

# *Uncertainties of regional climate projections*

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Validating Regional Climate Projections,  
4th VALUE Training School – Triest, 26-30 Oct 2015

## *Sources of the errors and uncertainties of downscaled climate simulations*

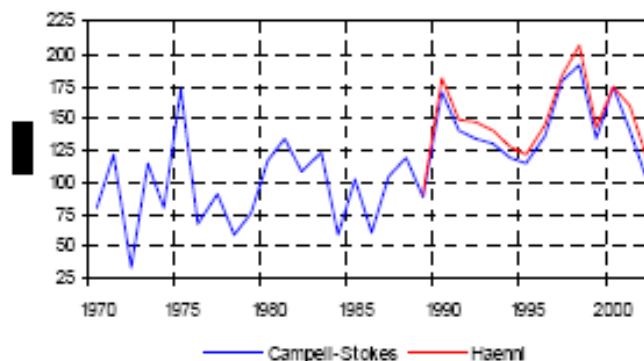
- ❑ *an imperfect model formulation,*
  - *errors of the driving GCM,*
  - *errors inherent in the downscaling approach,*
  - ***errors in observations themselves,***
- ❑ *uncertain future concentrations of GHGs,*
- ❑ ***internally generated climate variability.***

## *Uncertainty begins with the observations:*

- *How accurate are the instruments used for measurements?*
- *How the measurement device sensitivity influences the quality of measurements ?*
- *How unstable and affected by biases is the observing system over time?*
- *How intermittent are observations in space and time?*
- *How representative are those observations of the true ambient climate at that point in space and time?*



# *The influence of sensitivity of measuring device on measured variables*



## *What about the unhomogeneity of observations?*

- *Changes in environment around the station*
- *Changes in measurement procedures*
- *Changes in instruments*
- *Changes in observers*
- *Changes in density of measurement net*

## *Then the data have to be assimilated in models:*

- *How strong is the impact of processing steps to move from raw observational data to a gridded Climate Data Record (CDR)?*
- *Which method should be selected for doing it?*
- *How it influences the resulting simulations?*

