A spatial model for warm extremes in the High Arctic

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A SPATIAL MODEL FOR EXTREME EVENTS

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AN APPLICATION TO WARM EXTREMES

- A conventional EVT approach provides unphysical estimates.
- The spatial model improves over conventional EVT.

THE PHYSICAL BASIS OF WARM EXTREMES IN THE HIGH ARCTIC



Based on Messori et al., J. Clim. (2018)

Warm spells: based on domain-averaged T_{2m} anomalies above 80° N in ERA-Interim.



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Select extreme events (~50) over 1979-2016.



Warm/cold spell frequency over NDJFM

Removing the trend: moving average climatology.



Warm/cold spell occurrences









Moisture intrusions: intense, persistent and zonally extended moisture flux across 70 °N.

Basically mid-latitude airmass intrusion extremes.



Pithan et al., Nature Geosci. (2018)

Moisture intrusions: clear link with warm spells:



Mean intrusion duration and number relative to peak temperature anomalies

Moisture intrusions: clear link with warm spells:



Intrusion trajectories lags -8 to -2 days relative to peak temperature anomaly





Cyclone Climatology:



Hanley and Caballero, GRL (2012)

Cyclone Climatology: Cyclones Genesis Lysis (d) (e) Warm (h) (g) Cold -90 -60 -300 30 60 90 -12 -9 -6 -3 0 3 6 9 12 -12-9 -6-3 0 3 6 9 12 Cyclone freq. $\{2 \times 10^6 \text{ km}^2 \text{ NDJFM}\}^{-1}$ Lysis freq. $\{2 \times 10^6 \text{ km}^2 \text{ NDJFM}\}^{-1}$ Genesis freq. $\{2 \times 10^6 \text{ km}^2 \text{ NDJFM}\}^{-1}$

Frequency anomaly of cyclones, genesis and lysis for 4 days prior to warm and cold spells.

Cyclone Climatology:



Frequency anomaly of cyclones prior to warm spells.

THE PHYSICAL BASIS OF WARM EXTREMES IN THE HIGH ARCTIC

- Mid-lat airmasses/moisture intrusions with clear spatial footprint → warm Arctic extremes.
- Large-scale configurations favour the intrusions/ extremes, chiefly in the Atlantic sector.
- Important role of cyclones on synoptic scales.



Based on Wada et al., Ocean Eng. (2018)

The problem:

Estimate severity over long return times from short observational data series, for geophysical extremes with a narrow spatial footprint.

A good example:

Significant wave height from tropical cyclones.

A naïve attempt with classical EVT:



¥34 LON Max. significant Wave Height (m) for an

Todai hindcast dataset

significant Wave Height (m) for individual cyclone

A naïve attempt with classical EVT:



Max. significant Wave Height (m) in Todai hindcast dataset



100-year return value estimates for max. significant Wave Height (m)

STM-E Model:

Two-part model: Space-Time Maxima (STM) and Exposure (E)

STM: extract the locations of the maximum values during each extreme event.

E: a value normalized in [0, 1] for each event, in the form of a geographical map.

Assuming that the distributions of STM and E are independent, the two may be multiplied to derive the STM-E extreme behavior estimate at each location.

Define STMs:

Select maximum for each physical event (e.g. cyclone)

 $\{s\}_{i=1}^{n}$, n being the number of events

Can then use standard EVT (e.g. GPD to compute conditional distribution of threshold exceedances).

Define Es (N.B. sloppy notation for conciseness):

 $e_{i,j} = \max\left(\frac{h(j,t)}{s_i}\right)$, *j* being an index over spatial locations, and t being an index over timesteps for which the event lasts.

Combine to obtain STM-E:

- Event severity at a location *j* is $H_j = E \times S$, *E* and *S* random variables for STM and Exposure.
- Can then obtain the cumulative distribution of H_i .

N.B. We have thus «pooled» STMs so that all physical events contribute to the set $\{s_i\}$.

- Some geophysical extremes pose a challenge to conventional EVT estimates.
- STM-E approach presents a possible solution, taking into account spatial exposure and "pooling" space-time maxima.
- Subject to some assumption on independence of distributions.



AN APPLICATION TO WARM EXTREMES IN THE HIGH ARCTIC



Based on Messori et al., Q. J. Roy. Met. Soc., in review.

Focus on warm extremes driven by mid-latitude airmass intrusions:



Process:

- (i) Extract 2-metre air temperature anomalies for each intrusion. N.B. Can last for multiple days.
- (ii) Compute STM and retain events with STM > 22.5K (we tested sensitivity).
- (iii) Set a threshold of 8 K (we tested sensitivity).
- (iv) Identify contiguous geographical exceedance regions matching the intrusions and compute E.
- (v) Apply STM-E!

Example for one event:





Exposure:











AN APPLICATION TO WARM EXTREMES IN THE HIGH ARCTIC

- Arctic warm extremes are an example of geophysical extreme where a naïve EVT approach fails.
- The STM-E model seems to provide sensible estimates.
- Possible in the future to include non-stationarity in the framework?

Thank You!

Messori, G.*, Ryota, W.*, (equal contribution) Woods, C. A spatial model for warm temperature extremes in the High Arctic. *Q. J. Roy. Met. Soc., in review.*

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