



**Barcelona  
Supercomputing  
Center**  
*Centro Nacional de Supercomputación*

# Seasonal prediction of extreme events

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Climate Forecasting Unit





## Outline:

- Why focusing on extreme events?
- “Extremeness” metric
- Soil influence on extreme temperature
- Heatwave prediction

# What are extreme events?



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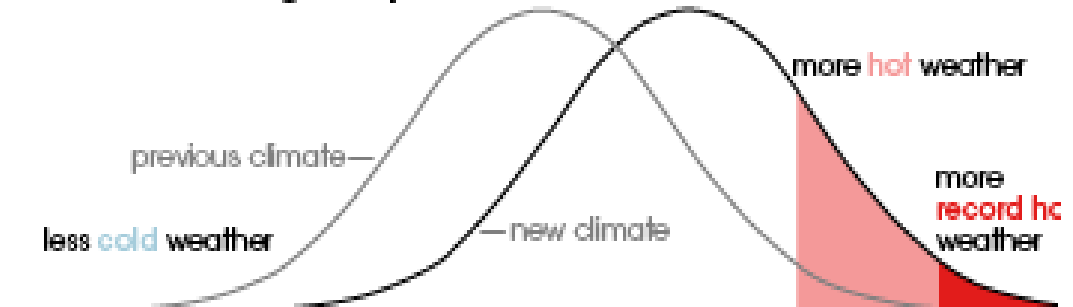
“There are a number of ways extreme climate events can be defined, such as extreme daily temperatures, extreme daily rainfall amounts, large areas experiencing unusually warm monthly temperatures, or even storm events such as hurricanes. Extreme events can also be defined by the impact an event has on society. That may involve excessive loss of life, excessive economic or monetary losses or both.” (Easterling et al. 2000)

# What are extreme events?

→ Global warming will shift and modify the distribution of temperature implying that more extremes will occur.

→ Seasonal prediction of extremes is thus indispensable for adaptation.

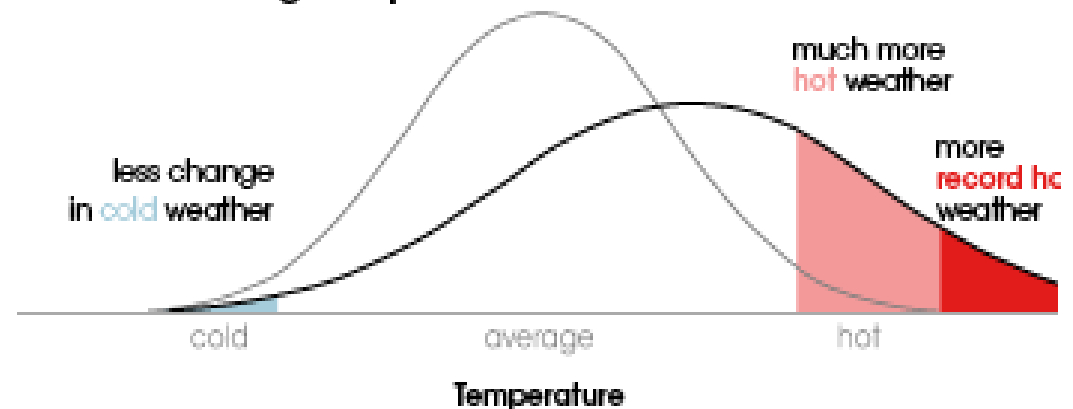
### Increase in Average Temperature



### Increase in Temperature Variance



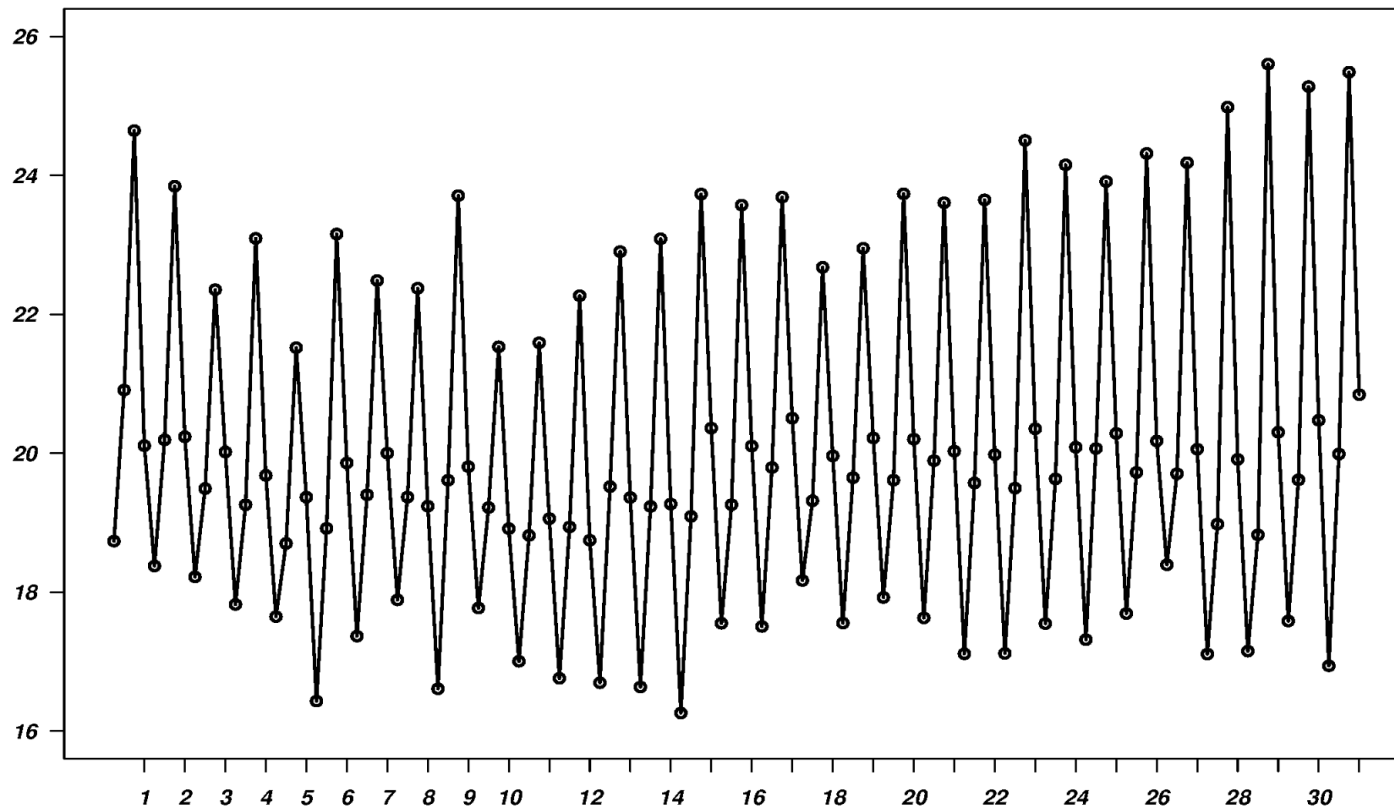
### Increase in Average Temperature and Variance



# “Extremeness” metric



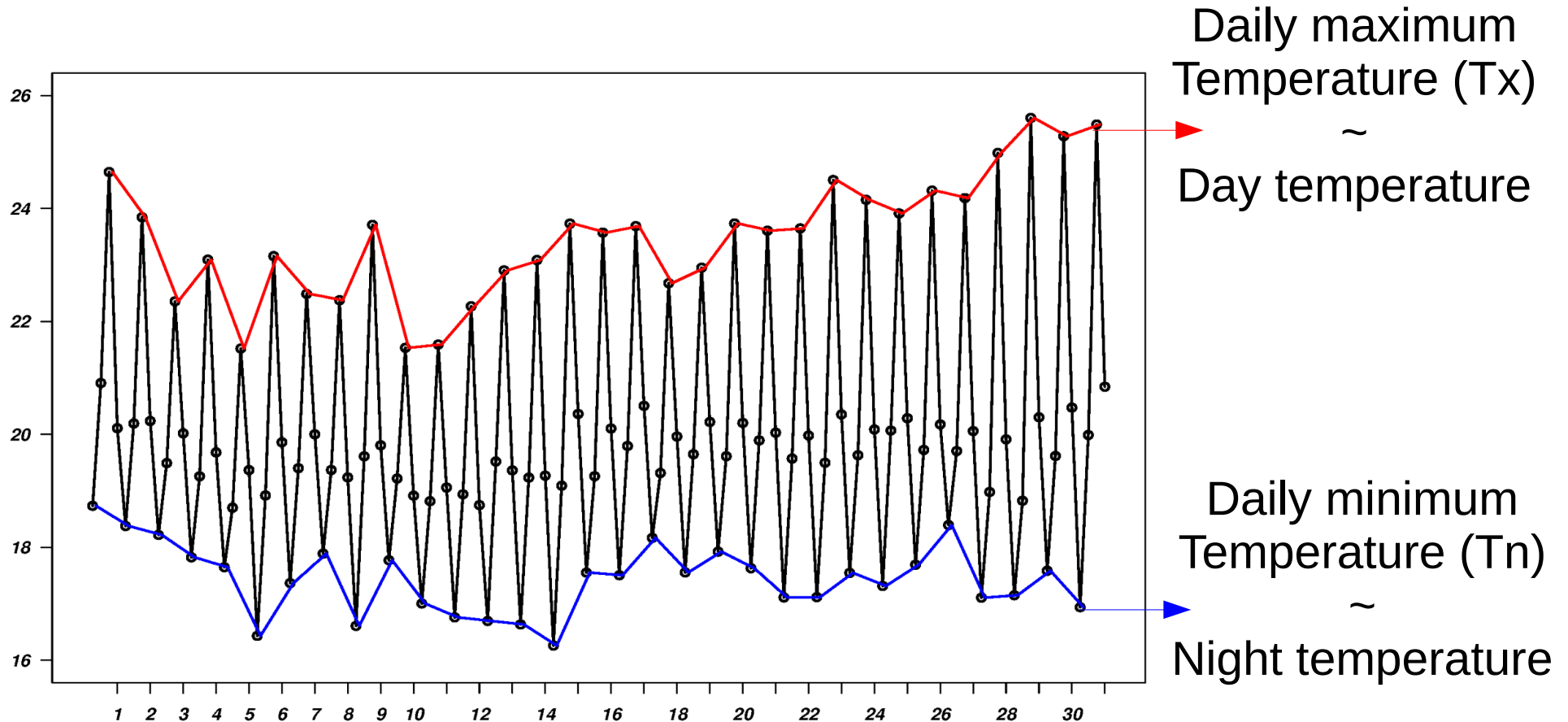
Six-hourly 2m-temperature (°C) in May 1985 from ERA-Interim in one grid point in Europe



