

Extreme events and Climate Change: a statistical perspective

Aurélien Ribes

27 June 2022, Kelowna (and Toulouse)

Motivation / This talk

- ◇ Most impacts of climate change are expected to occur through extreme events.
e.g., sea level rise,
- ◇ How extreme events are / will be affected by climate change?
What events to expect in the future?
- ◇ Statistics play a central role.

This talk:

- ◇ A review of recent research and statistical challenges in investigating extreme events and climate change.
- ◇ I assume the audience is familiar with stats (e.g., GEV), but not familiar with climate data.

- 1 Extreme events in climate
- 2 Data
- 3 Questions to investigate
- 4 Some statistical challenges and personal work
- 5 Conclusion

Which events and the current knowledge

- ◇ **Temperature** (easy): warm ↗, cold ↘
- ◇ **Heavy precipitation** (quite easy): ↗ (most regions)
- ◇ **Droughts** (quite difficult): depends on def, region, period, length, ...

Projected changes in extremes are larger in frequency and intensity with every additional increment of global warming

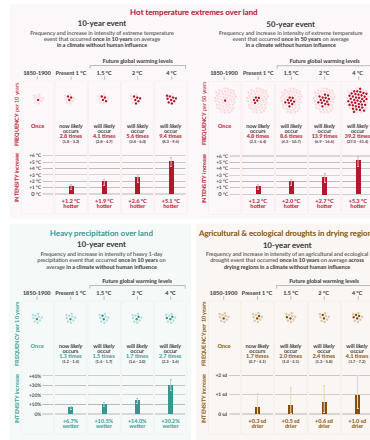


Figure SPM.6, IPCC WGI AR6.

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- ◇ **Droughts** (quite difficult): depends on def, region, period, length, ...
- ◇ **Floods**
- ◇ **Coastal floods / waves**
- ◇ **Storms**: tropical cyclones, extratropical storms
- ◇ **Convective storms**: thunderstorms, convective gusts, lightning, hail, etc
- ◇ **Compound events**: hot and dry, frost on active vegetation

Projected changes in extremes are larger in frequency and intensity with every additional increment of global warming

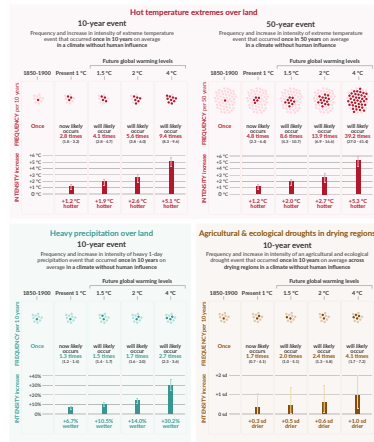


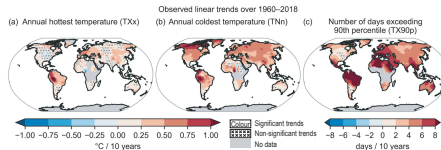
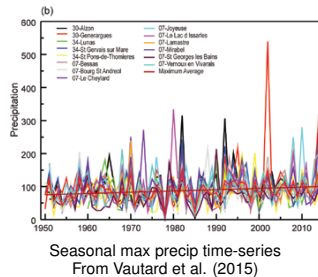
Figure SPM.6, IPCC WGI AR6.

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Data

◇ Observations

Big bunch of literature to describe observed changes in extreme events

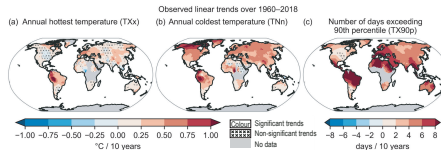
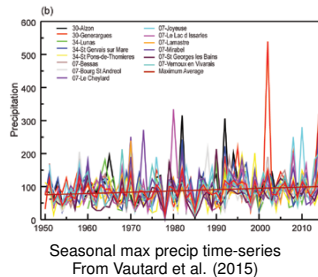


Data

◇ Observations

Big bunch of literature to describe observed changes in extreme events

◇ Models



Climate models: principle

Models are based on:

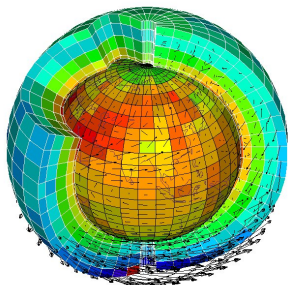
- physical equations (Navier-Stokes, thermodynamics, radiation, etc).
- numerical resolution of PDEs.