## *An example of station density used for gridding*

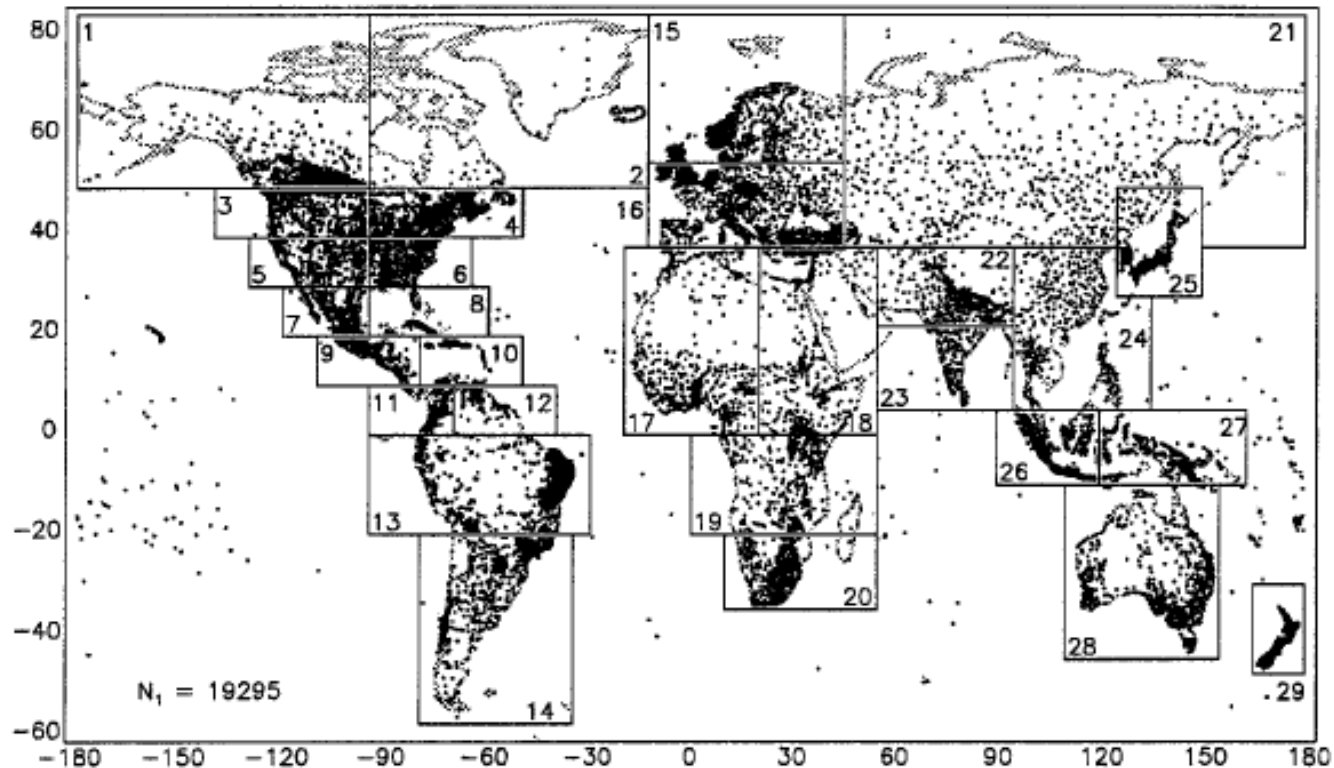
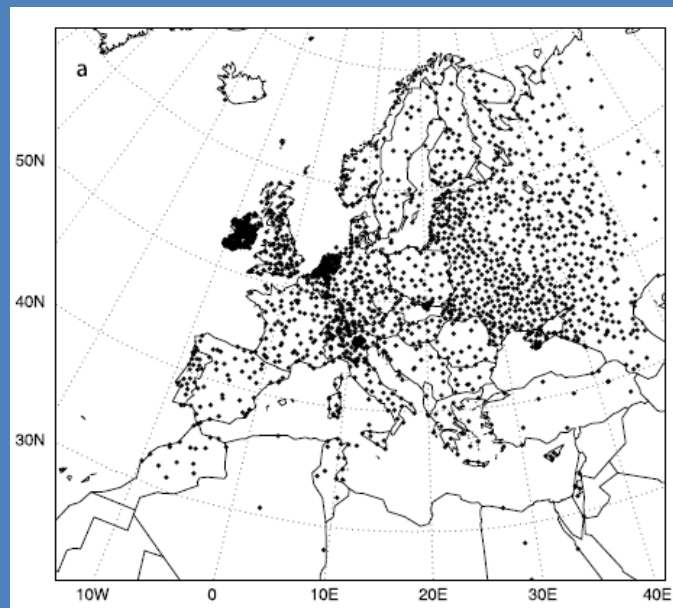
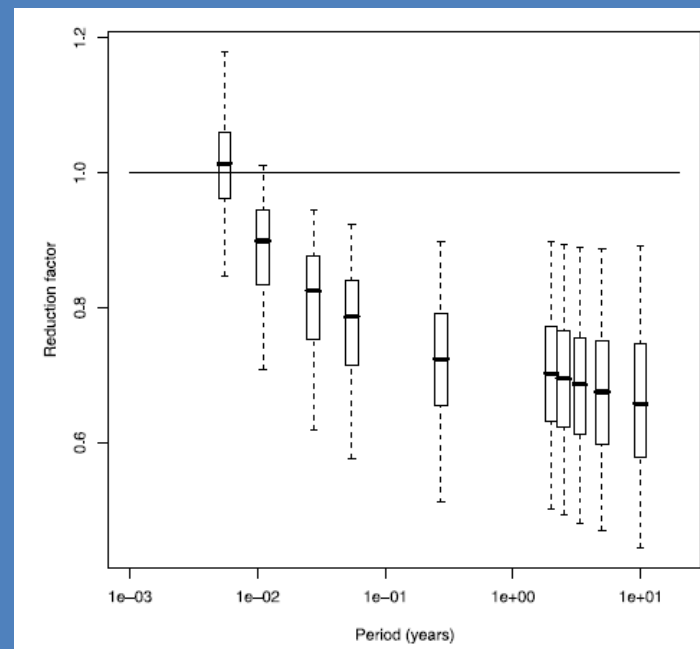


FIG. 1. Location of stations with precipitation normals. Geographic tiles used in the interpolation are shown and  $N$  signifies the total number of stations used. Note that (i) for all variables, oceanic stations were used during the interpolation of a global "background" tile and (ii) tile numbers and sizes differ between variables.