# “Extremeness” metric



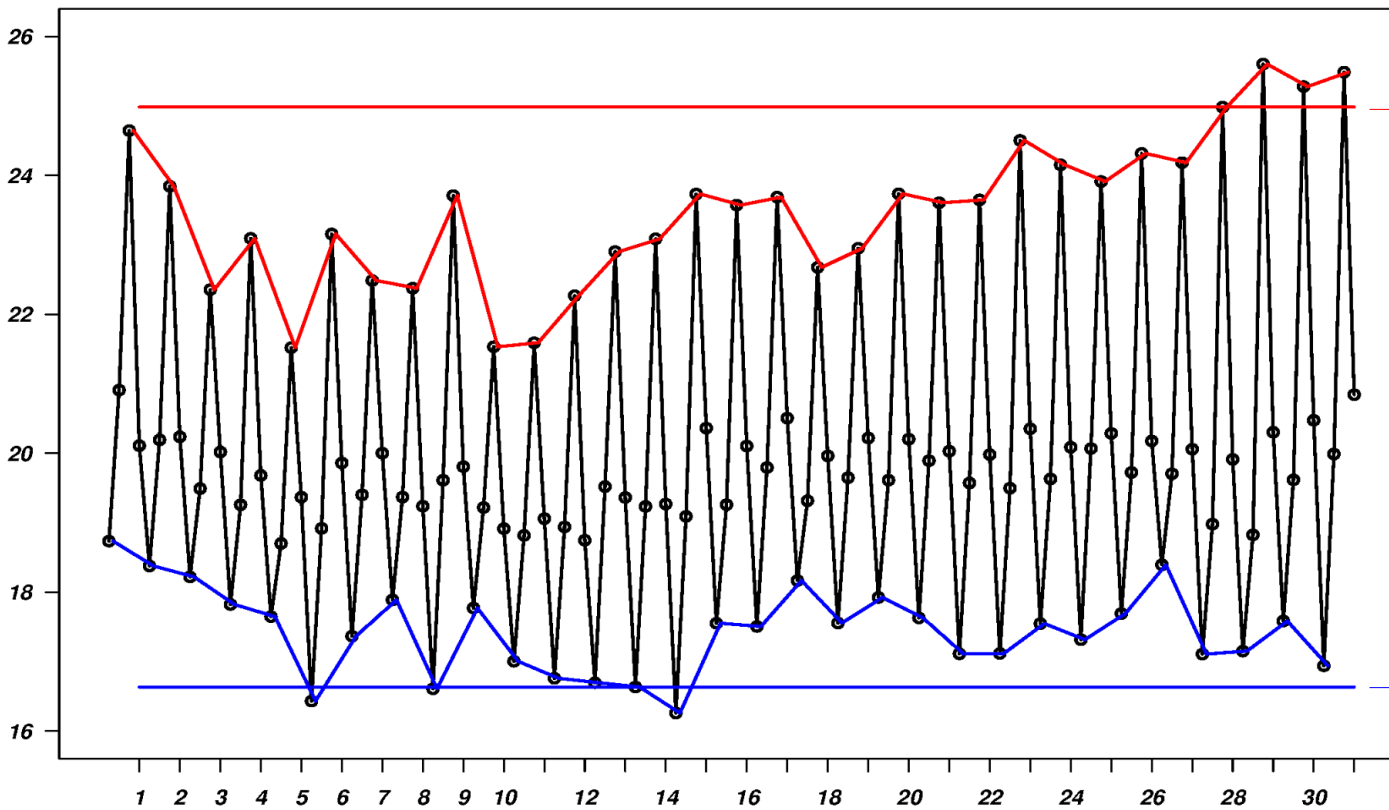
Six-hourly 2m-temperature ( $^{\circ}\text{C}$ ) in May 1985 from ERAint  
in one grid point in Europe



# “Extremeness” metric



Six-hourly 2m-temperature ( $^{\circ}\text{C}$ ) in May 1985 from ERA-Interim in one grid point in Europe



Monthly 90th percentile of  $T_x$

~  
Temperature during the warmest days

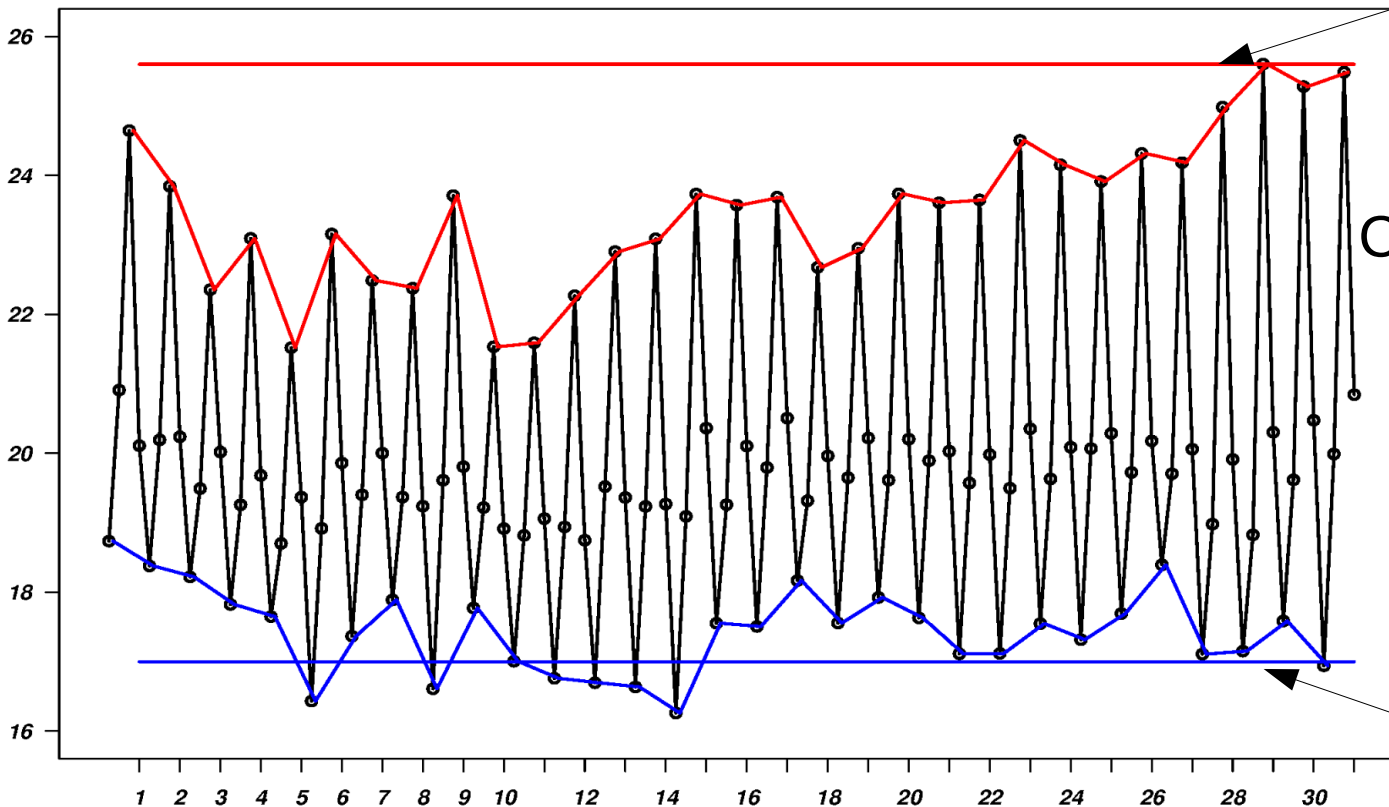
Monthly 10th percentile of  $T_n$

~  
Temperature during the coldest nights

# “Extremeness” metric



Six-hourly 2m-temperature ( $^{\circ}\text{C}$ ) in May 1985 from ERA-Interim in one grid point in Europe



Climatological 90th  
Percentile of  $T_x$   
over a given period  
(1981-2010)

~  
Climatological temperature  
of the  
warmest days

Climatological 10th  
Percentile of  $T_n$   
over a given period  
(1981-2010)

