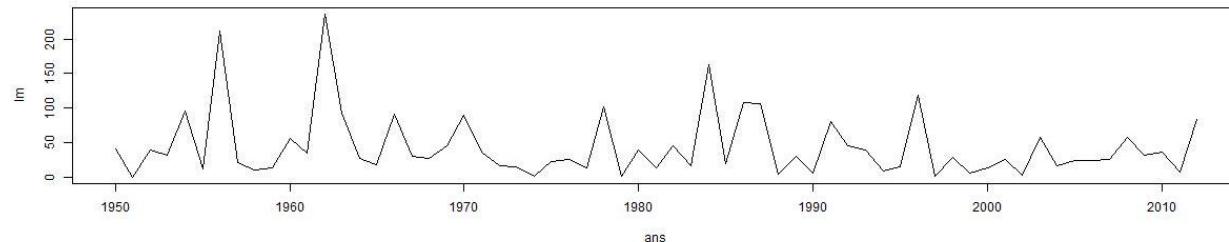
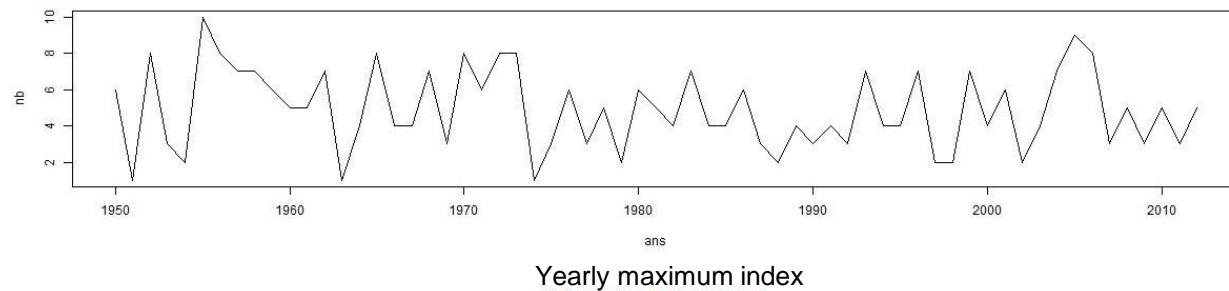
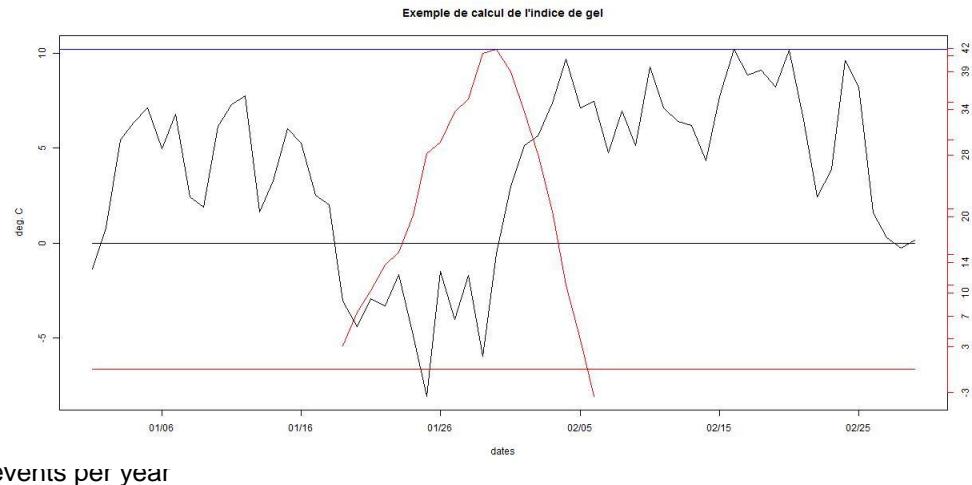
Return levels of frost indices

