

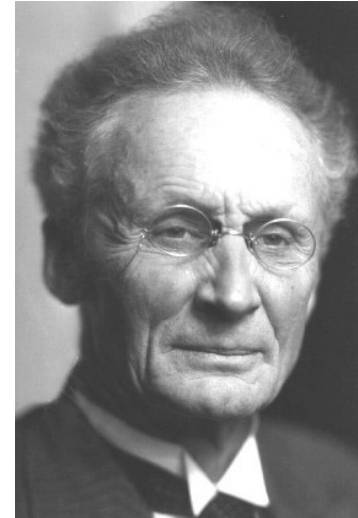


Global climate models and downscaling

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Institute of Geosciences and Earth Resources
National Research Council of Italy

with contributions from
J. von Hardenberg and E. Palazzi

Meteorological predictions and the use of numerical methods



Wilhelm Bjerknes
(1862-1952)

He suggested (1904) to consider weather forecast as an initial value problem, to be solved using the equations of Mathematical Physics.

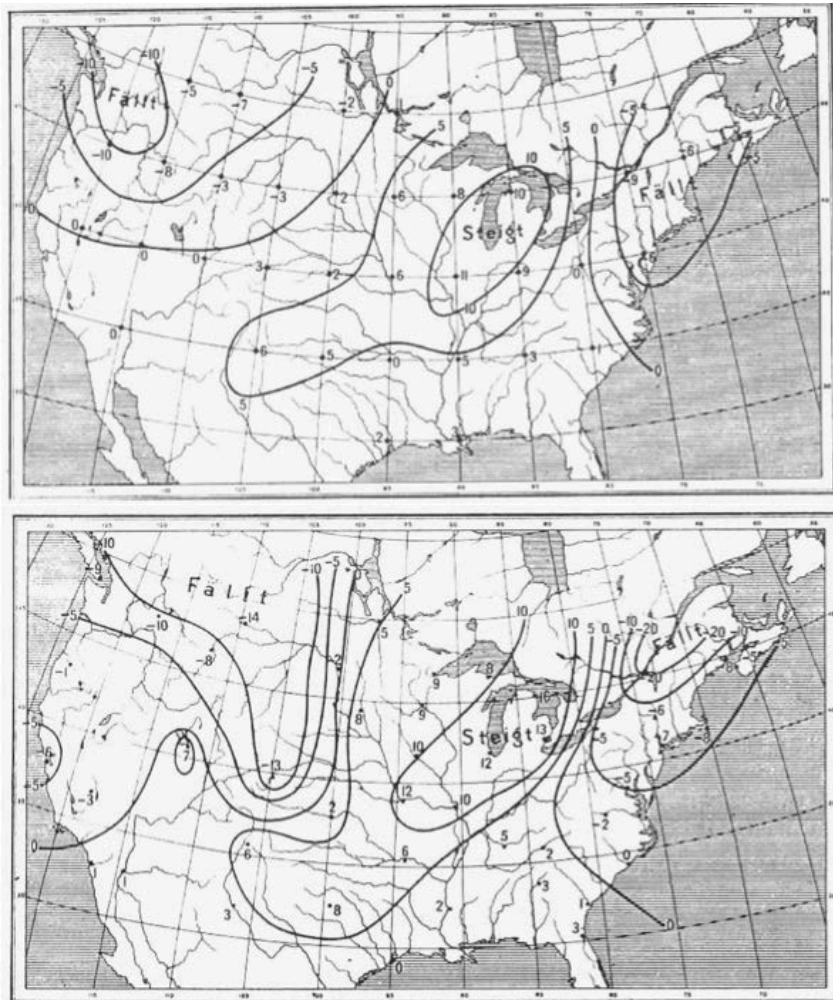
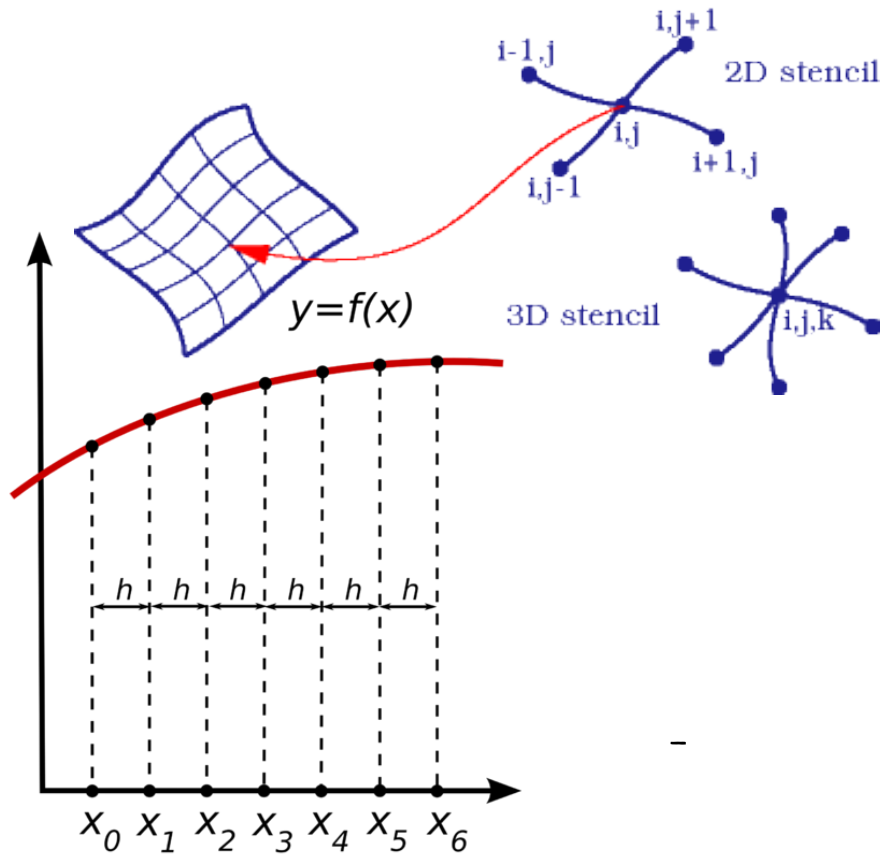


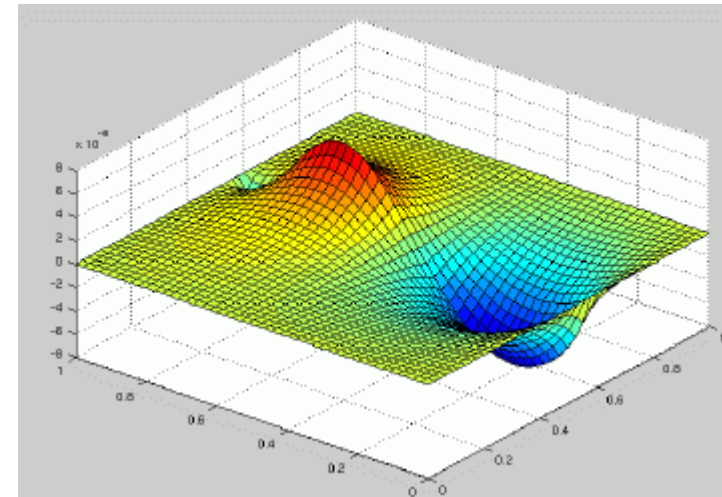
Figure 1.3 Top: Exner's calculated pressure change between 8 p.m. and midnight, 3 January 1895. Bottom: observed pressure change for the same period [Units: hundredths of an inch of mercury. *Steigt* = rises; *Fällt* = falls]. (Exner, 1908)