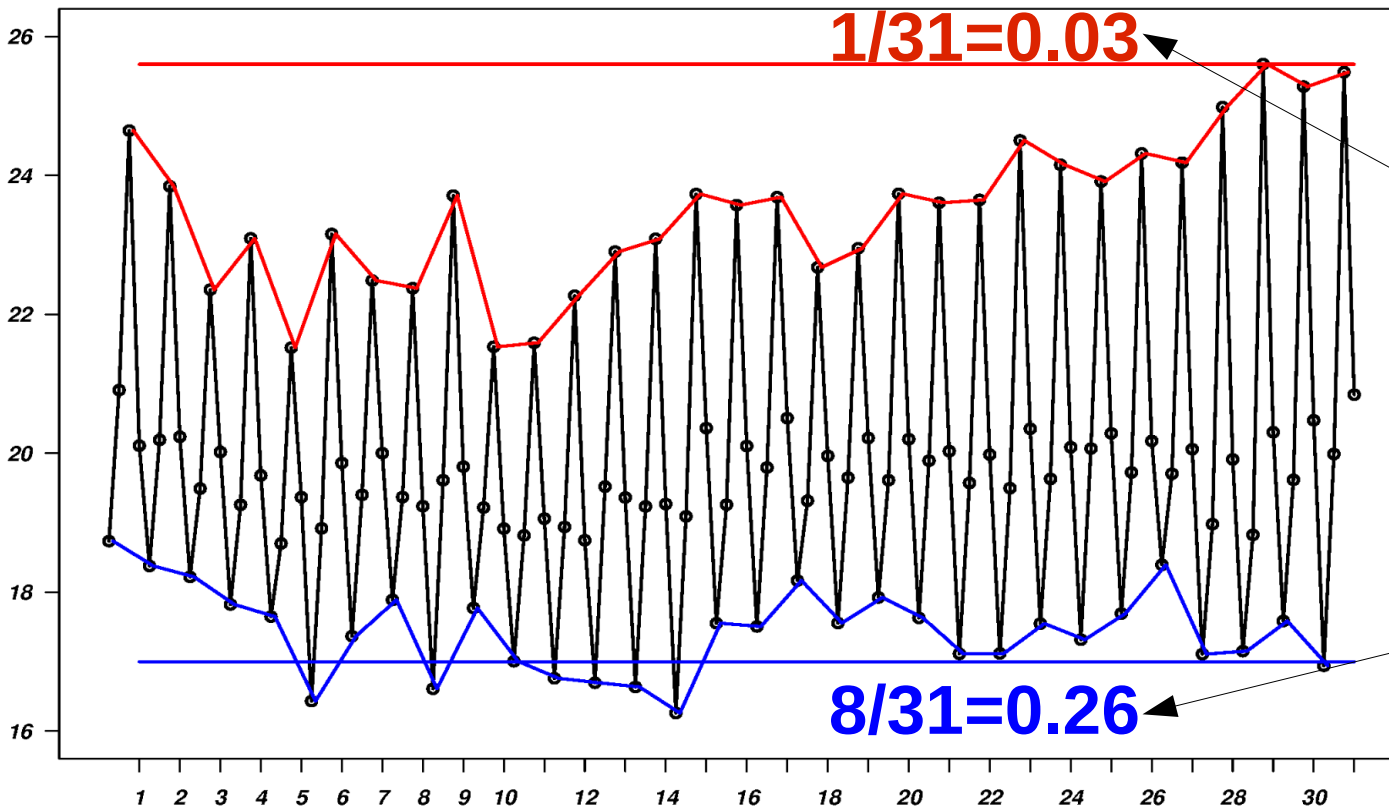
~  
Climatological temperature  
of the  
coldest nights



# “Extremeness” metric



Six-hourly 2m-temperature (°C) in May 1985 from ERA-Interim in one grid point in Europe



**Percentage of days**  
Over the  
Climatological 90th  
Percentile of Tx  
over the whole period  
(1981-2010)

~  
Number of warm days

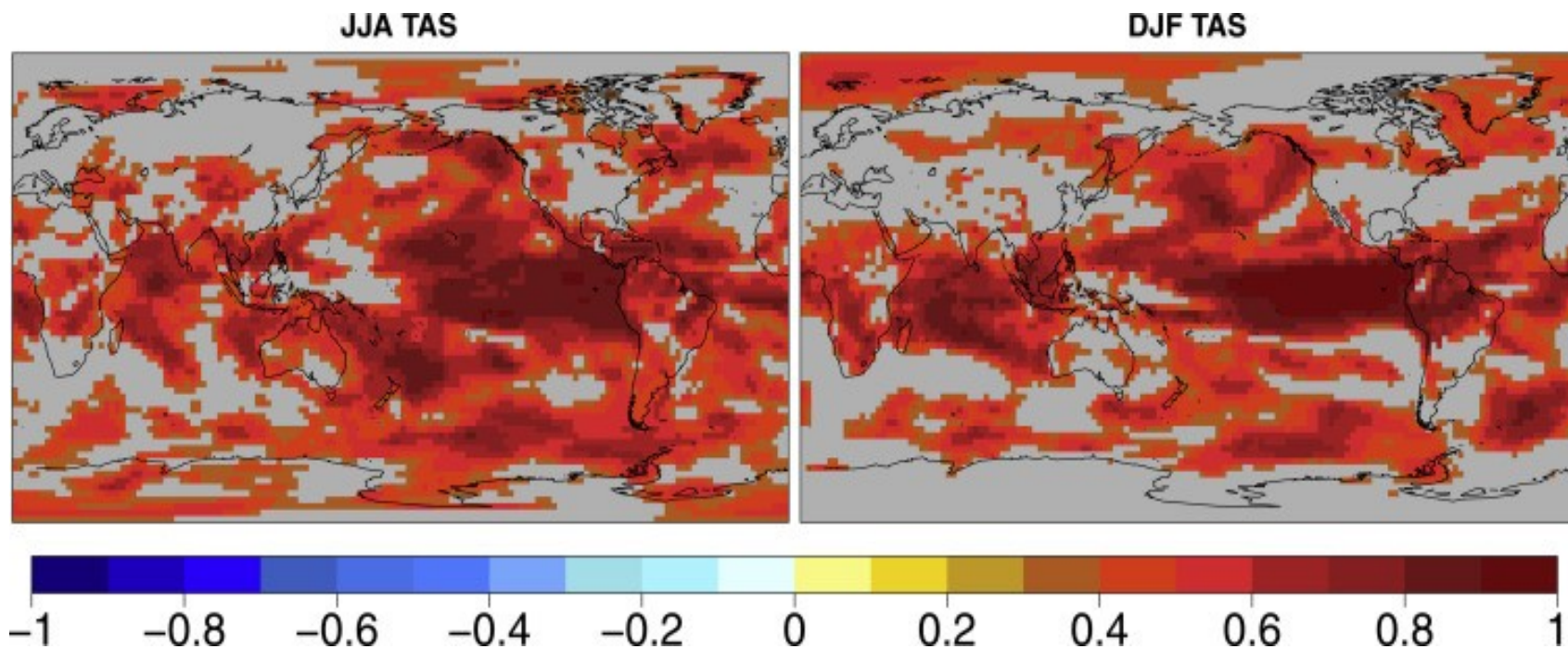
**Percentage of days**  
Under the  
Climatological 10th  
Percentile of Tn  
over the whole period  
(1981-2010)

~  
Number of cold nights

**More computationally expensive!**

**This method “bias-corrects” the hindcasts temperature distribution.**

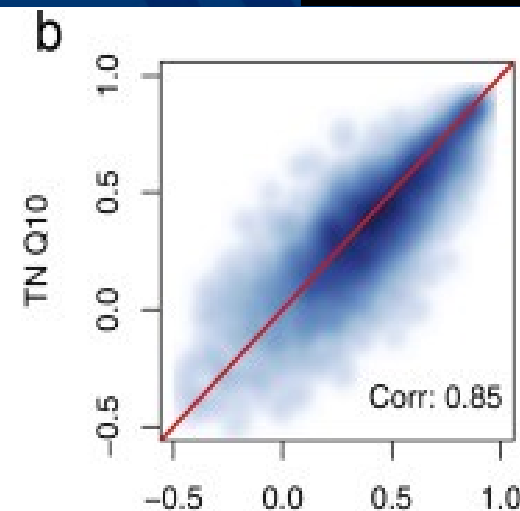
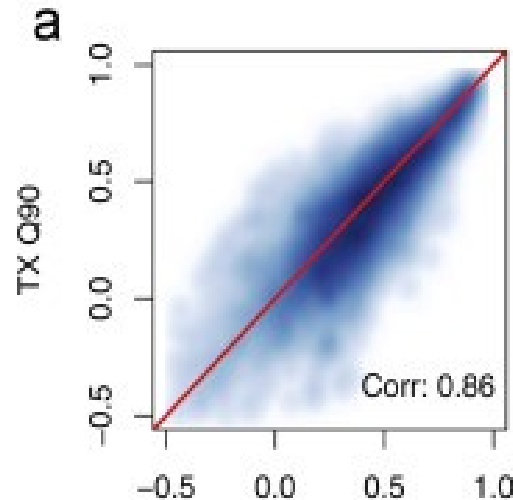
- The ENSEMBLES multi-model seasonal forecasts (Weisheimer et al., 2009):
- UK Met Office (UKMO)
  - Météo-France (MF)
  - European Centre for Medium-Range Weather Forecasts (ECMWF)
  - Leibniz Institute of Marine Sciences at Kiel University (IFM-GEOMAR)
  - Euro-Mediterranean Centre for Climate Change (CMCC-INGV) in Bologna. .