Resolution:

- spatial (H) ~ 100 km,
- spatial (V) ~ 500 m,
- temporal $\sim 15'$.

Models simulate the climate system:

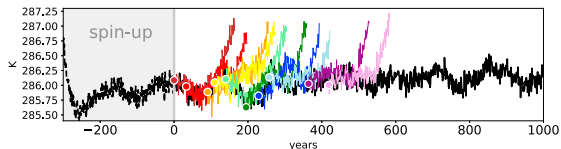
- atmosphere, ocean, land surfaces, biosphere, cryosphere, rivers, ...
- IV and forcings.



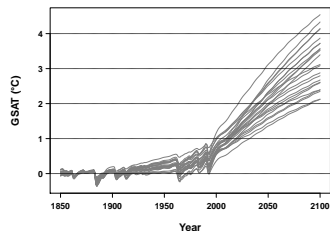
Source : IPSL

Climate models: practical use

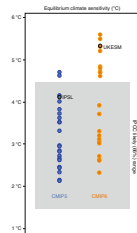
- ◇ Model development and use is a huge effort!!
- ◇ Numerical experiments: pi-control, historical, scenarios, theoretical experiments, ...
- ◇ Coordinated simulations: CMIP ensemble (~40 models in CMIP6).
- ◇ Several types of models/resolution: global (~100km), regional (~10km), km-scale (~1km), ...



Model spin-up, pre-industrial control (black), and historical (colors) simulations with the model CNRM-CM6. From Voldoire et al., 2019, JAMES.



Left: Response of CMIP6 models to SSP2-4.5 scenario



Right: Warming response to $2 \times \text{CO}_2$ in CMIP5/6 models

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Questions

How extreme events are / will be affected by climate change?

In IPCC AR6 WG1 Chapter 11:

- ◇ Mechanisms and Drivers
- ◇ Observed trends
- ◇ Model evaluation
- ◇ Detection and Attribution, Event attribution
- ◇ Projections

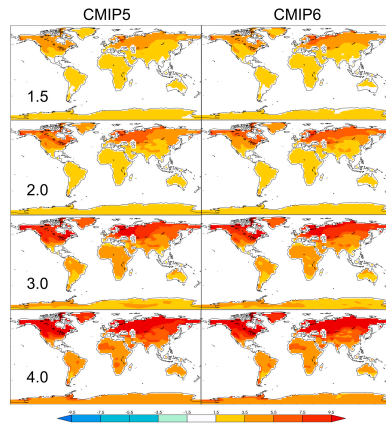
Projections

Describe how extreme events respond in (model) projections.

Examples:

Temp: Wehner et al. (2020, WACE);

Precip: Kharin et al. (2012, below); John et al. (2022).



Change in TNn 20-yr RV, at various Global Warming Levels (GWLs); from Wehner et al., 2020, WACE.

Projections

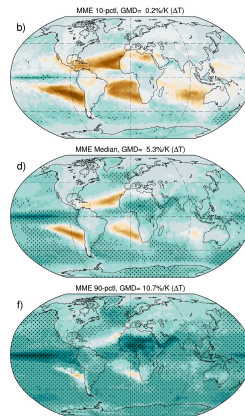
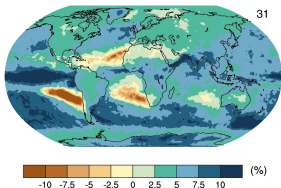
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Daily precipitation 20-yr RV change per 1°C warming



Confidence interval on daily precip 20-yr RV change; from John et al. (2022).

Detection / attribution

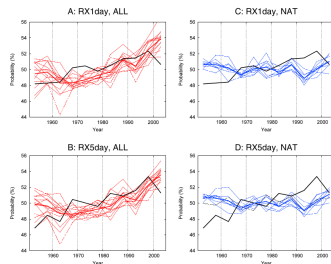
- ◇ **Detection:** are observations consistent with internal variability only?
- ◇ **Attribution:** assess the relative contribution of (various) external forcings.
e.g., ANT / NAT, GHG / AER, etc.

Example: Min et al (2011), Zhang et al (2013) looked at annual max 1d rainfall (RX1D) worldwide and report a statistically discernible human influence.