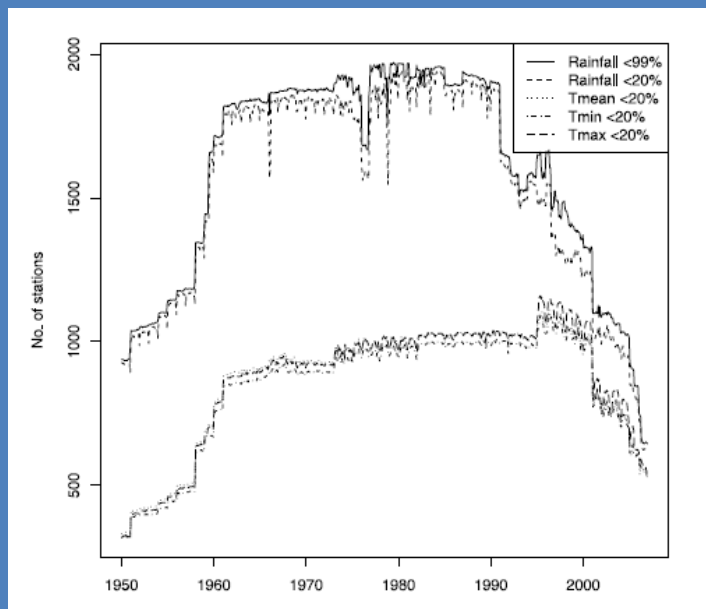


The complete gridding region (land-only), showing the station network for precipitation



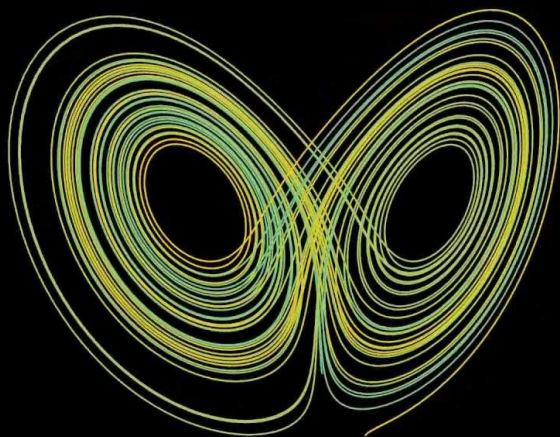
Areal reduction factor for daily quantiles of precipitation from the median (50% quantile) up to the 10-year return level.

The number of stations with less than 99% and 20% missing observations for each month.





length: 5033.51  
rho = 28.00  
sigma = 10.00  
b = 2.67



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[https://commons.wikimedia.org/wiki/File:Lorenz\\_attractor\\_yb.svg#/media/File:Lorenz\\_attractor\\_yb.svg](https://commons.wikimedia.org/wiki/File:Lorenz_attractor_yb.svg#/media/File:Lorenz_attractor_yb.svg)

In chaos theory, the **butterfly effect** is the sensitive dependence on initial conditions in which a small change in one state of a deterministic nonlinear system can result in large differences in a later state. The name of the effect, coined by Edward Lorenz, is derived from the metaphorical example of the details of a hurricane (exact time of formation, exact path taken) being influenced by minor perturbations such as the flapping of the wings of a distant butterfly several weeks earlier. Lorenz discovered the effect when he observed that runs of his weather model with initial condition data that was rounded in a seemingly inconsequential manner would fail to reproduce the results of runs with the unrounded initial condition data. A very small change in initial conditions had created a significantly different outcome.