*Anomaly correlations between the ENSEMBLES ensemble mean temperature forecast and ERA-Interim reanalysis. May (left) and November (right) start date.*

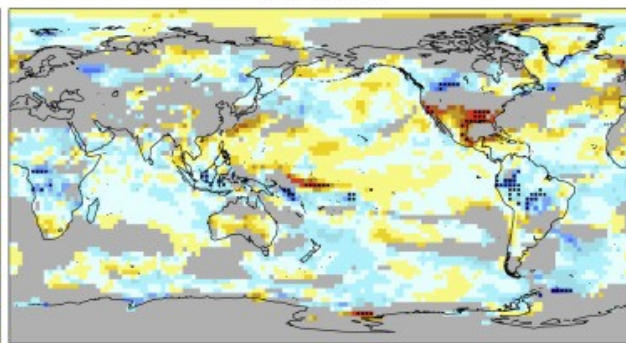
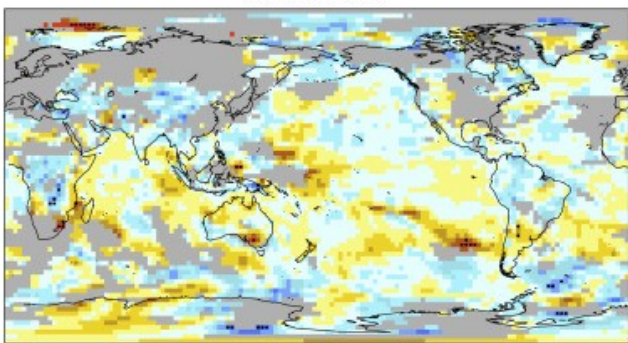
# “Extremeness” verification

*The difference between the anomaly correlation between the ERAI observations and the ENSEMBLES multi-model forecast mean as obtained for indices of seasonal extremes, and the same correlation for the seasonal mean temperature.*



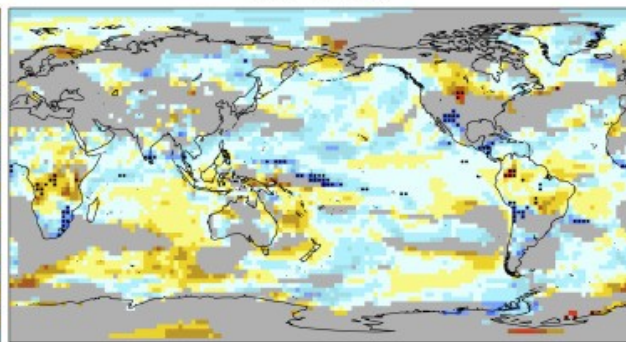
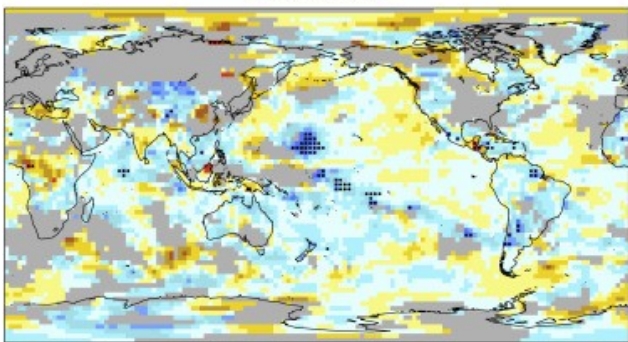
JJA TX Q90

DJF TX Q90



JJA TN Q10

DJF TN Q10



TAS

TAS

*Scatter plot of anomaly correlation for the mean daily temperature and the 90th and 10th percentiles over the entire globe in JJA*

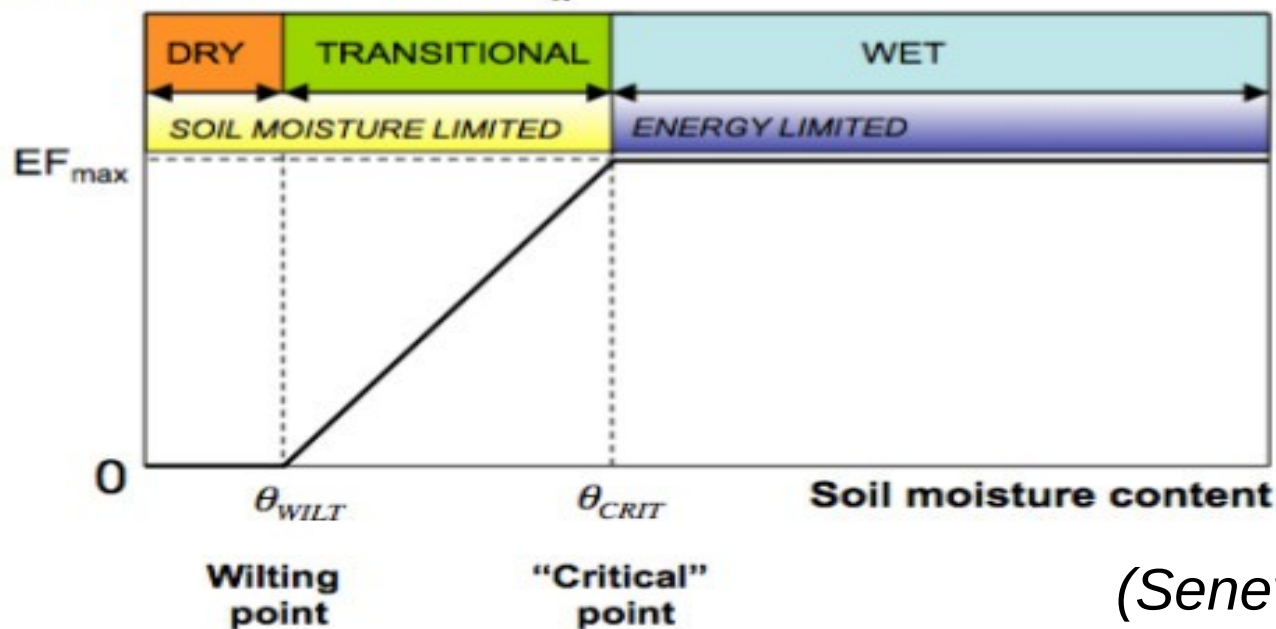
**Skill for mean and extreme is similar.**



According to the **user needs**, many other extreme variables can be defined:

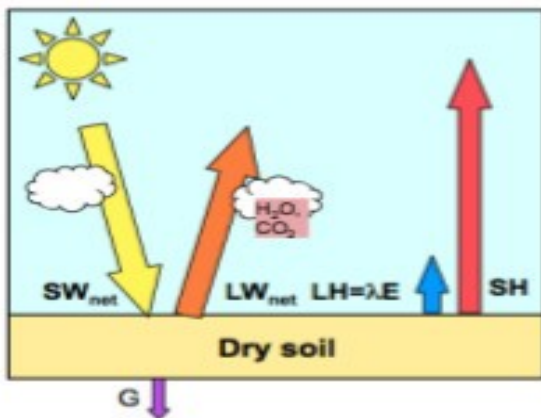
- Temperature and number of warm nights (Impact on human and animals health)
- Number of nights over a temperature threshold (cattle loss)
- Number of nights under 0°C (crop damage, road...)
- Extreme precipitation (floods)
- Number of consecutive dry/warm days (drought)
- Wind module exceeding a threshold (wind energy)
- .....

Evaporative fraction  $EF = \lambda E / R_n$

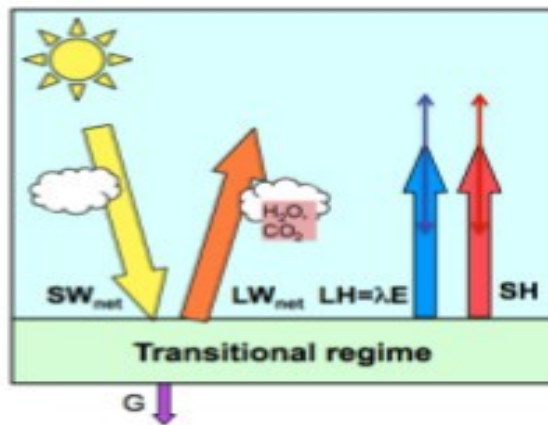


(Seneviratne et al., 2010)