Finite difference methods

- We use a discretized formulation on a grid in space and time:



$$f'(a) \approx \frac{f(a+h) - f(a)}{h}$$



The Euler method

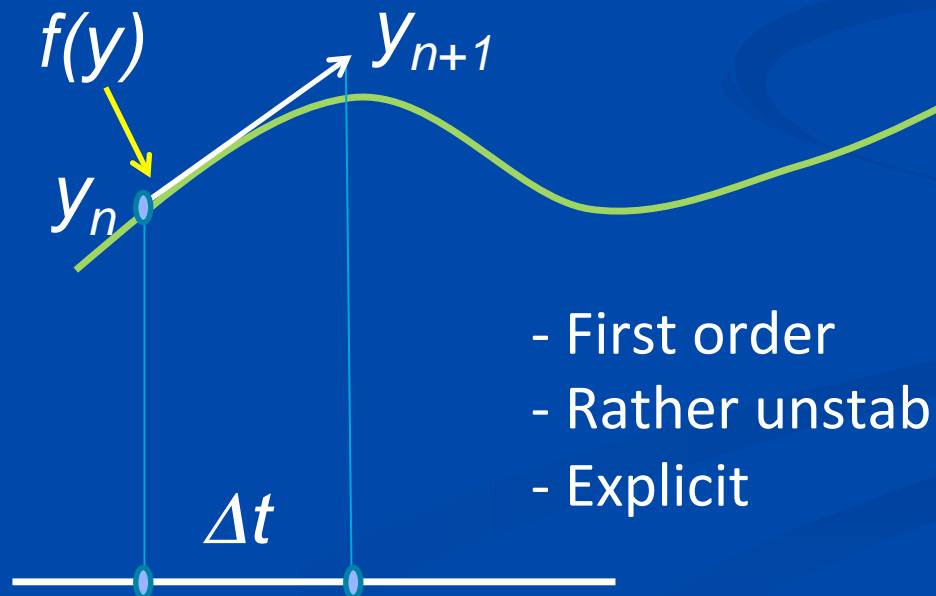
$$\frac{dy}{dt} = f(y)$$



$$y(t + \Delta t) = y(t) + \Delta t f(y(t))$$

$$y_{n+1} = y_n + \Delta t f(y_n)$$

+ Truncation
error



- First order
- Rather unstable
- Explicit



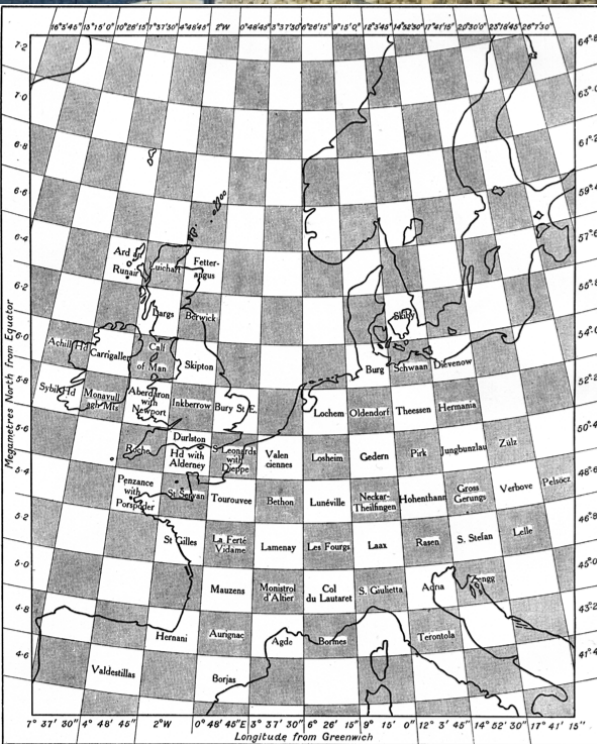


L.F. Richardson, Weather Prediction by Numerical Process (1922):

The forecast factory



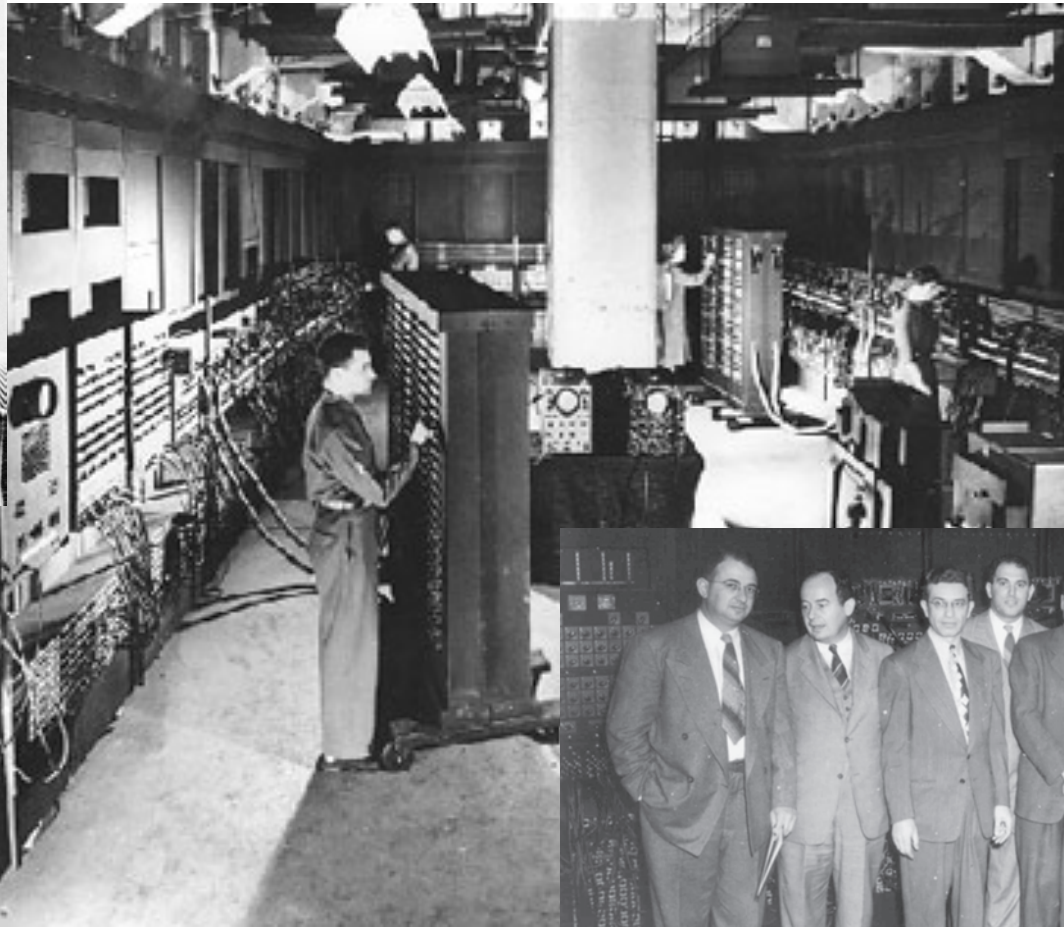
L. F. Richardson, 1931



“Imagine a large hall like a theater except that the circles and galleries go right round through the space usually occupied by the stage. The walls of this chamber are painted to form a map of the globe. . . . From the floor of the pit a tall pillar rises to half the height of the hall. It carries a large pulpit on its top. In this sits the man in charge of the whole theatre.” (Weather Prediction by Numerical Process)