# Frost index

- Sum of negative daily mean temperatures

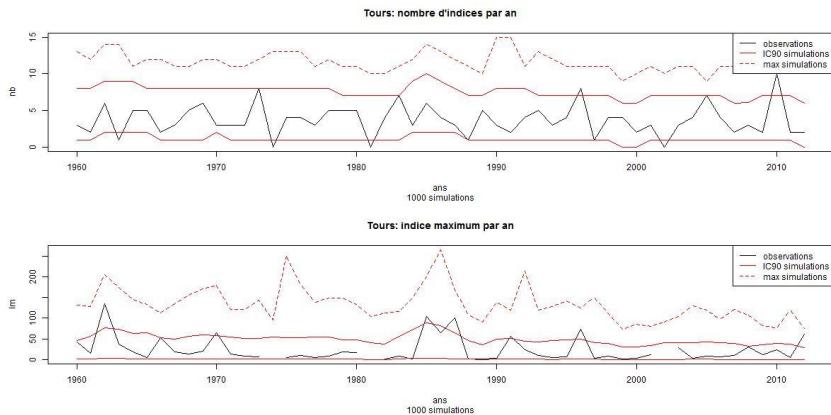
- Computation:

- $t_0$  1<sup>st</sup> time  $T(t_0) < 0^\circ\text{C}$
- $N(t_0) = \min\{t > t_0, T(t_0) + T(t_0+1) + \dots + T(t) > 0\}$
- $Ig(t_0) = \max\{t_0 < k < N(t_0), -(T(t_0) + T(t_0+1) + \dots + T(k))\}$



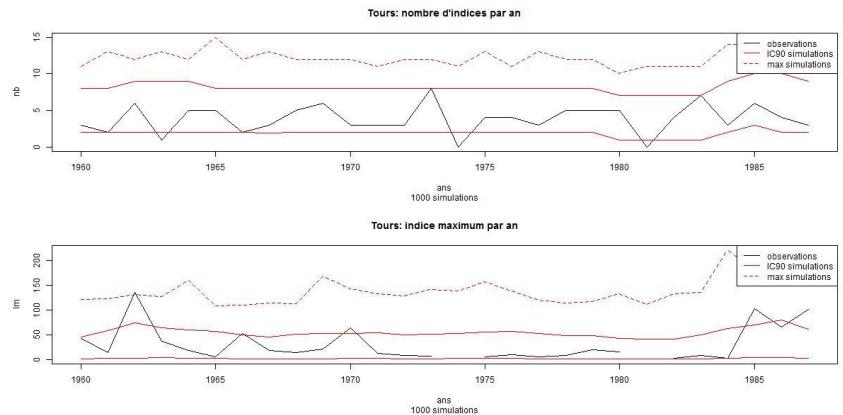
# Use of the generator

Tours 1960-2012:  $0^\circ\text{C} \cong q3.8\%$



7 indices  $> q95$  (5% 53 years  $\Rightarrow 2,65$ )

Tours 1960-1987:  $0^\circ\text{C} \cong q4.4\%$



3 indices  $> q95$  (5% 28 years  $\Rightarrow 1,4$ )

The generator tends to under-estimate the frequency of long events when the threshold lies in the far tail of the T distribution