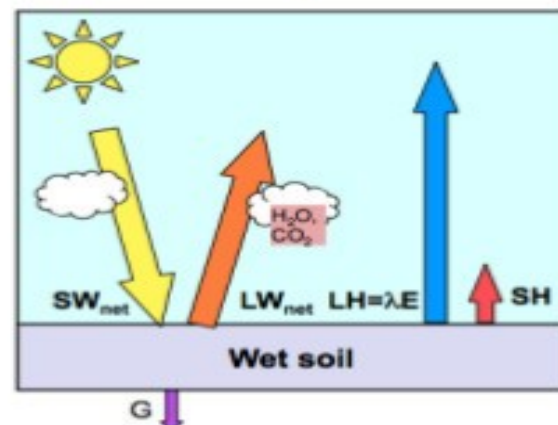
Dry climate regime

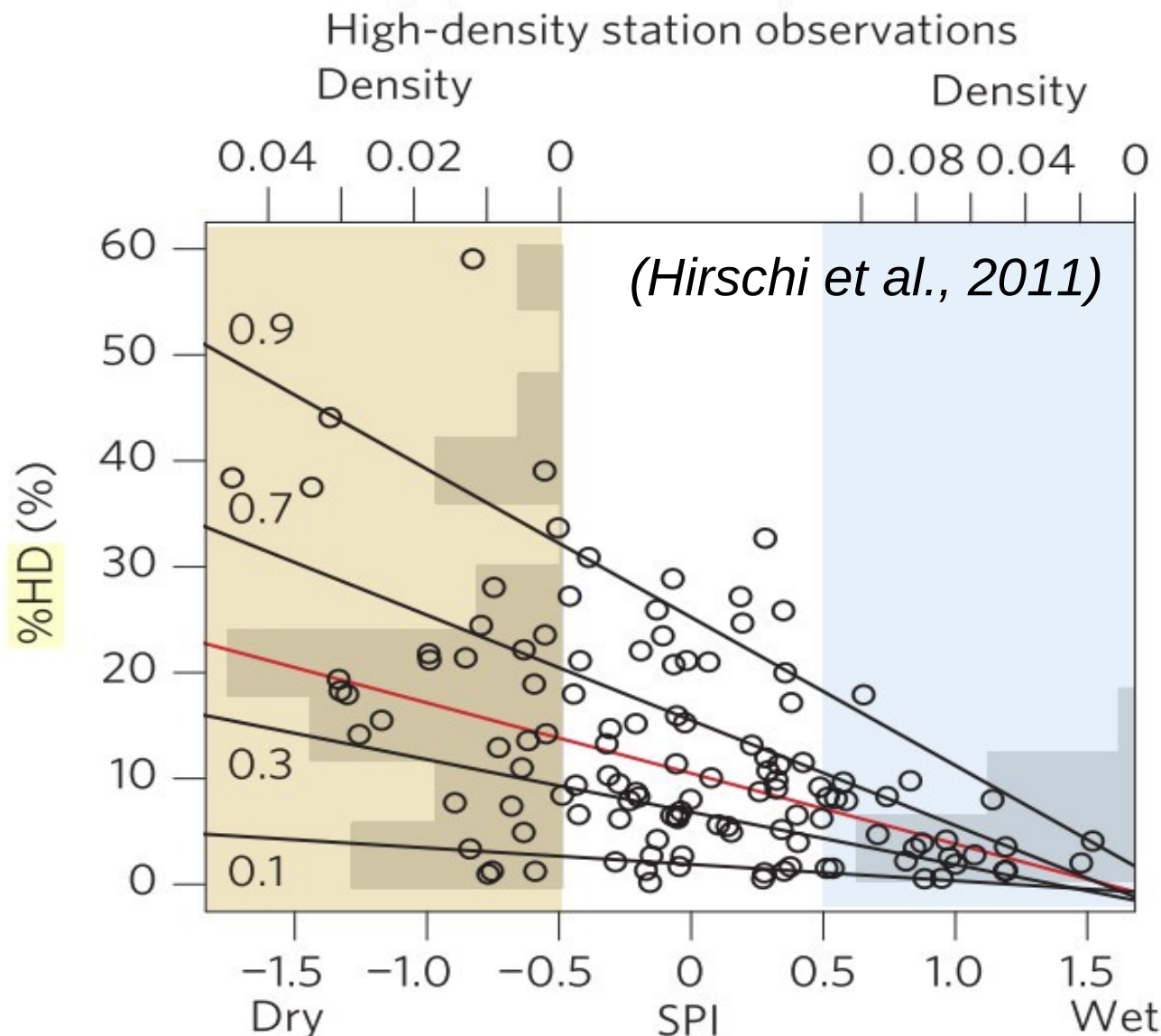


Transitional climate regime



Wet climate regime





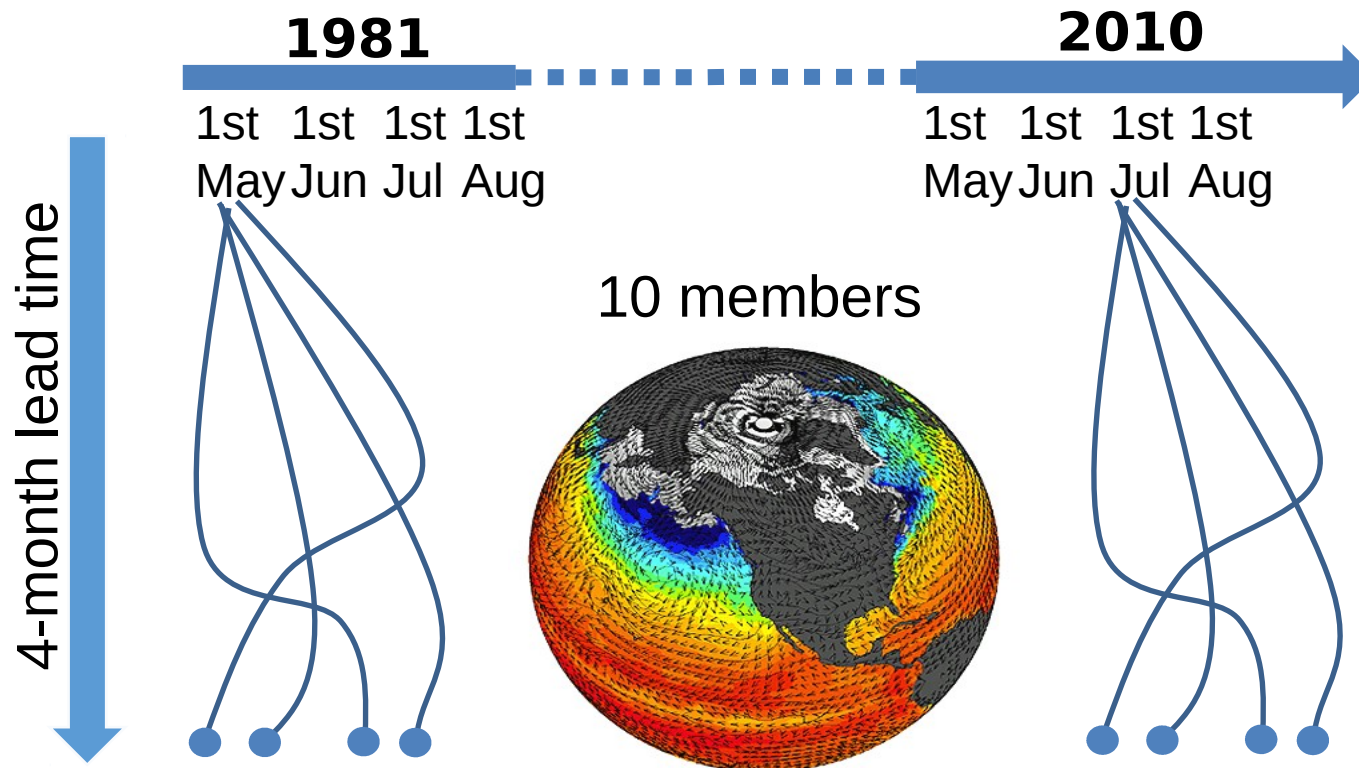
**Warm extremes only occur when the soil is dry.**

*Percentage of Hot Days (%HD) vs the Standardized Precipitation Index (SPI) in the southeast European domain (1961-2000 period).*

# Experiment description

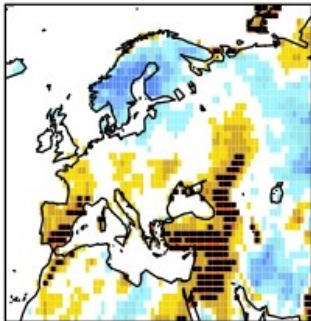


	Model	Start dates	Land IC	Atm IC	Oce/Ice IC
<b>INIT</b>	EC-Earth 2.3	May	<b>ERA-Land</b>	ERAInt	ORA-S4
<b>CLIM</b>	EC-Earth 2.3	May	<b>ERA-Land Climatology</b>	ERAInt	ORA-S4