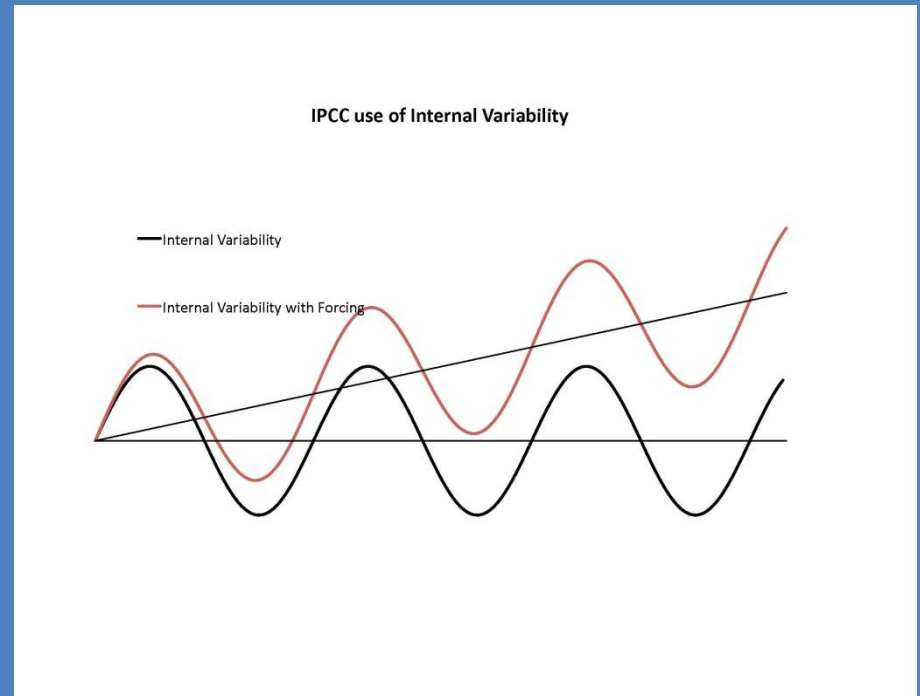
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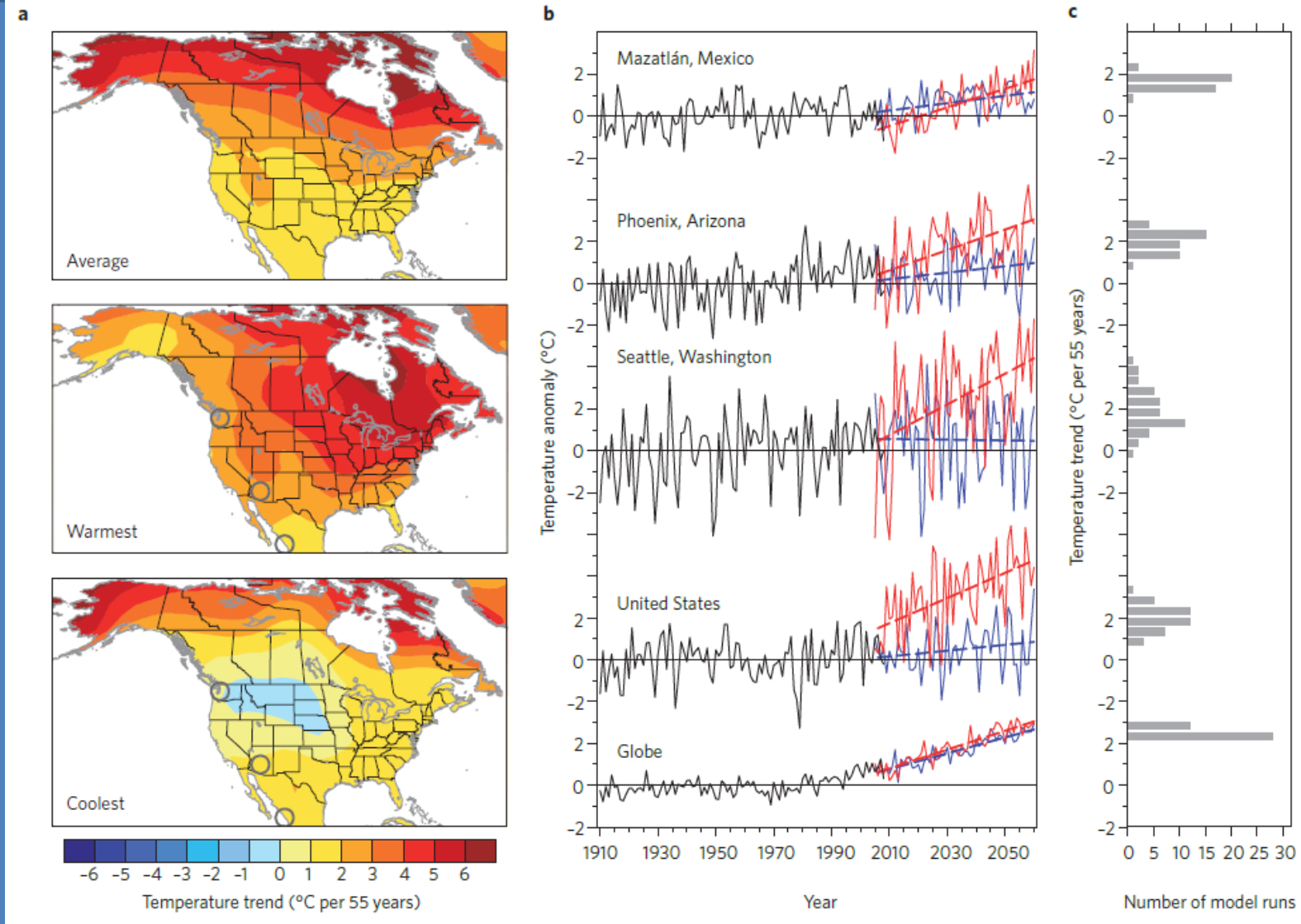
# *What is internal variability?*