LETTER

doi:10.1038/nature09763

Human contribution to more-intense precipitation extremes

Seung-Ki Min¹, Xuebin Zhang¹, Francis W. Zwiers^{1†} & Gabriele C. Heger²

From Zhang et al. (2013)

Event attribution

- ◇ Concept: consider a specific event that did happen
- ◇ Investigate / Describe human contribution to that event
Assess human influence on probability / intensity of the event?
What to expect in the future?
- ◇ Historical example: European 2003 HeatWave
Stott et al., 2004, Nature
- ◇ Now: *rapid* attribution (eg, WWA).
- ◇ Potential interest: awareness-raising, legal aspects.

Human contribution to the European heatwave of 2003

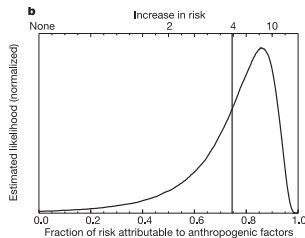
Peter A. Stott¹, D. A. Stone^{2,3} & M. R. Allen²

¹Met Office, Hadley Centre for Climate Prediction and Research (Reading Unit), Meteorology Building, University of Reading, Reading RG6 6BB, UK

²Department of Physics, University of Oxford, Oxford OX1 3PU, UK

³Department of Zoology, University of Oxford, Oxford OX1 3PS, UK

The summer of 2003 was probably the hottest in Europe since at latest AD 1500¹⁻⁴, and unusually large numbers of heat-related deaths were reported in France, Germany and Italy⁵. It is an ill-posed question whether the 2003 heatwave was caused, in a simple



Causality

◇ The 2010 Russian heatwave (R10) example:

Dole et al, 2011: R10 is “mainly natural in cause”,

Rahmstorf and Coumou, 2010: R10 “*would not have occurred*” without climate change.

Reconciling (Otto et al, 2012): most of the heat anomaly is natural; CC increased the risk significantly.

◇ Relationship between “(event) Attribution” and “Causal theory”,

e.g., Hannart et al., 2016, BAMS; Hannart et al., 2018, JCLim

$$PN = \max\left(1 - \frac{p_1}{p_0}, 0\right),$$

$$PS = \max\left(1 - \frac{1-p_1}{1-p_0}, 0\right),$$

$$PNS = \max(p_1 - p_0, 0).$$

◇ Usually, “*human influence*” is a necessary, but not sufficient cause.

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Related to the extremes events

- ◇ Statistical model selection, e.g., non-stationary models

$$y_t \sim GEV(\mu_0 + \mu_1 x_t, \sigma_0 + \sigma_1 x_t, \xi).$$

- ◇ Select appropriate covariate x_t ,
- ◇ Constant ξ ?
- ◇ GEV models may not fit annual max.
Ben Alaya et al., 2020 (JCLim), 2021 (WACE).

- ◇ Specific investigation of *Low Likelihood High Impact* scenarios / events, sse, e.g., Sutton, 2019, BAMS.

- ◇ If no data: statistical investigation of environmental conditions (incl. AI).
e.g., Tropical cyclogenesis, Menkes et al., 2009, Clim Dyn; CAPE, Singh et al., 2017, PNAS.

Related to the data

- ◇ Observations
 - ◇ measurement uncertainty,
 - ◇ homogeneity,
 - ◇ missing data,
 - ◇ (spatial) representativity,
 - ◇ short records,
 - ◇ ...

Related to the data

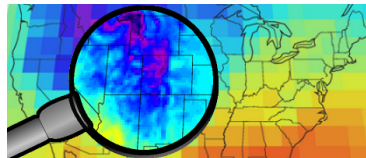
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◇ Models

- ◇ Systematic bias – need for bias correction (= stats),
- ◇ Too coarse resolution – statistical downscaling,
- ◇ Models do not necessarily agree,
- ◇ Combining models: model uncertainty, non-independence, poor design.

Knutti et al., 2010, Jclim; Knutti et al., 2013, GRL



Combining observations and models (1)

- ◇ Obs. now provide indication about warming strength
- ◇ Combination of model and obs to assess past and future changes
- ◇ GSAT warming in IPCC AR6...
Tokarska et al., 2020, Sci Adv; Liang et al., 2020, GRL; Ribes et al., 2021, Sci Adv.

