# Étude de saisons caniculaires extrêmes en Asie du Sud à l'aide d'un algorithme d'événement rares

Clément Le Priol<sup>1,2</sup>, with Joy M. Monteiro<sup>3,4</sup>, Freddy Bouchet<sup>2</sup>

Journées du GDR « Défis théoriques pour les Sciences du Climat », Grenoble, 27 mai 2024

- <sup>1</sup> Laboratoire de Physique à l'ENS de Lyon, Lyon, France
- <sup>2</sup> Laboratoire de Météorologie Dynamique, IPSL, ENS-PSL, CNRS, Paris, France
- <sup>3</sup> Department of Earth and Climate Science, IISER Pune, Pune, India
- <sup>4</sup> Department of Data Science, IISER Pune, Pune, India





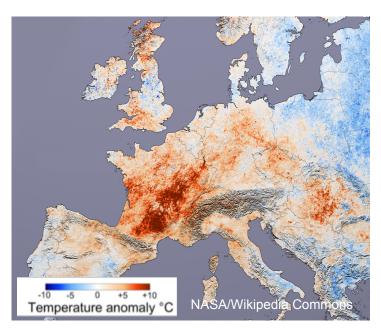






#### **Extremely rare events matter**

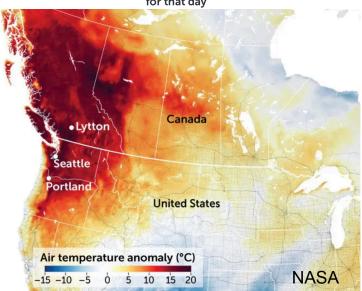
#### 2003 West European heatwave



July 20 - August 20 temperature anomaly

#### **2021 Pacific Northwest heatwave**

June 29, 2021 temperature compared with the 2014–2020 average for that day



NASA

#### **Extremely rare events matter**

# An event that has a **return time** of **1000 years** has:

• 1/1000 chance to occur in any year

$$\operatorname{rt}(x) = 1/\mathbb{P}(X \ge x)$$

•6% chance to occur in the next 60 years.

(In a stationary climate)

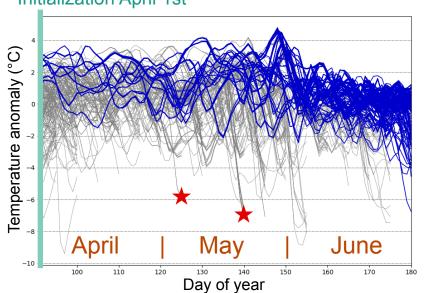
#### Difficulties in studying extremely rare events

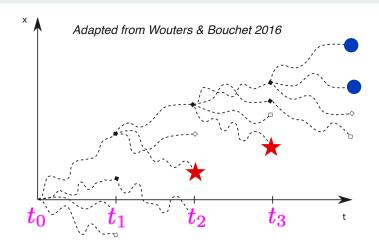
- Observational records (60-150 years): Too short for observing most events but useful for GEV/GPD fits
- Climate models: Obtaining good statistics on events that
  have a return time of centuries using direct simulations
  requires millennia of simulation -> extremely large
  computation cost with the best models
- How to sample extremely rare events in climate models?
  - —> Rare event algorithms

#### Rare event algorithm's principle

**Duplicate** the trajectories most likely to produce the desired event, **eliminate** the others (*resampling step*).







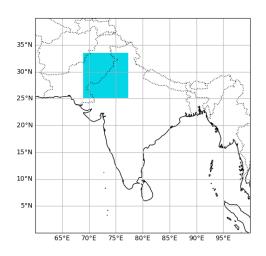
N=200 trajectories (constant)

Resample every 5 days

Total integration time: 90 days (3 months)

#### Application: extreme heatwave seasons in South Asia

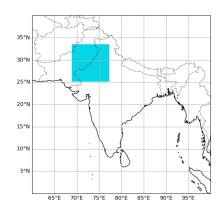
- Heatwave season = pre-monsoon season (March to June)
- We want to sample extrema of April-May-June (AMJ) averaged temperature



#### Choice of the score function

- ullet We resample according to weights  $W_n$  assigned to each trajectory
- The computation of the weights relies on a **score function**  $W\left(\left\{x(t)\right\}\right)$
- A good score function, tailored to the events of interest, is crucial to the effectiveness of the algorithm.