Jule Charney
(1917-81)



ENIAC

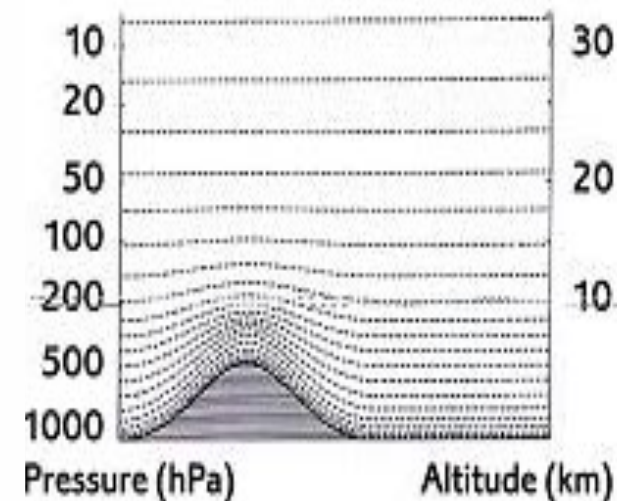
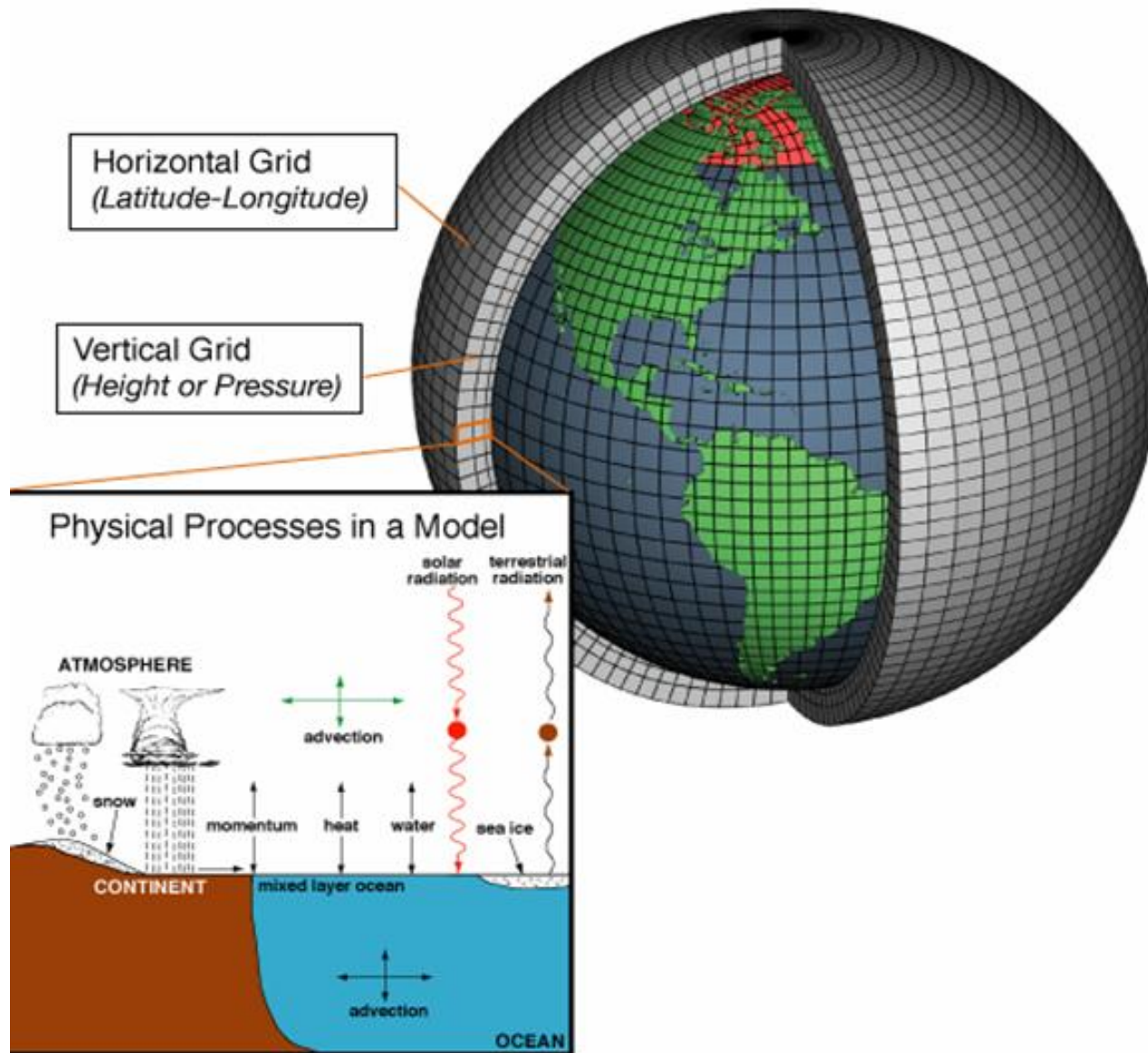
500 hPa geopotential
height: solid=observed
dashed=forecasted
change



FIG. 1. Visitors and some participants in the 1950 ENIAC computations. (left to right) Harry Wexler, John von Neumann, M. H. Frankel, Jerome Namias, John Freeman, Ragnar Fjørtoft, Francis Reichelderfer, and Jule Charney. (Provided by MIT Museum.)

1950: First weather forecast for 24h using the first electronic computer (ENIAC) and **simplified equations for the atmosphere (QG)**

General circulation models



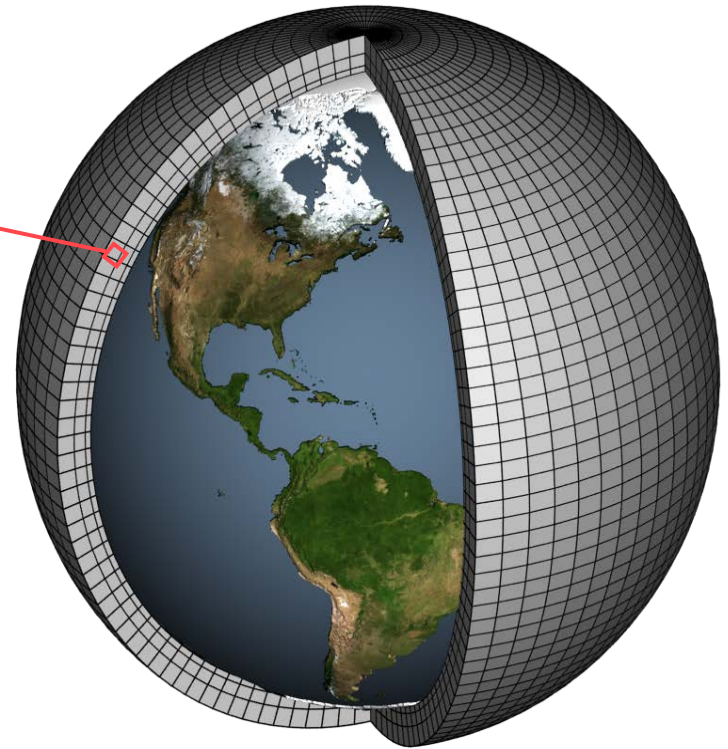
Mathematical equations that represent the physical characteristics and processes are entered for each box