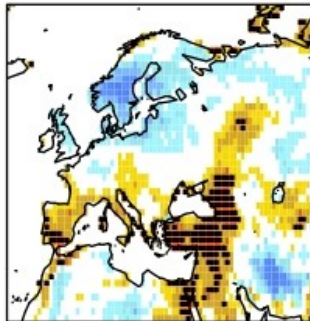


## Correlation between CLIM and ERA-interim

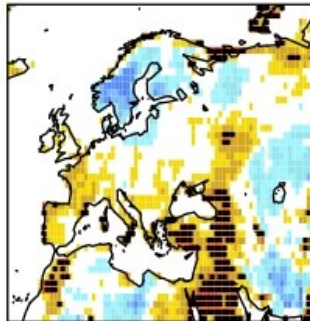
a) q90 of Tx



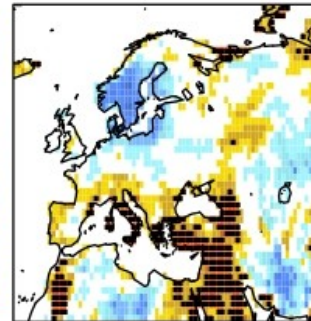
b) nb of warm days



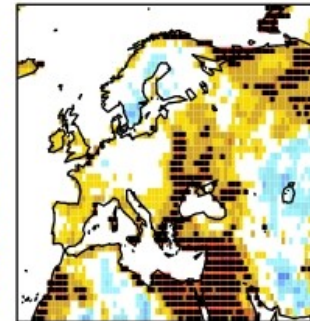
c) q90 of Tn



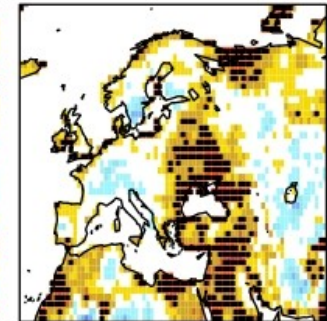
d) nb of warm nights



e) q10 of Tn

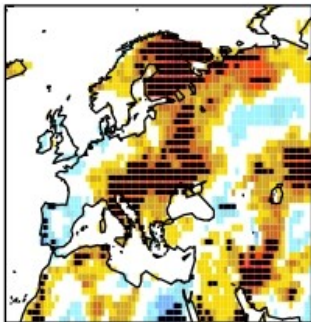


f) nb of cold nights

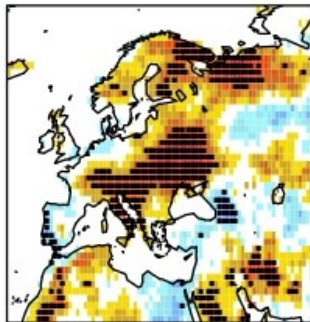


## Difference of correlation: INIT-CLIM

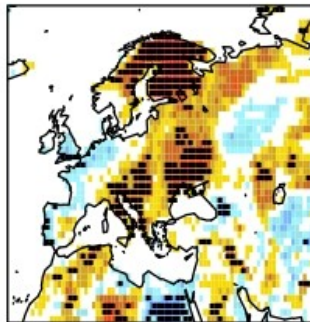
g) q90 of Tx



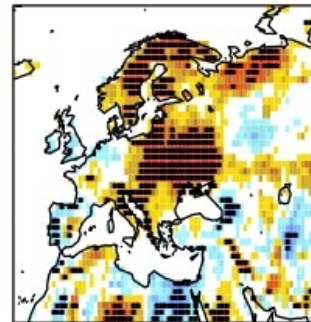
h) nb of warm days



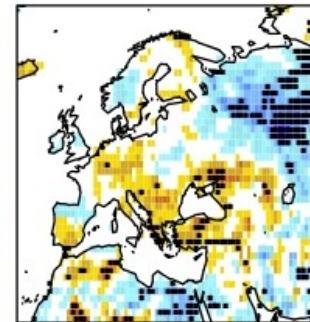
i) q90 of Tn



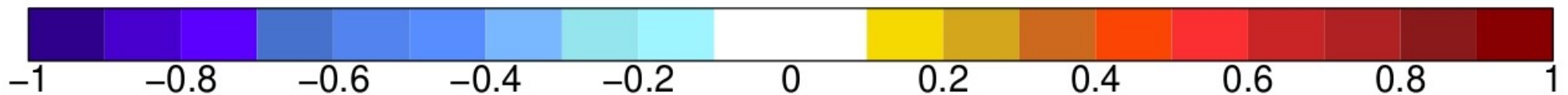
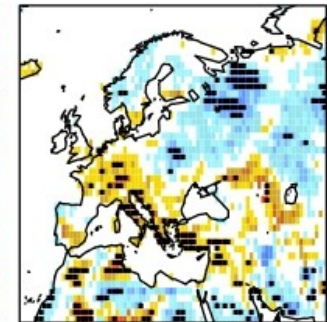
j) nb of warm nights



k) q10 of Tn



l) nb of cold nights



**With realistic soil initialization skill of warm extremes significantly increases**



2m-Temperature anomalies (JJA)

**2003**



The European heat-wave of 2003 caused the death of 35,000 people and damages of \$15 billion.

**2010**

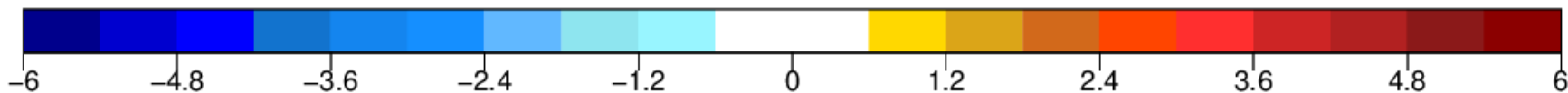
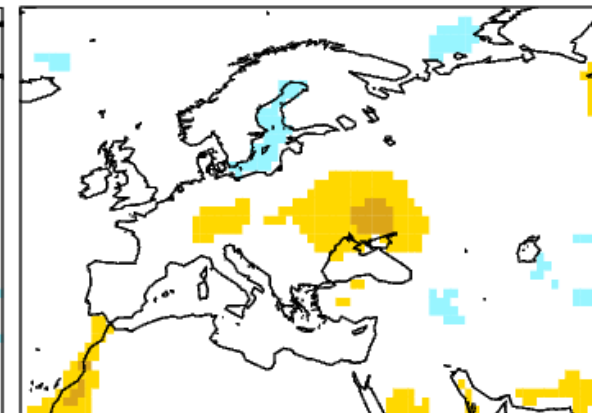
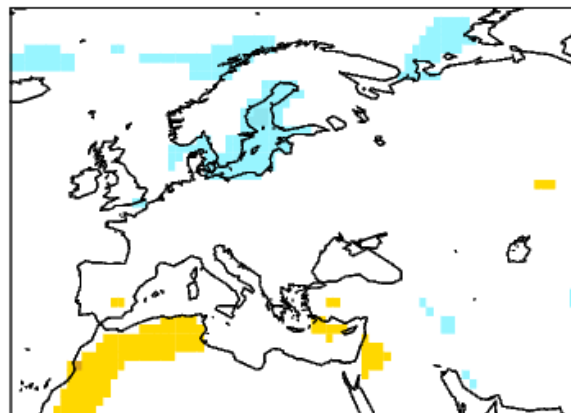
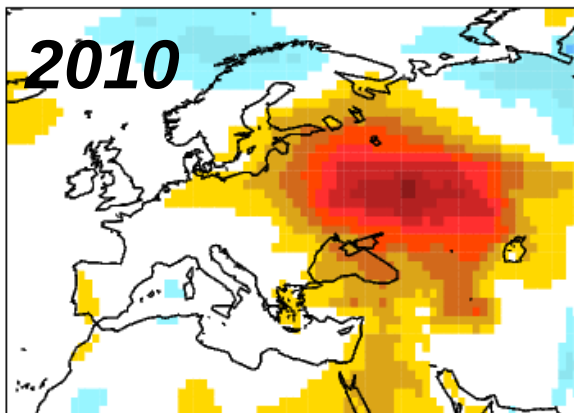
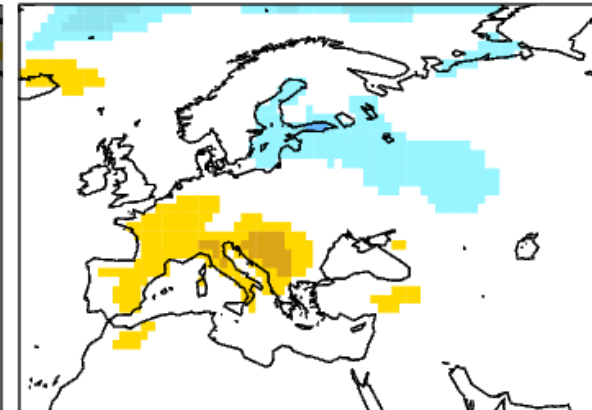
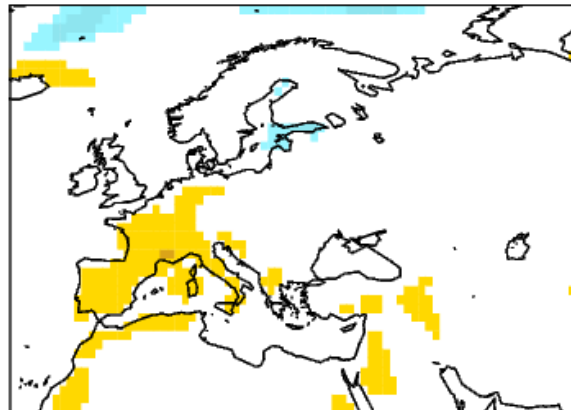
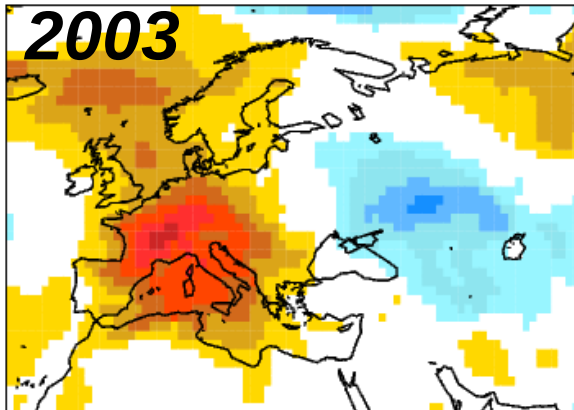