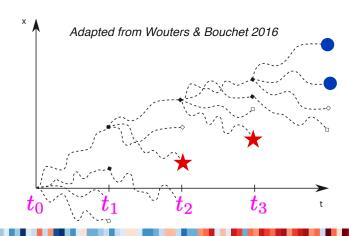
We want to sample extremes of  $\tilde{A}\left(\{x(t)\}\right)=rac{1}{T}\int_{\mathrm{AMJ}}A(x(t))dt\,,\;T=90\,\mathrm{days}$ 



where 
$$A(x(t)) = \frac{1}{\mathcal{A}} \int_{Area} T_{2m}(\mathbf{r}, t) d\mathbf{r}$$

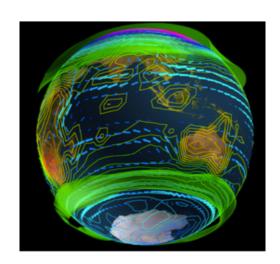
#### Score function suited for long-lasting extremes:

$$W\left(\left\{x(t)\right\}_{t_{i} \leq t \leq t_{i+1}}\right) \propto \exp\left(k \int_{t_{i}}^{t_{i+1}} A\left(x(t)\right) dt\right)$$



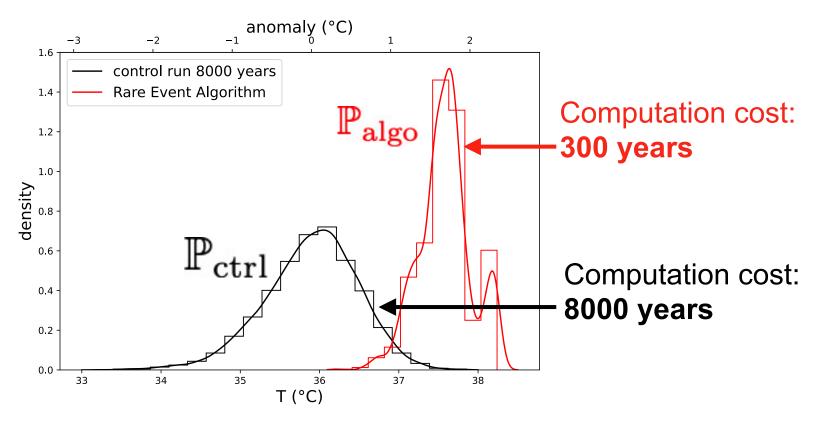
#### Model used: PlaSim

- Land-atmosphere coupling configuration.
- Runs fast (less than 1hCPU/year)
- A control run of 1200 years (1990's climate) provides initial conditions for the algorithm
- Independent 8000-year long control run
- **Goal:** Compare the algorithm extreme event statistics with the long control run statistics.

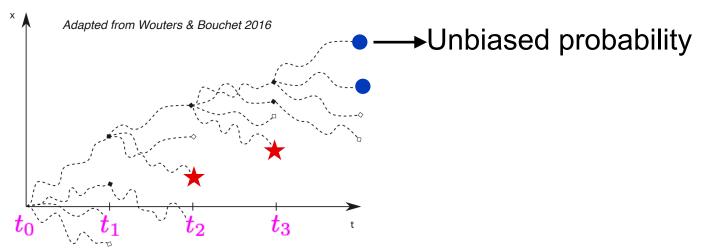


#### Outcome: Sampling of extremely rare heatwave seasons

### PDF of AMJ averaged temperature



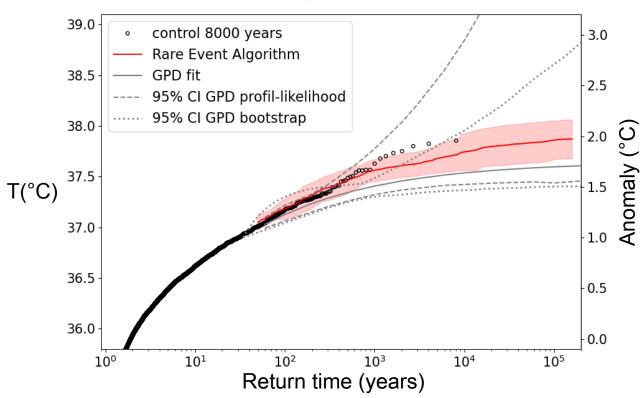
#### We can recover the unbiased probability of each trajectory.