Primitive equations (3D):

Hydrostatic approximation
Boussinesq approximation

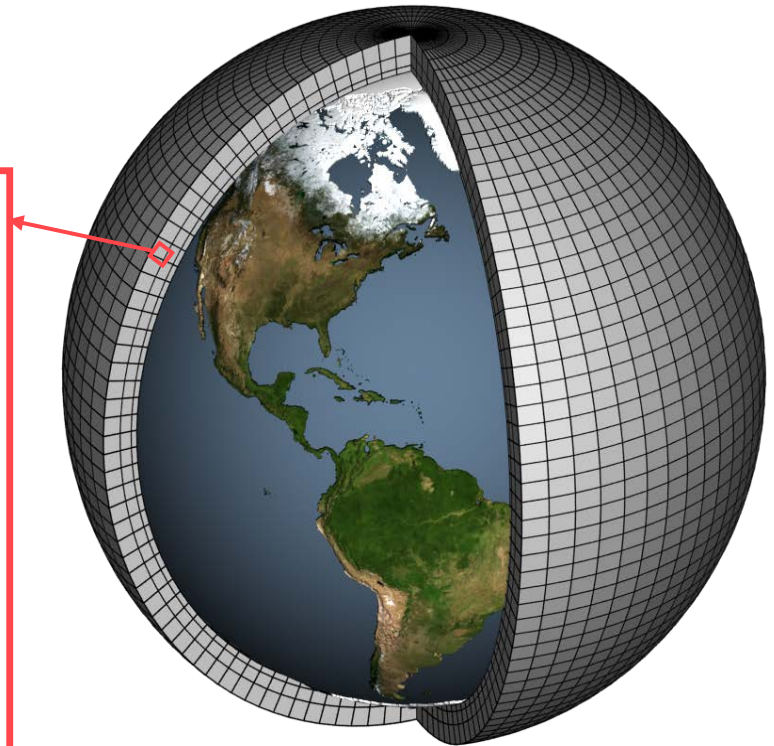
Vertical coordinate:
pressure, or entropy

Perfect gas (atmosphere)
Thermodynamics



Equations are converted to computer code and climate variables are set

```
if (diagts .and. eots) then
  do 1500 m=1,nt
    do 1490 k=1,km
      fx = cst(j)*dyt(j)*dzt(k)/(c2dttts*dtxccl(k))
      do 1480 i=2,imtml
        boxfx      = fx*dxt(i)*fm(i,k,jc)
        sddt       = (ta(i,k,m)-t(i,k,jc,nm,m))*boxfx
        svar       = (ta(i,k,m)**2-t(i,k,jc,nm,m)**2)
                  *boxfx
        n          = 0
        termbt(k,1,m,n) = termbt(k,1,m,n) + sddt
        tvar(k,m,n)     = tvar(k,m,n) + svar
        n             = nhreg*(mskvr(k)-1) + mskhr(i,j)
        if (n .gt. 0 .and. mskhr(i,j) .gt. 0) then
          termbt(k,1,m,n) = termbt(k,1,m,n) + sddt
          tvar(k,m,n)     = tvar(k,m,n) + svar
```



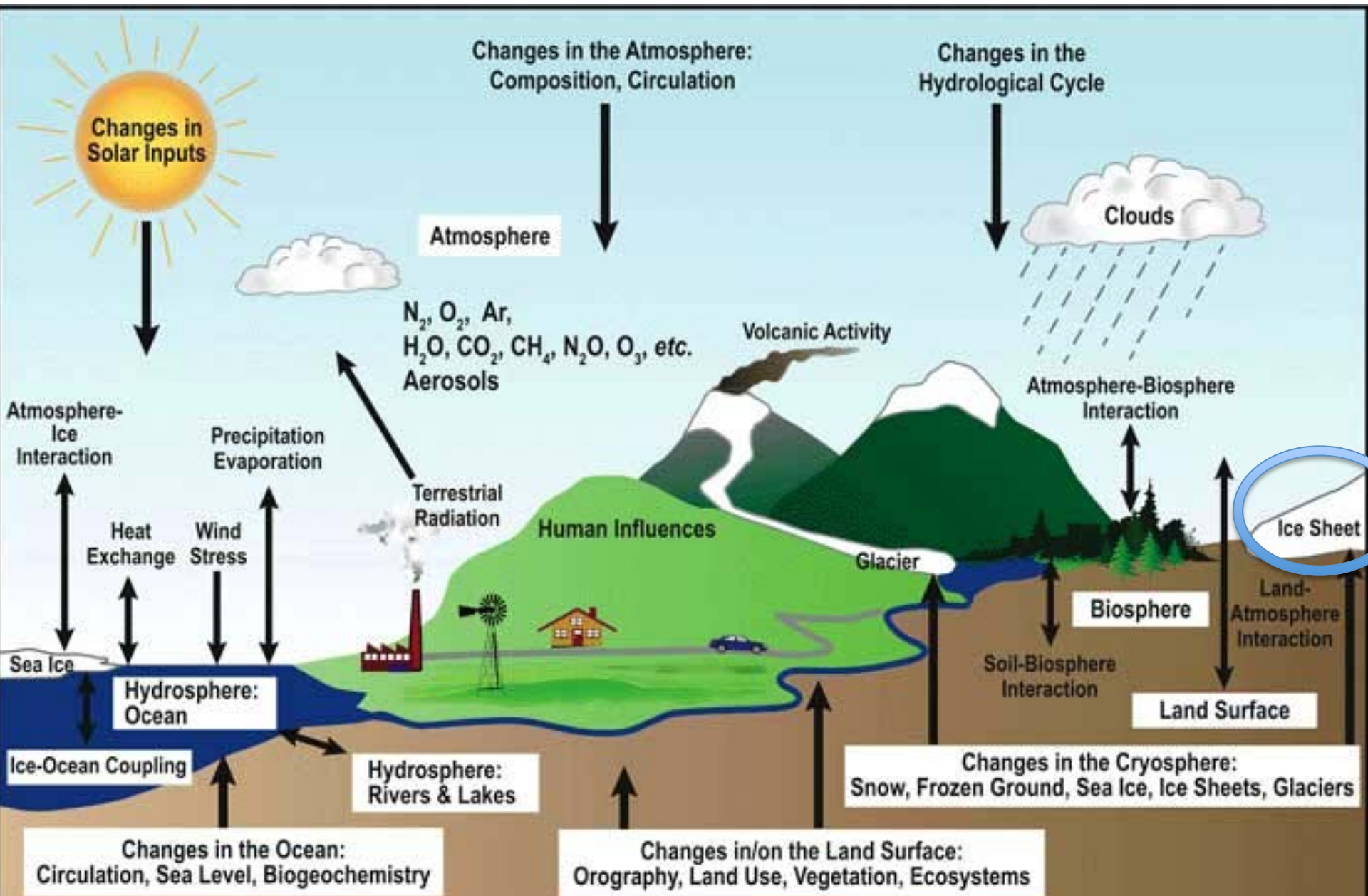
From meteorology to climate:

There is much more than the atmosphere:
Earth System processes

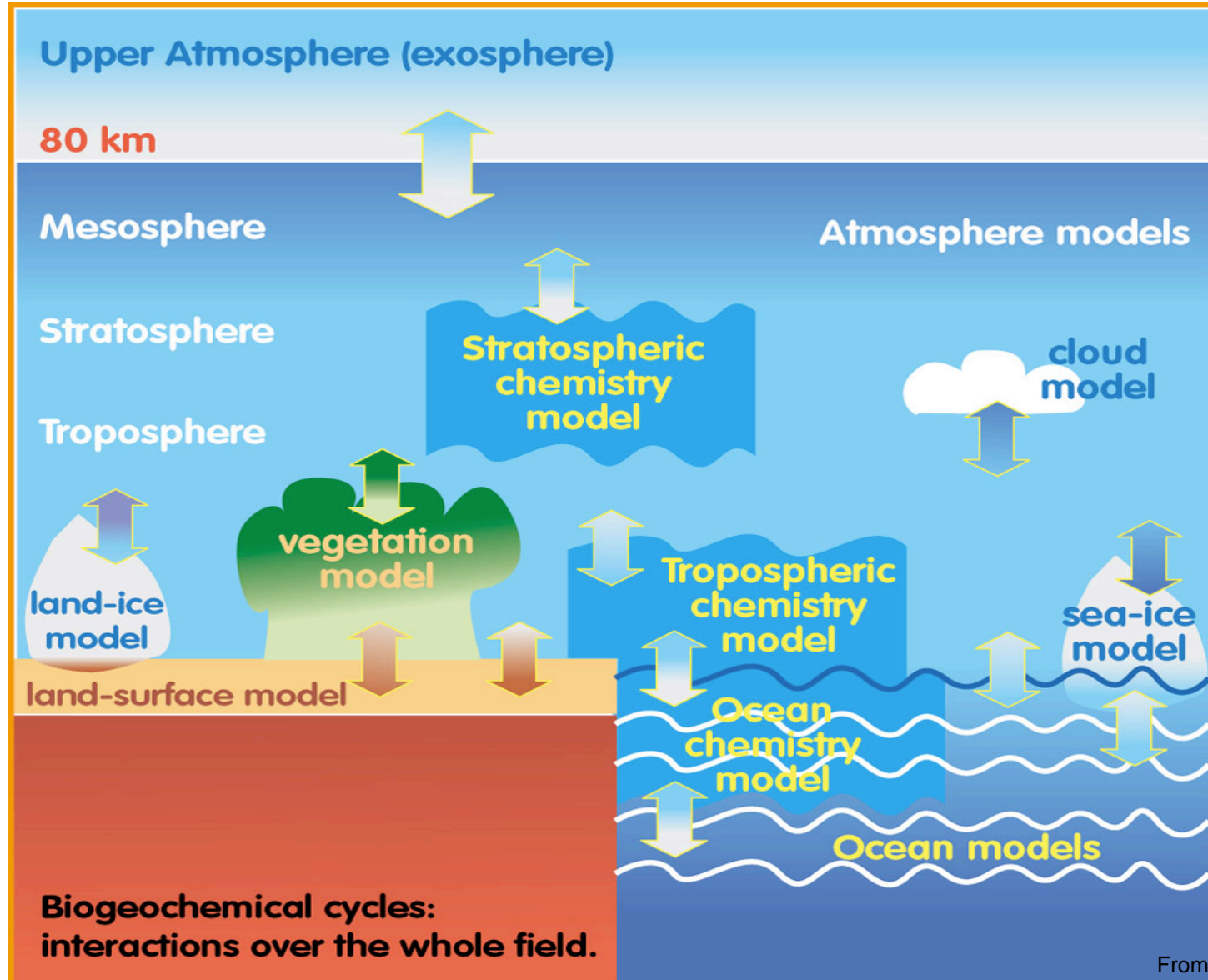
What are the climatic processes
that we need to take into account?

It depends, of course, on the time scale

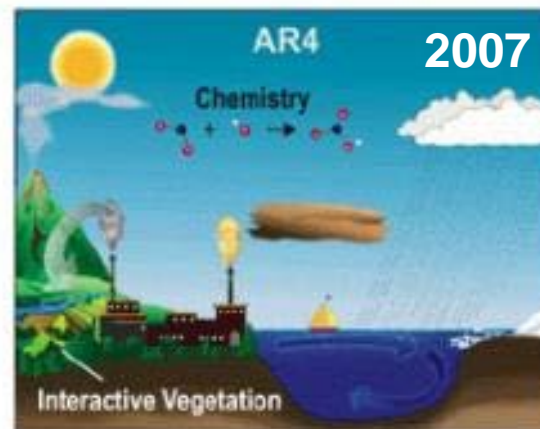
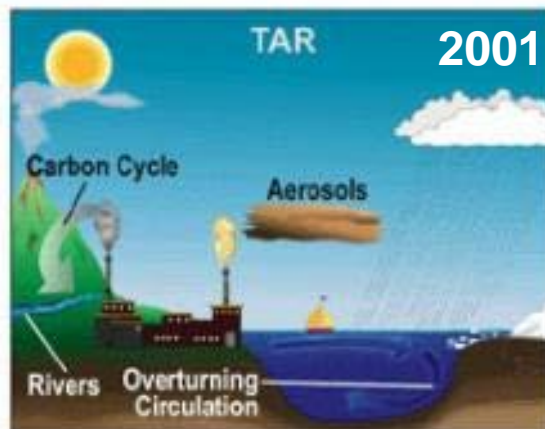
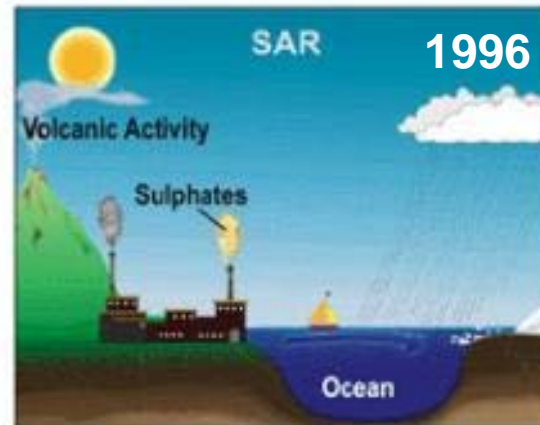
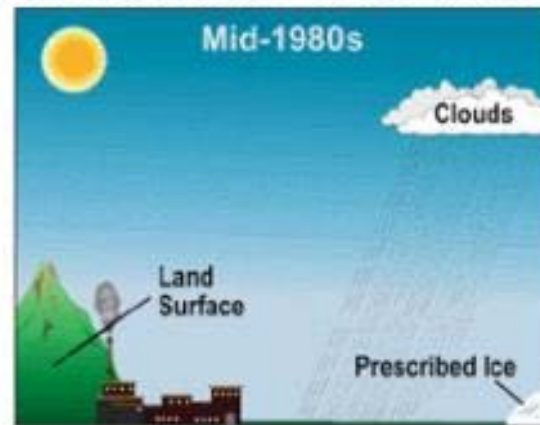
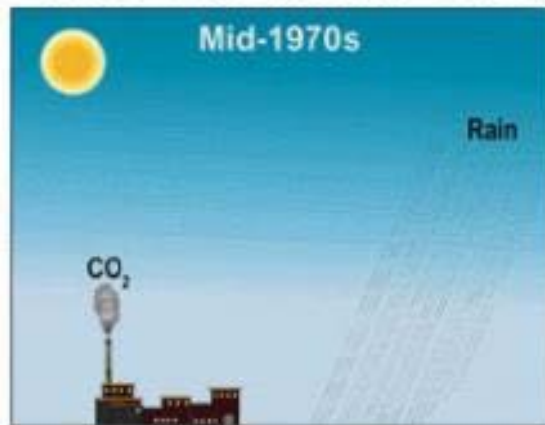
Earth System processes (100 yr)