# Are they predictable?

ERAint: 2m temperature

CLIM: 2m temperature

INIT: 2m temperature



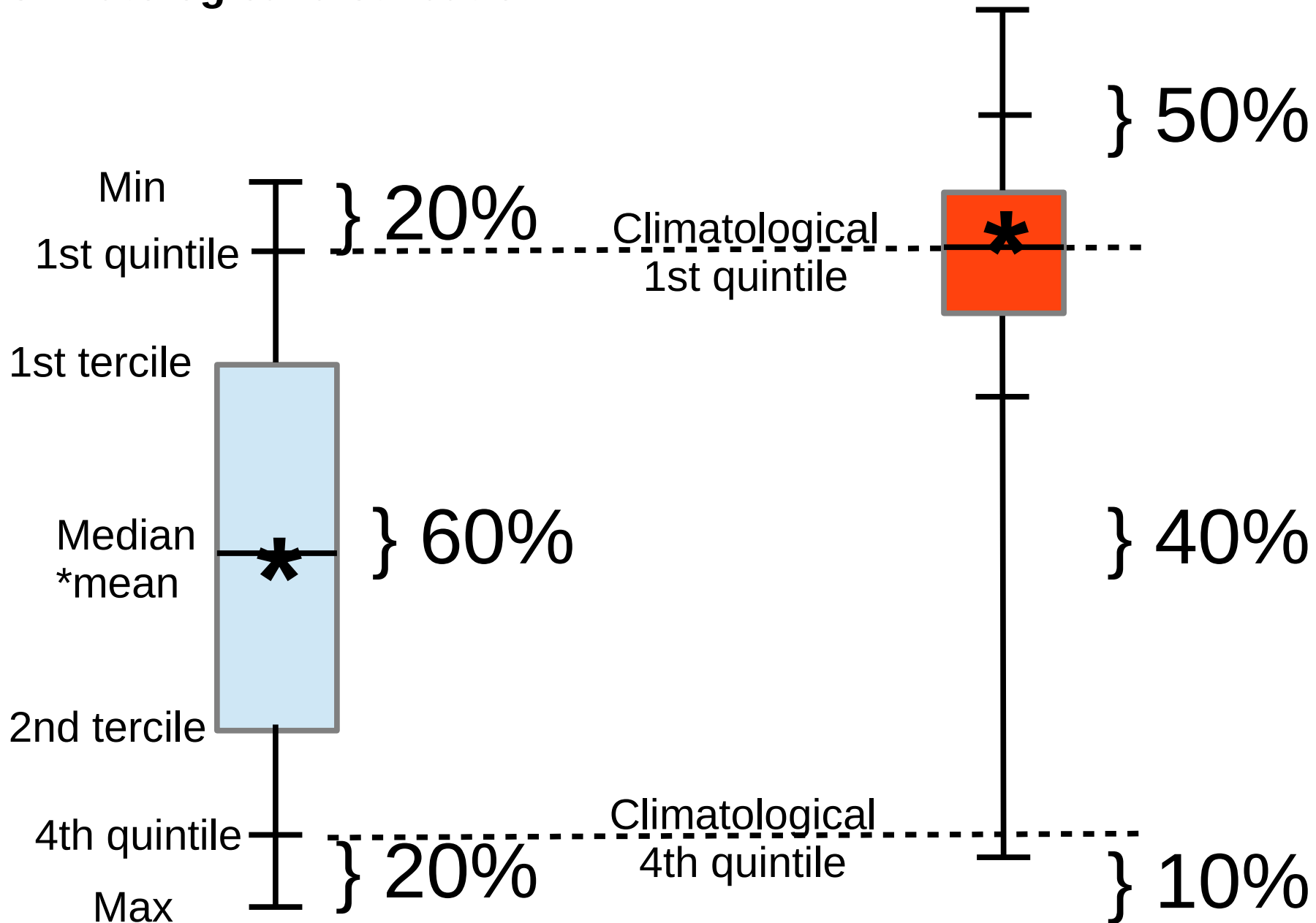
- 2003 and 2010 seems to be predictable, soil initial conditions seem to be important only for 2010.
- Using ensemble mean anomaly smooths out the signal and is not adequate to assess the ensemble predictions.