*Variability due to natural internal processes within the climate system.*

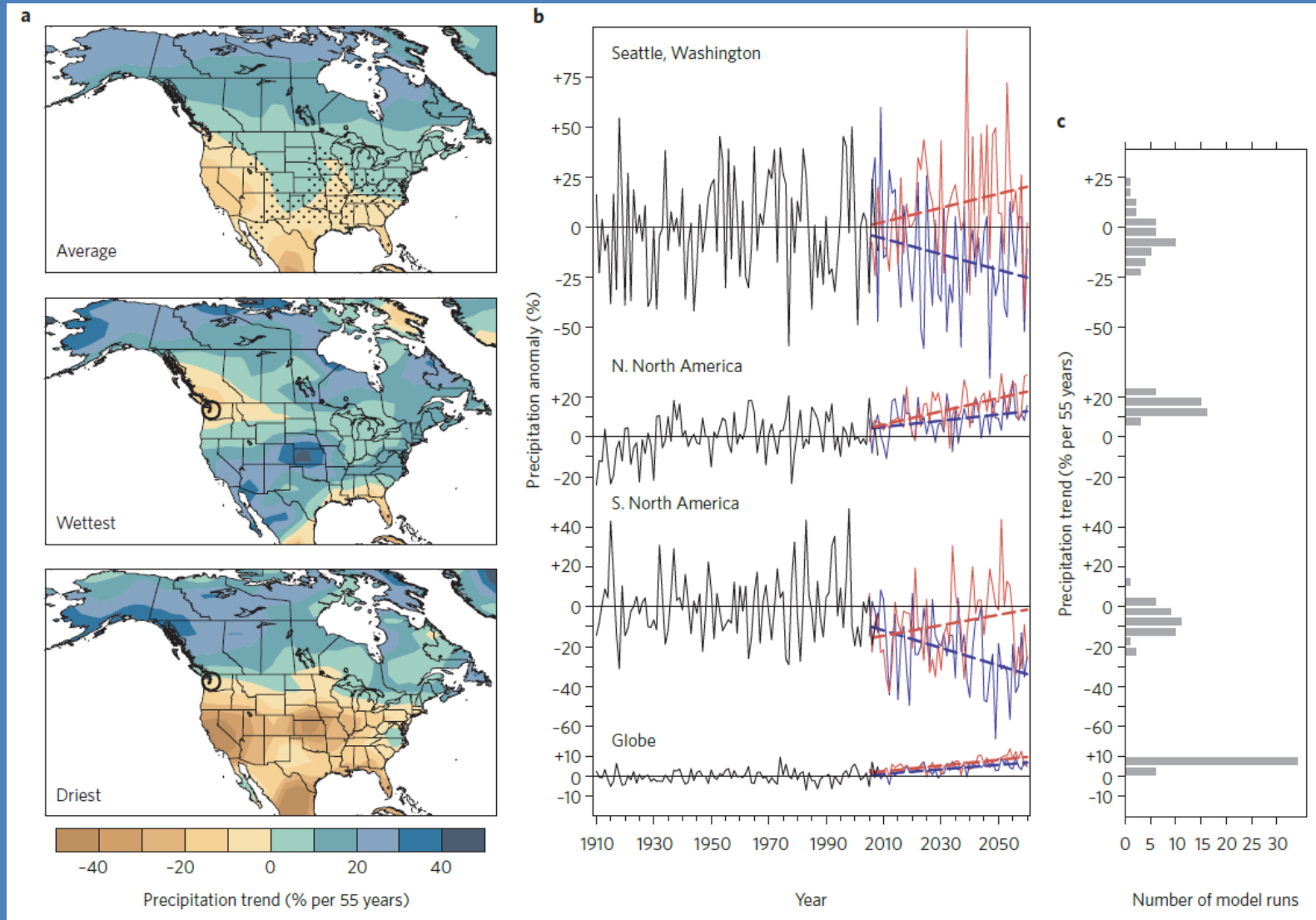
*Known examples:*

- *ENSO,*
- *AMO,*
- *PDO,*
- *Thermohaline Circulation*





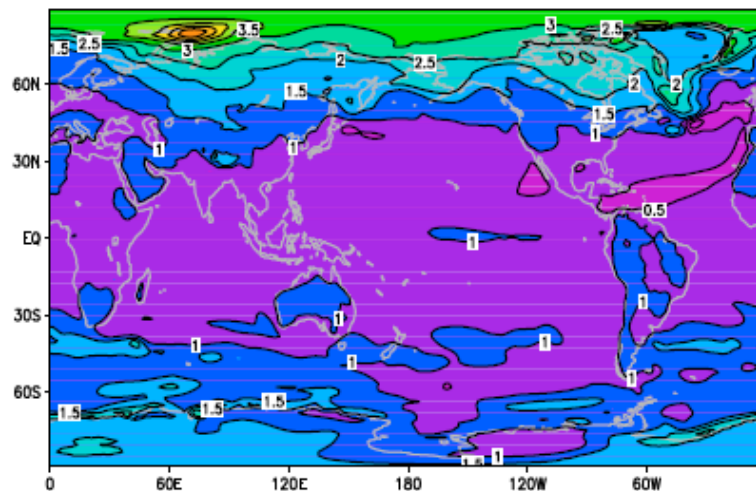
Range of future climate outcomes. **a**, (DJF) temperature trends during 2005–2060; **b**, DJF temperature anomaly time series for selected places. Black - observed anomalies from 1910 to 2008; red and blue - model projections for 2005–2060 from the realizations with the largest and