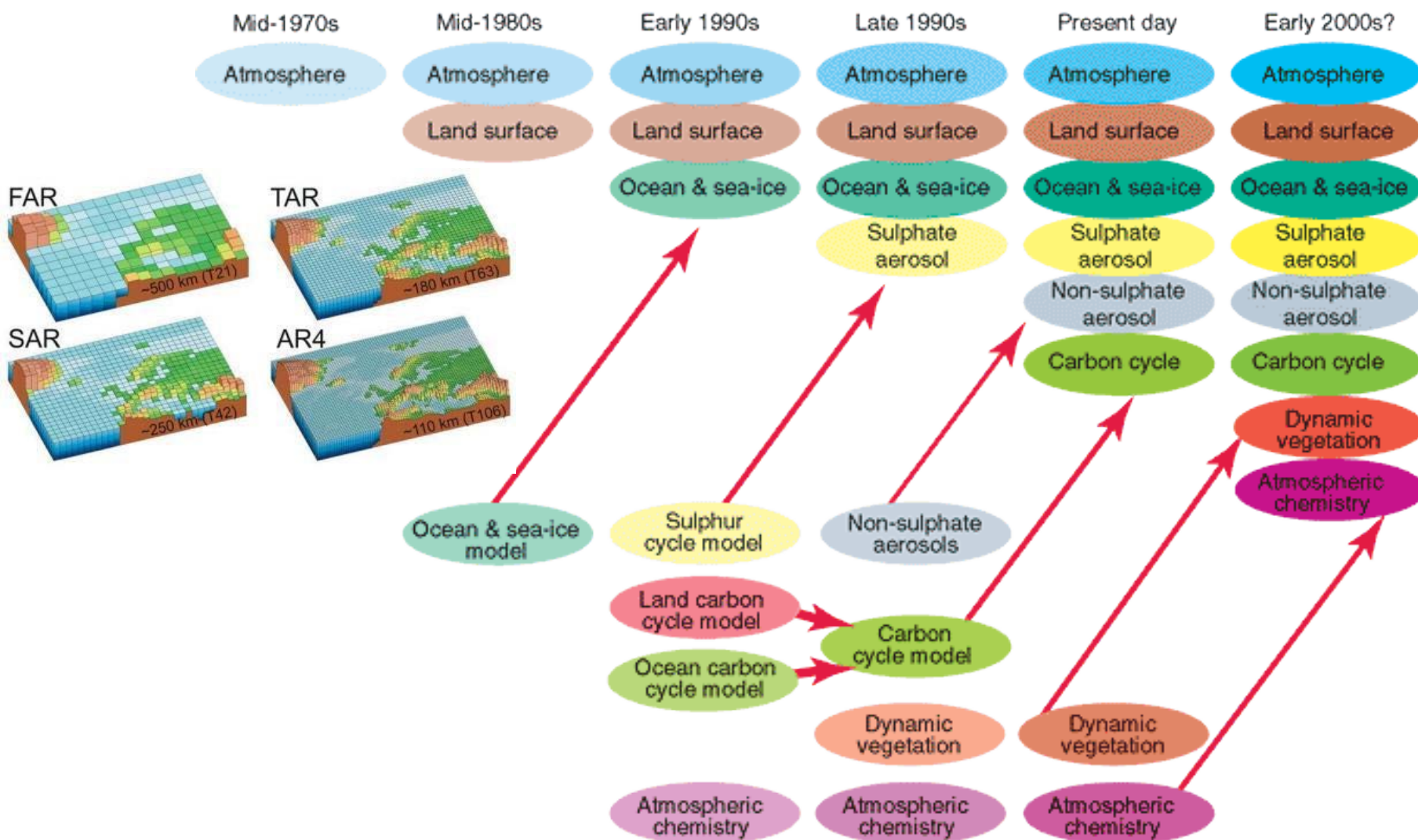
Main components of a global Earth System model



The World in Global Climate Models



The Development of Climate models, Past, Present and Future



But remember that sometimes “Less is more”

IPCC TAR, 2001

Other crucial elements of climate modelling:

External forcings

Solar variability

Orbital variability

Volcanoes

GHG concentrations

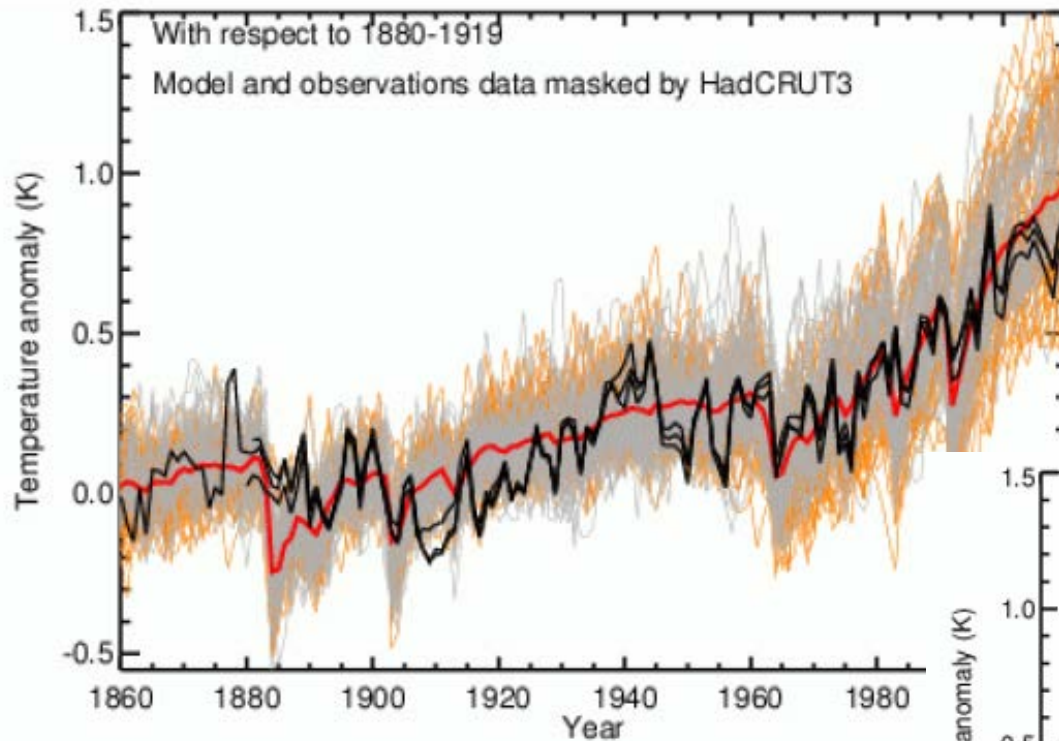
Aerosols

(often, use equivalent radiative forcing at TOA)