# Odds ratio: Metric for heat waves



## Climatological distribution

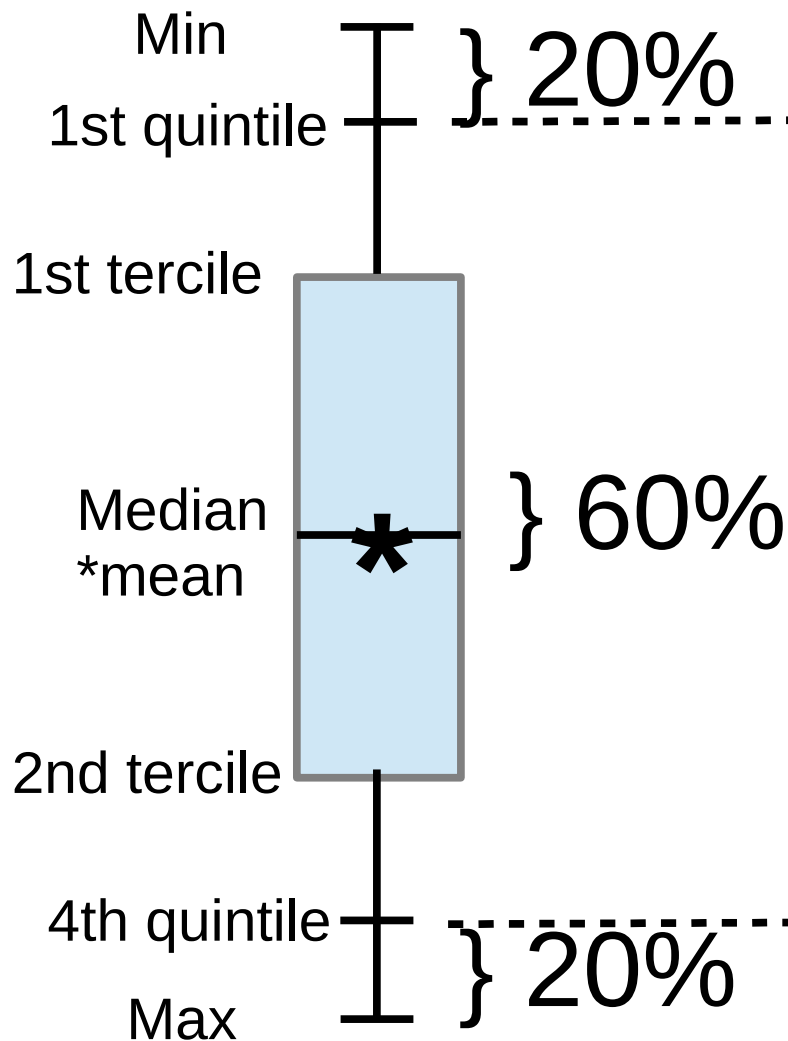
## Warm-event distribution



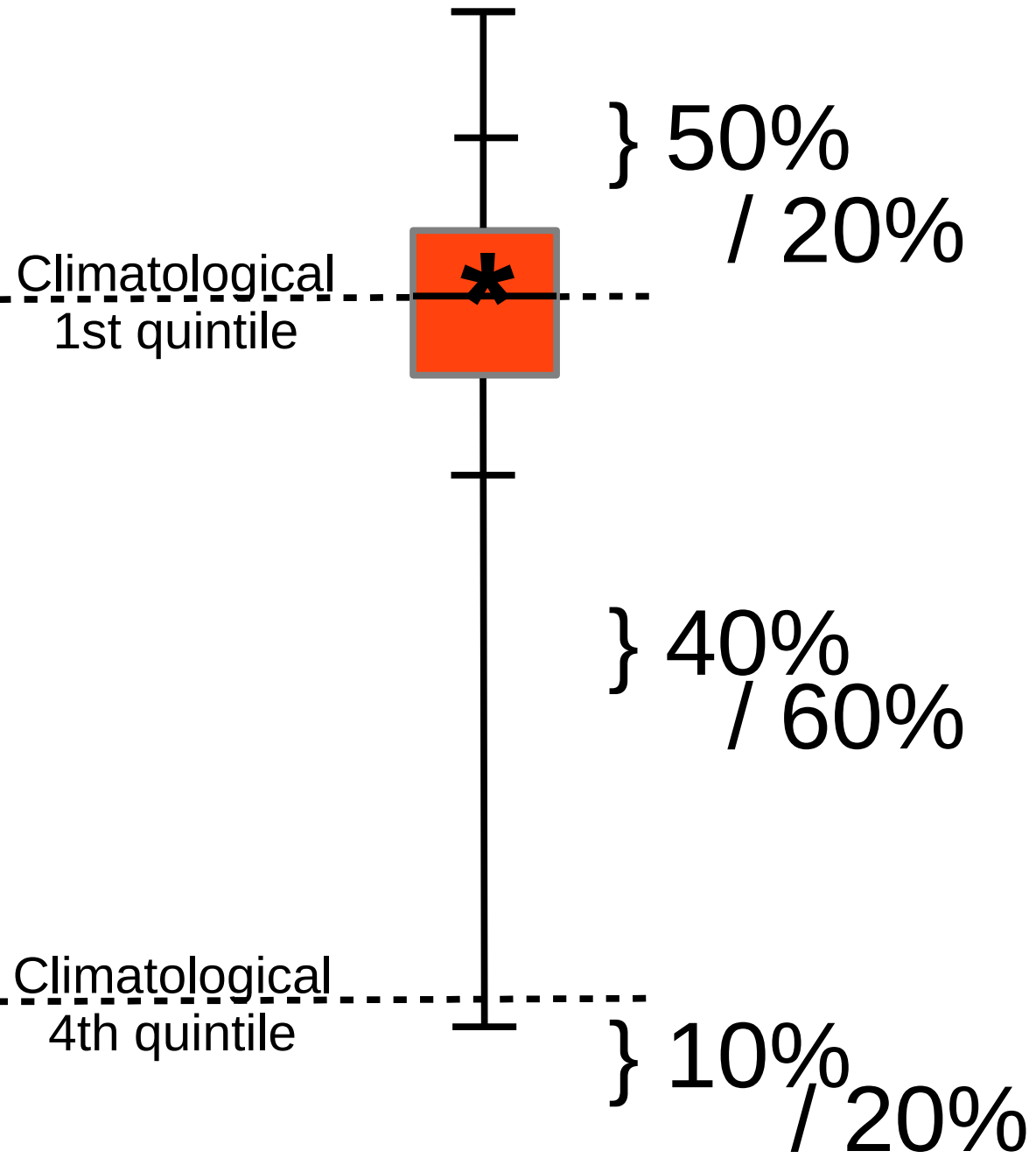
# Odds ratio: Metric for heat waves



## Climatological distribution



## Warm-event distribution

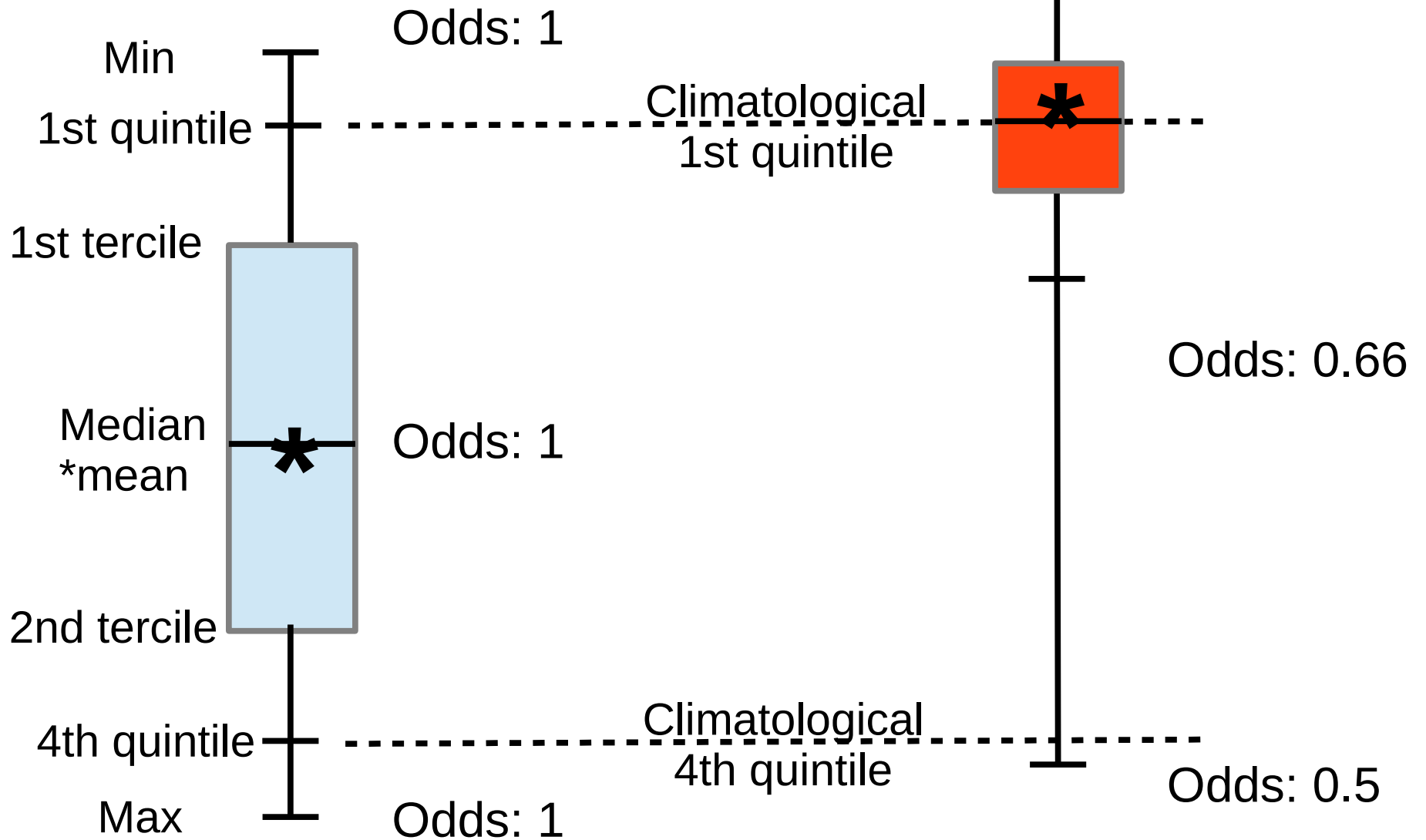


# Odds ratio: Metric for heat waves

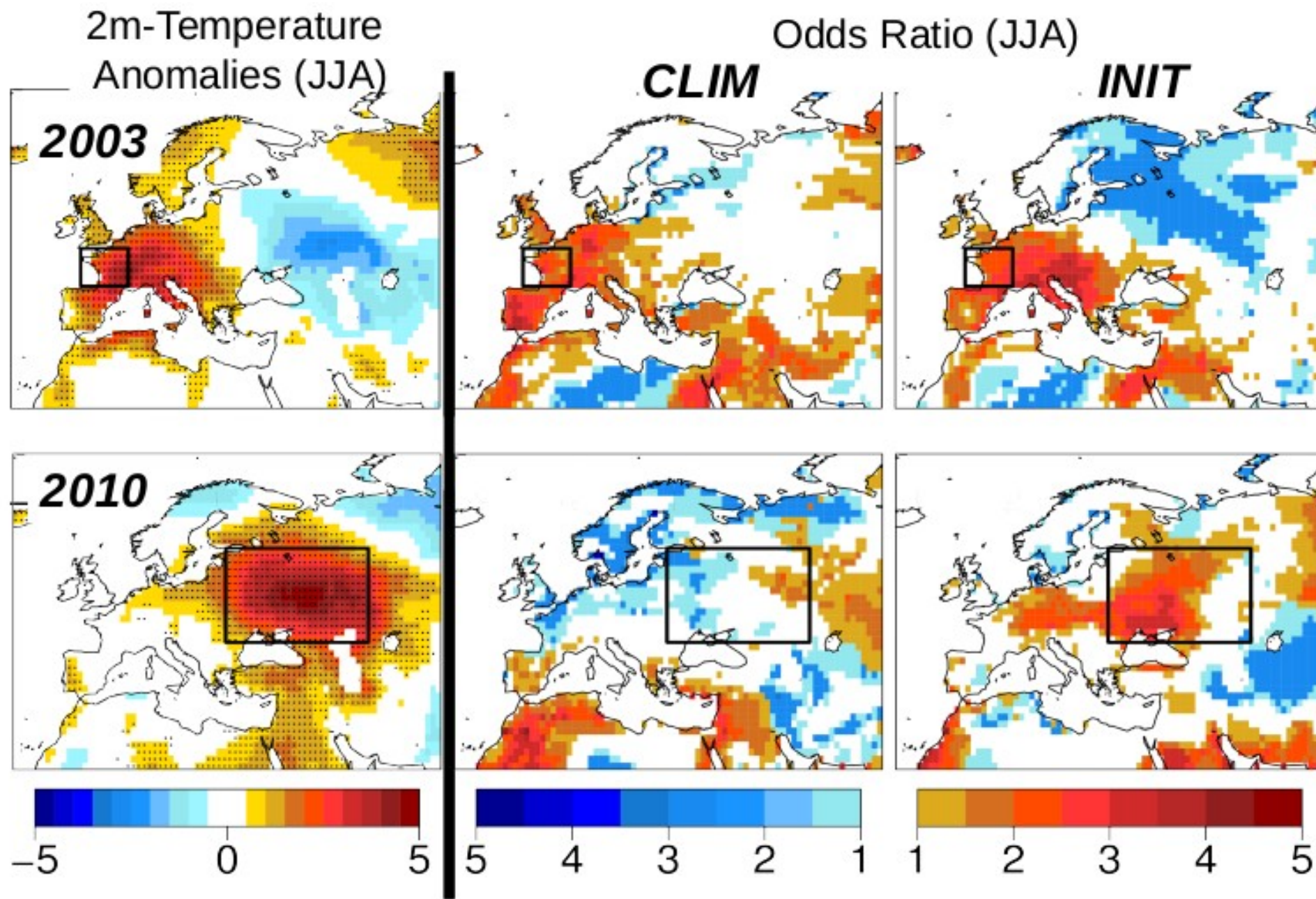


## Climatological distribution

## Warm-event distribution



# Odds ratio: Metric for heat waves



- Extreme temperature variables can be predicted with skill equivalent to the skill of the mean.
- Land-atmosphere coupling is extremely important for extreme temperature and realistic soil initial conditions are important for the skill of extreme temperature.
- Heatwaves can be predicted and soil initial conditions can be important for predicting the heatwave development.
- The ensemble-mean anomaly is not suitable for expressing the extremes; metrics based on the forecast distribution such as the odds ratio are more adequate.
- Extreme variables should also be defined from the users needs.