smallest future trends, respectively, with the best-fit linear trends. **c**, Distribution of projected DJF temperature trends (2005–2060) across the 40 ensemble members at the locations shown in panel



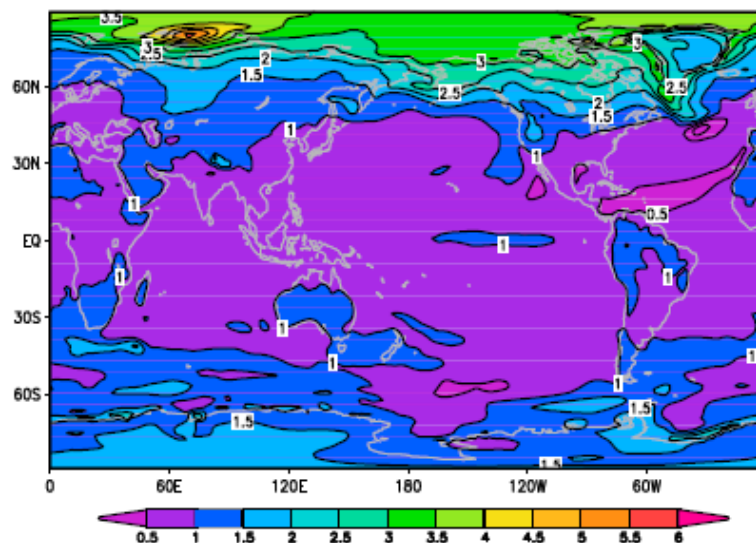
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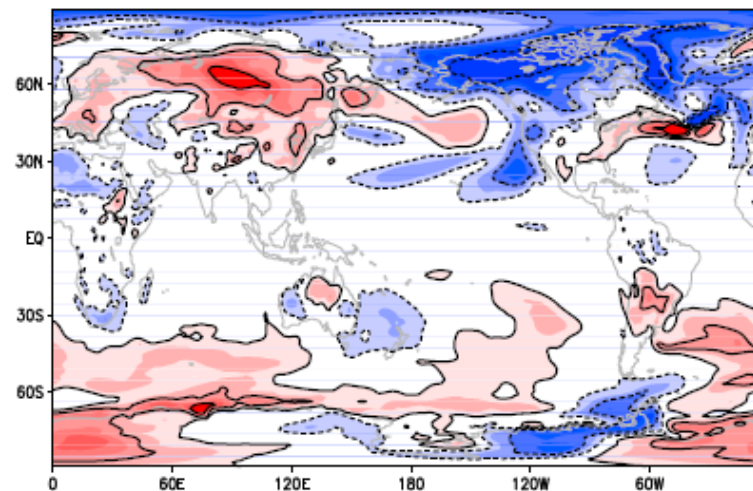
(a) SAT change (1990–1850) under AMV+ , glb mean:1.05 K



(b) SAT change (1990–1850) under AMV- , glb mean:1.07 K



(c) Difference (a) – (b)



(d) SAT Difference from linear regression