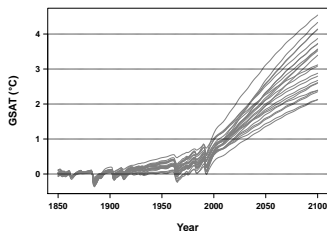


Illustration from Ribes et al. (2021)

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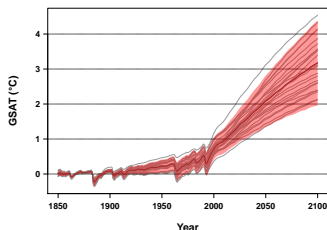


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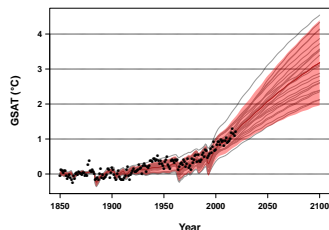


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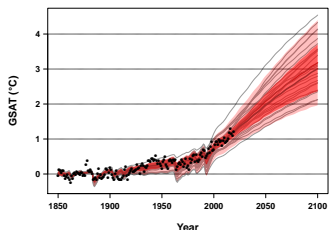


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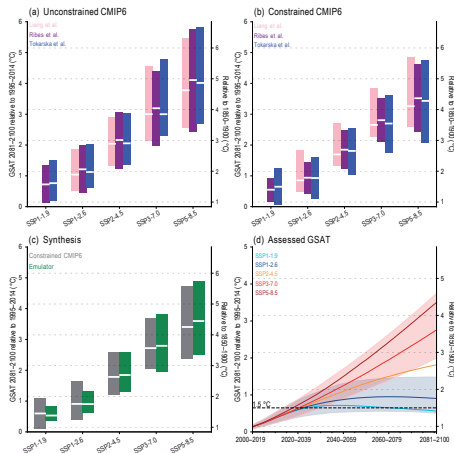


Fig 4.11 from IPCC AR6 (2021)

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Tokarska et al., 2020, Sci Adv; Liang et al., 2020, GRL; Ribes et al., 2021, Sci Adv.
- ◇ ... now moving towards extremes.
Thackeray et al., 2022, NCC

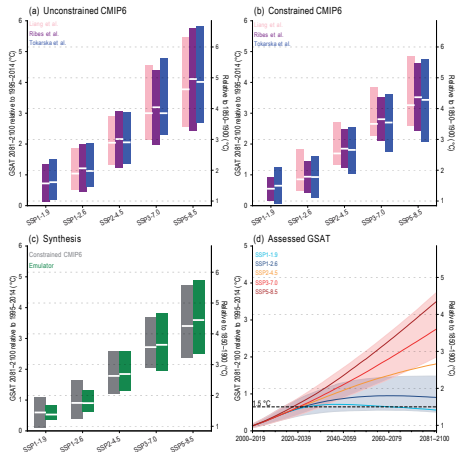


Fig 4.11 from IPCC AR6 (2021)

Combining observations and models (1)

- ◇ “*Best available information*” requires blending models + obs
- ◇ There might be interest in doing so.

Illustration:

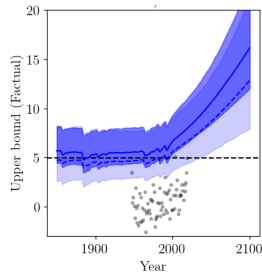
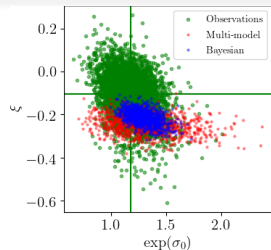
$$y_t \sim GEV(\mu_0 + \mu_1 x_t, \sigma_0 + \sigma_1 x_t, \xi).$$

- ◇ Parameters: $\theta = (\mu_0, \mu_1, \sigma_0, \sigma_1, \xi)$,
- ◇ ξ hard to estimate from obs (short record),
- ◇ μ_0, σ_0 often biased in models.

We use a Bayesian treatment:

- ◇ Models provide a prior $\pi(\theta) \sim N(\mu, \Sigma_{mod})$.
- ◇ We derive the posterior $p(\theta|y)$.

Ref: Robin & Ribes (2020), *ASCMO*



Conclusion

- ◇ A wide range of events – some are very poorly known.
- ◇ One central question: evaluating / quantifying the future hazard.
- ◇ Causal theory involved in relating extreme events to the human influence on climate, with potential societal implications.
- ◇ A number of statistical challenges that need continued statistical research, e.g., smart blending of model + observation data.