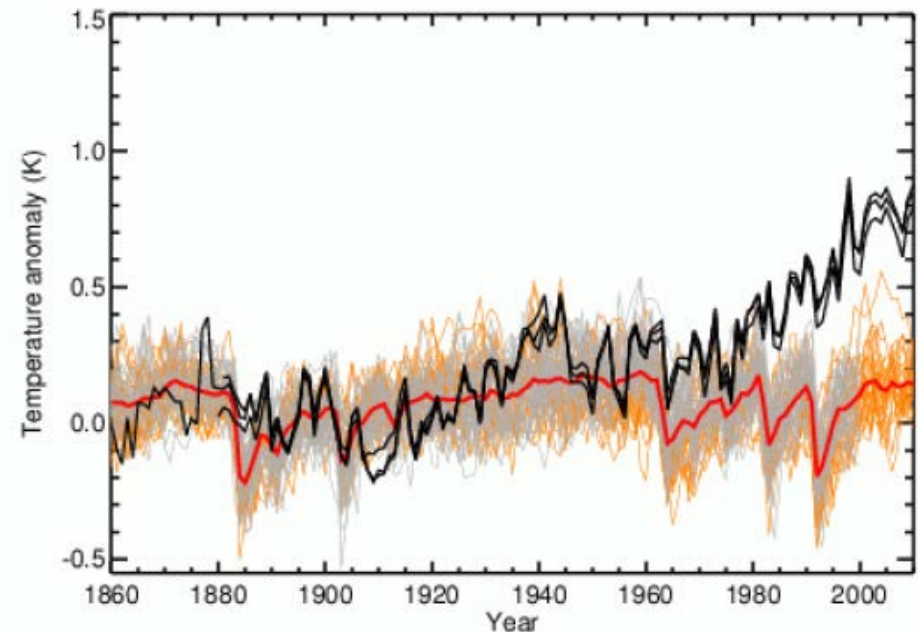
Initial conditions

Parameterization choices

Model validation: reproduction of current climate



— CMIP3
— CMIP5



IPCC 2013

Model validation: Global precipitation

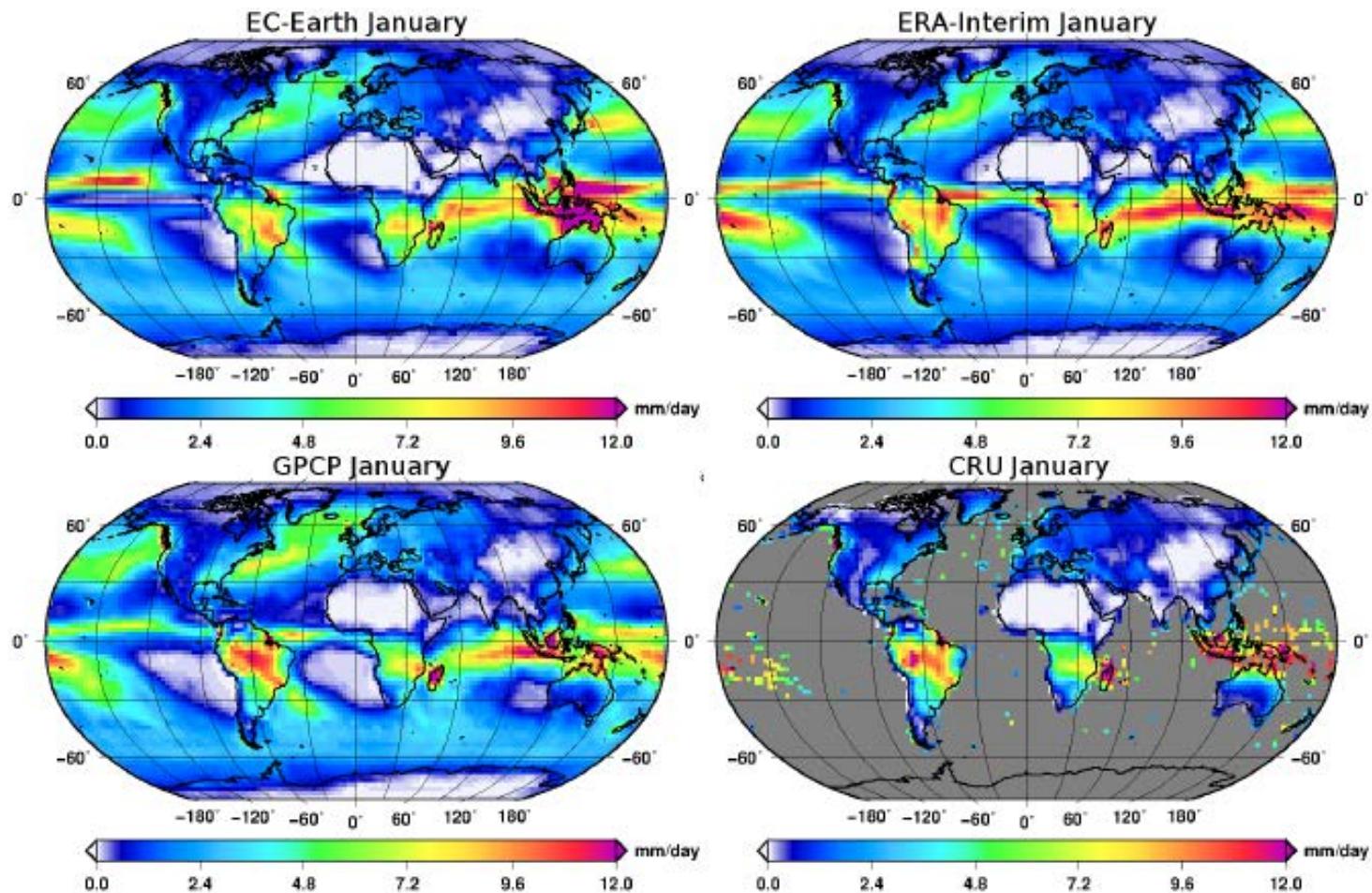
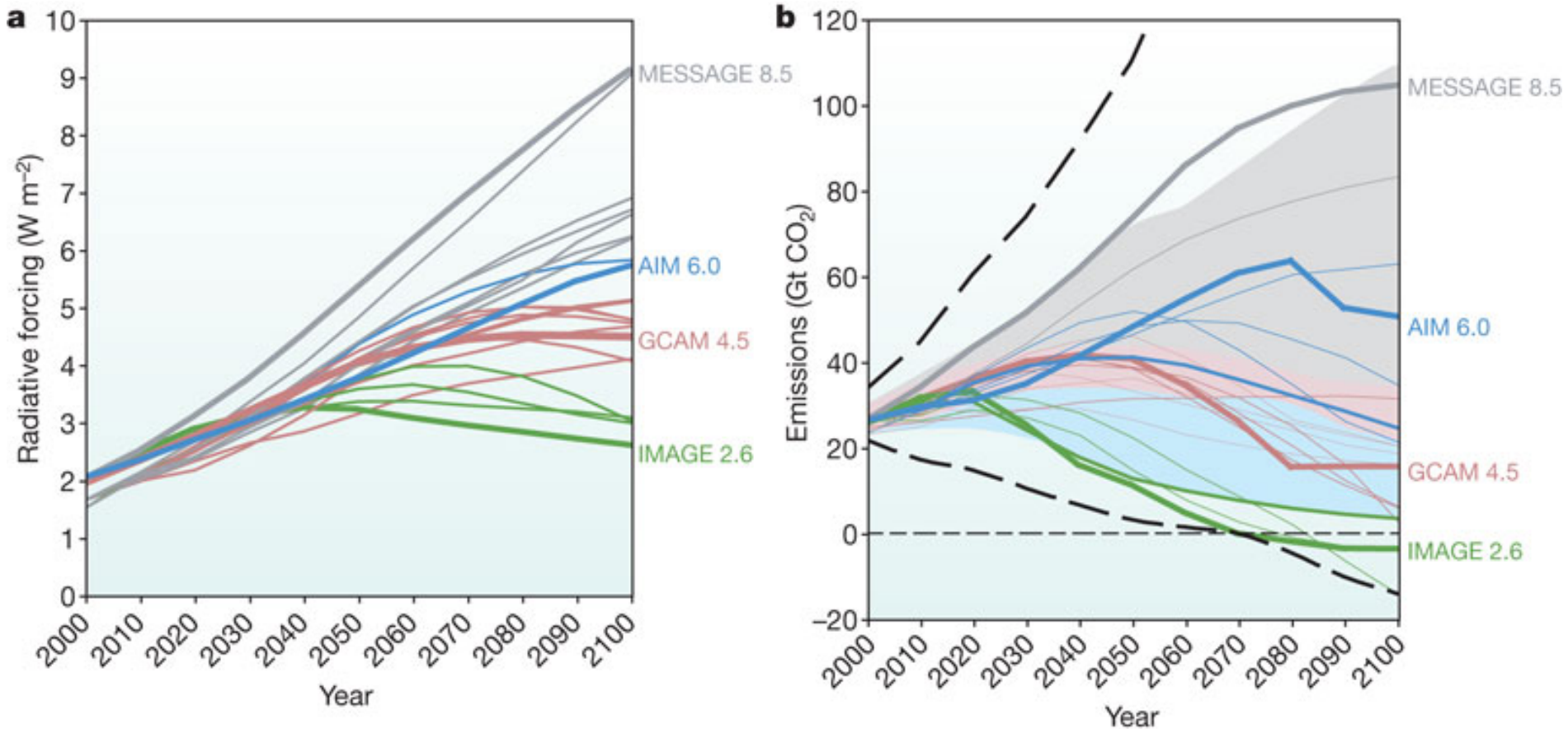