The knowledge of the trajectories probabilities is crucial to compute return time curves and any other statistics

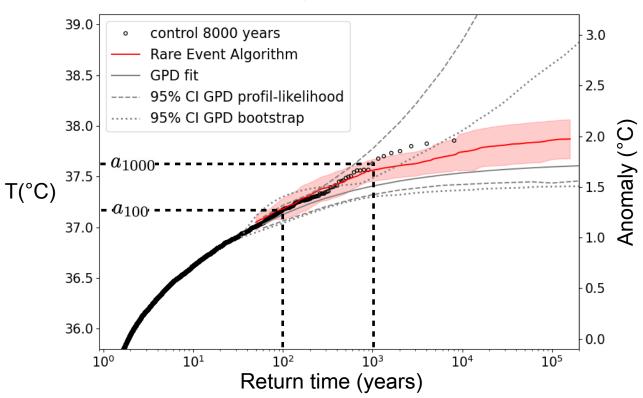
#### **Outcome: Return time curve**

#### **AMJ** averaged temperature



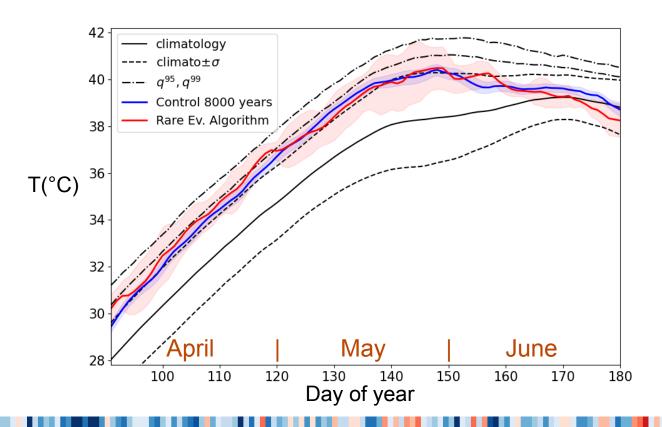
#### **Outcome: Return time curve**

#### **AMJ** averaged temperature

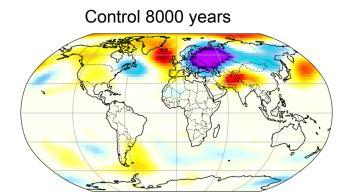


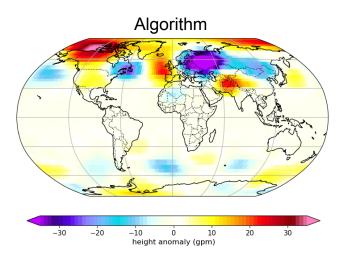
#### **Evolution of the temperature during centennial heatwave seasons**

#### **Centennial heatwave season**



# Composite maps of Zg500 AMJ anomaly

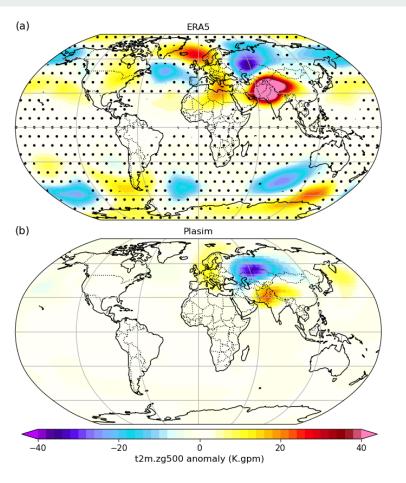




# 1000-year heatwave

season:  $T_{\rm AMJ} \geq a_{1000}$ 

# Can we trust the teleconnection pattern of the model?



T2m over the study region and Zg500.

MAMJ average

#### **Perspective**

Cost of running the algorithm: 300 years

but

we needed initial conditions: 1200-year long run

#### **Solution:**

Draw initial conditions from existing Single Model Large Ensembles

#### Key messages

- Extremely rare events matter
- Rare event algorithms can sample a large number of very extreme events
- The knowledge of the trajectories probabilities is crucial to compute any statistics.
- They provide a **precise estimate** of the (model-dependent) **return time curve**.
- Rare event simulations could be combined with Single Model Large
   Ensembles to explore extremely rare events in future warmer worlds

Our work is on arXiv: http://arxiv.org/abs/2404.07791

# **Appendix**

## Algorithm experiment and computational cost

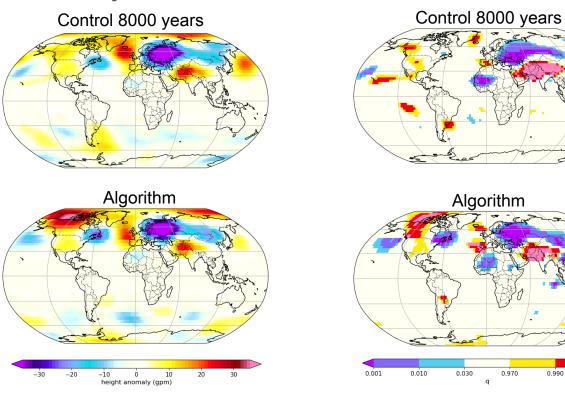
1200 independent initial conditions (April 1st)

6 algorithm experiments with N=200 trajectories running from April to June

Total computational cost: 6\*200\*3 months = 300 years

## Composite maps of Zg500 AMJ anomaly





Statistical significance (T-test)

0.990