# Examples of use

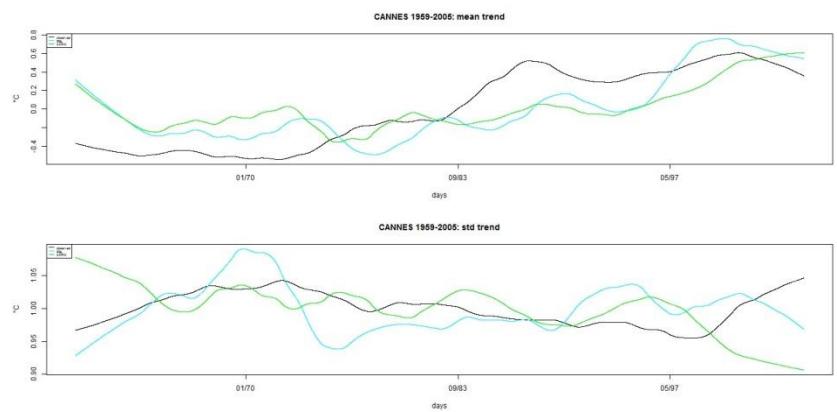
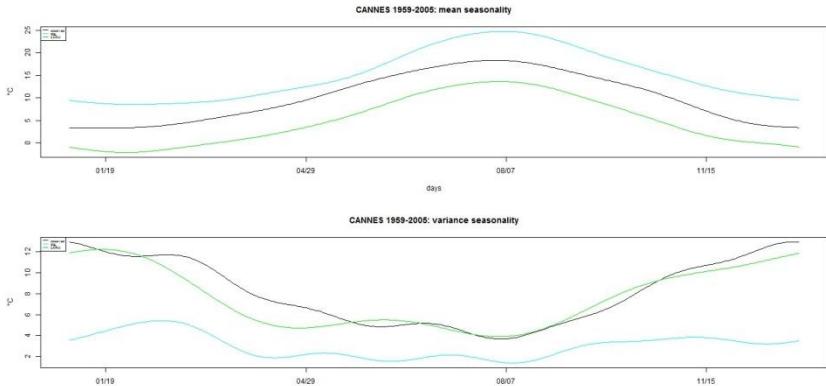
Future cold waves

# Cold waves

- 22 temperature time series in France
- 3 ≠ thresholds: 0°C, q10%, q5%
- Frequency of events of ≠ lengths (from 1 day to > 15 days)
- 2 climate models (IPSL-CM5-MR and CNRM-CM5)
- 2 future periods (2060; 2100); 2 RCPs (4.5, 8.5)

# Use of the stochastic model

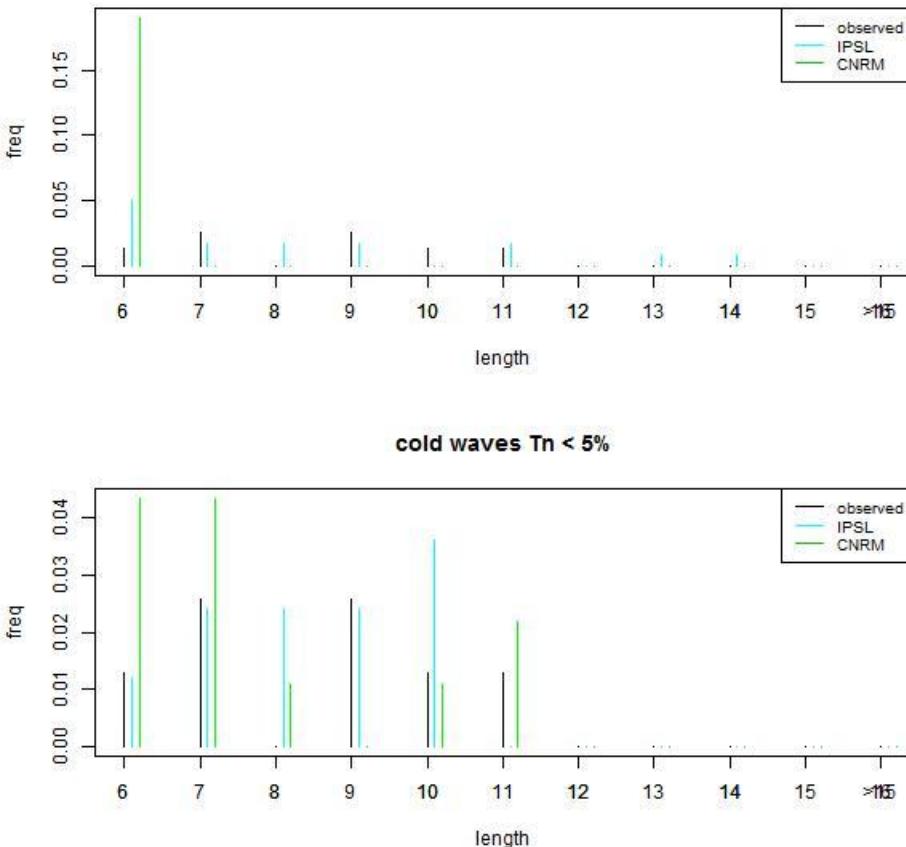
- 100 simulations of  $Z(t)$  based on observations
- Climate model:
  - Present:  $X(t)$  with observed seasonalities and model trends
  - Future:  $\text{Smf} = \text{Smo} + (\text{Smm2} - \text{Smm1})$   
 $\text{Svf} = \text{Svo} * \text{Svm2} / \text{Svm1}$



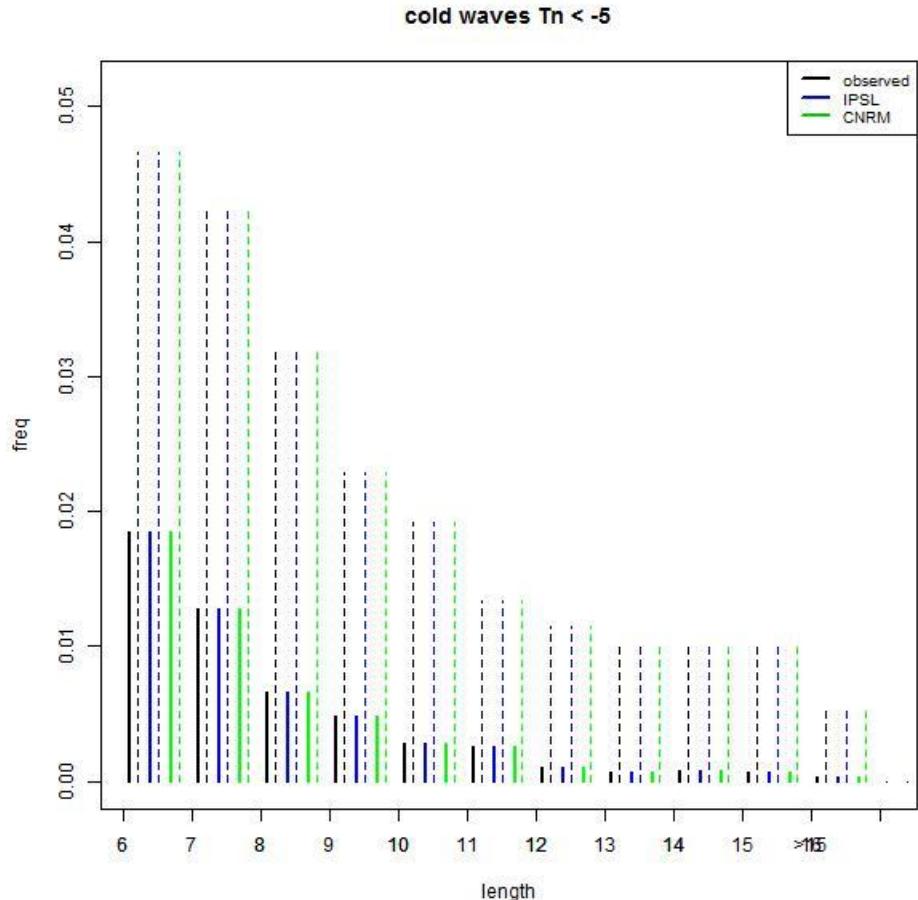
# Generator's advantage

Example: Boulogne sur mer; Threshold = q5%

cold waves  $T_n < -5$



cold waves  $T_n < 5\%$



Ability to produce long events, even for extreme thresholds  
gives confidence intervals

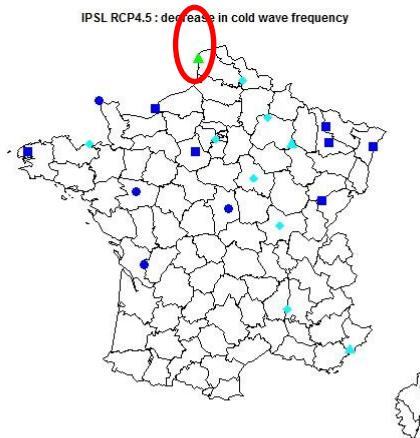
# Significant changes in frequency

Period 2: 2100 RCP 4.5

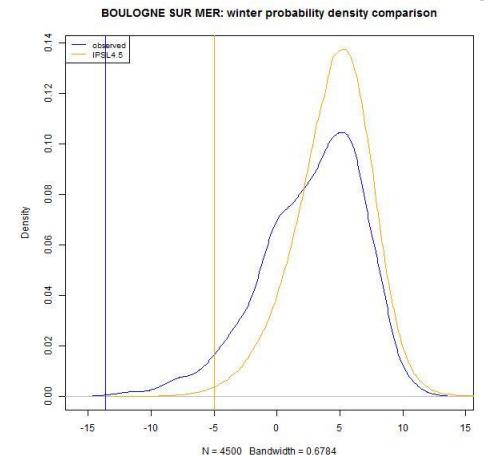
Climate model, threshold present



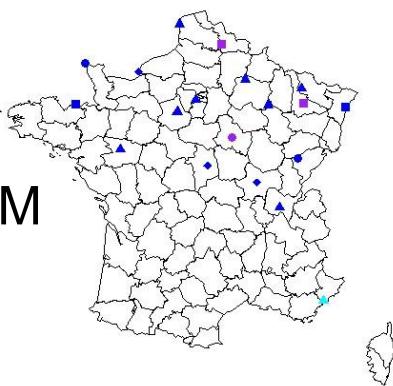
Observations with threshold shift



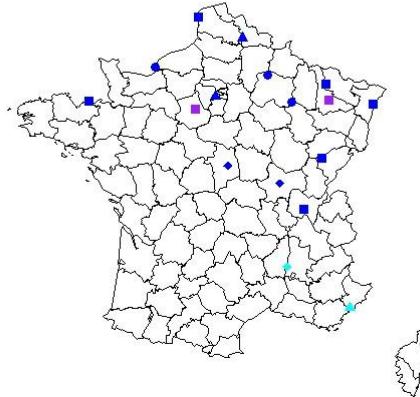
Importance of variance change



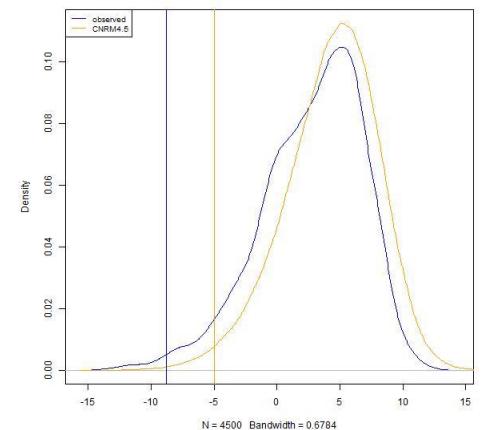
CNRM RCP4.5 : decrease in cold wave frequency



CNRM RCP4.5 : decrease in cold wave frequency



BOULOGNE SUR MER: winter probability density comparison



- 1d ● 2d ▲ 3d ● 4d ■ 5d ● 6d ▲ 7d ● 8d ■ 9d ■ 10d ▲ 11d ● 12d ■ 13d ● 14d ▲ 15d ■ >15d

# Conclusion and perspectives

- Generator => reliable representation of distribution of temperature time-series, even the tails and heat or cold wave frequencies
- Weakness: use of EVT to estimate the shape parameters and the bounds
- Identified limitation: long events when the threshold lies in the far tail
- Works in progress:
  - Improvement of estimation of the shape parameter  $\xi$  with different non parametric estimators (Hill and his extensions)
  - Improvement of the quality of extremes by carefully analyzing correlation over high (and below low) percentiles
  - Direct estimation of the bounds not using the estimation of extreme parameters



# THANKS!