Figure 1: Multiannual mean (1980-2005) January precipitation from EC-Earth, ERA-Interim, GPCP and CRU.

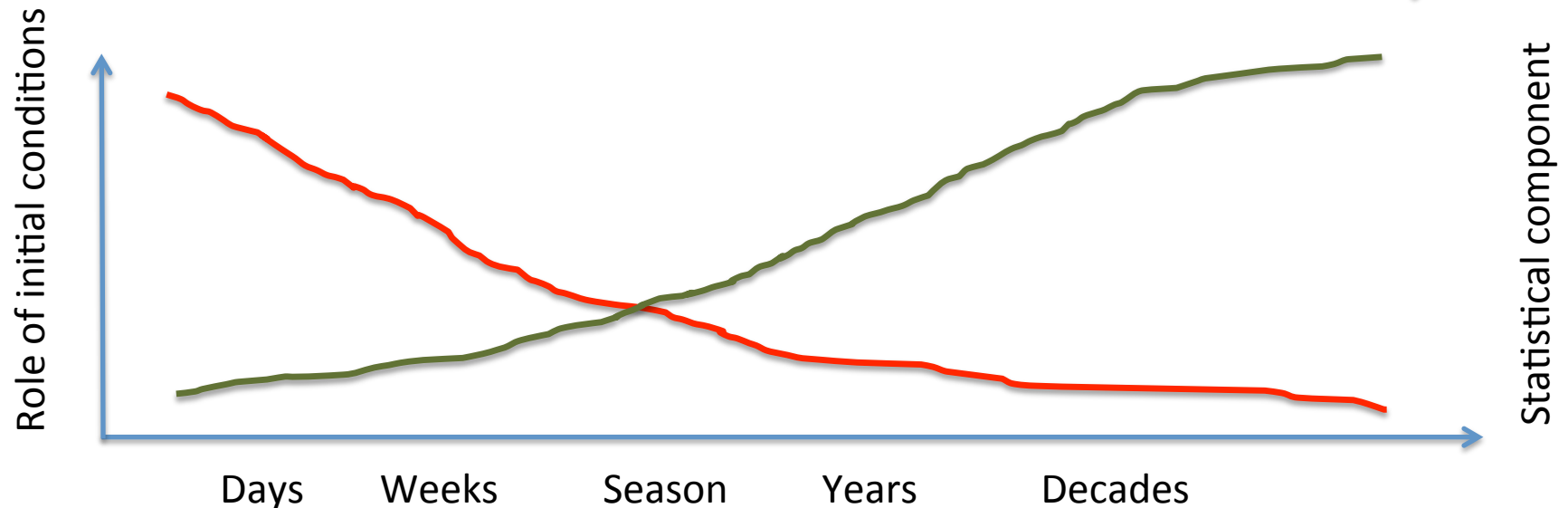
Representative concentration pathways



The difference between weather and climate

Predictions of the first and second kind (Edward Lorenz)

What is climate predictability?
(position on the attractor vs
invariant measure on the attractor...)



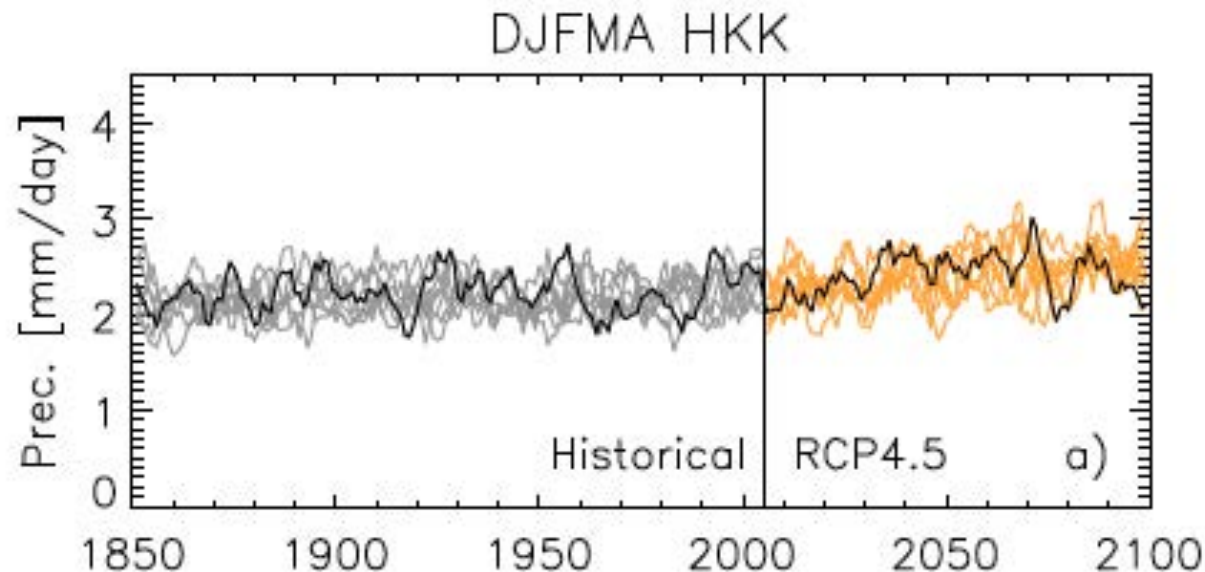
Climate simulations are statistical

A given year in a climate simulation
does not mean anything

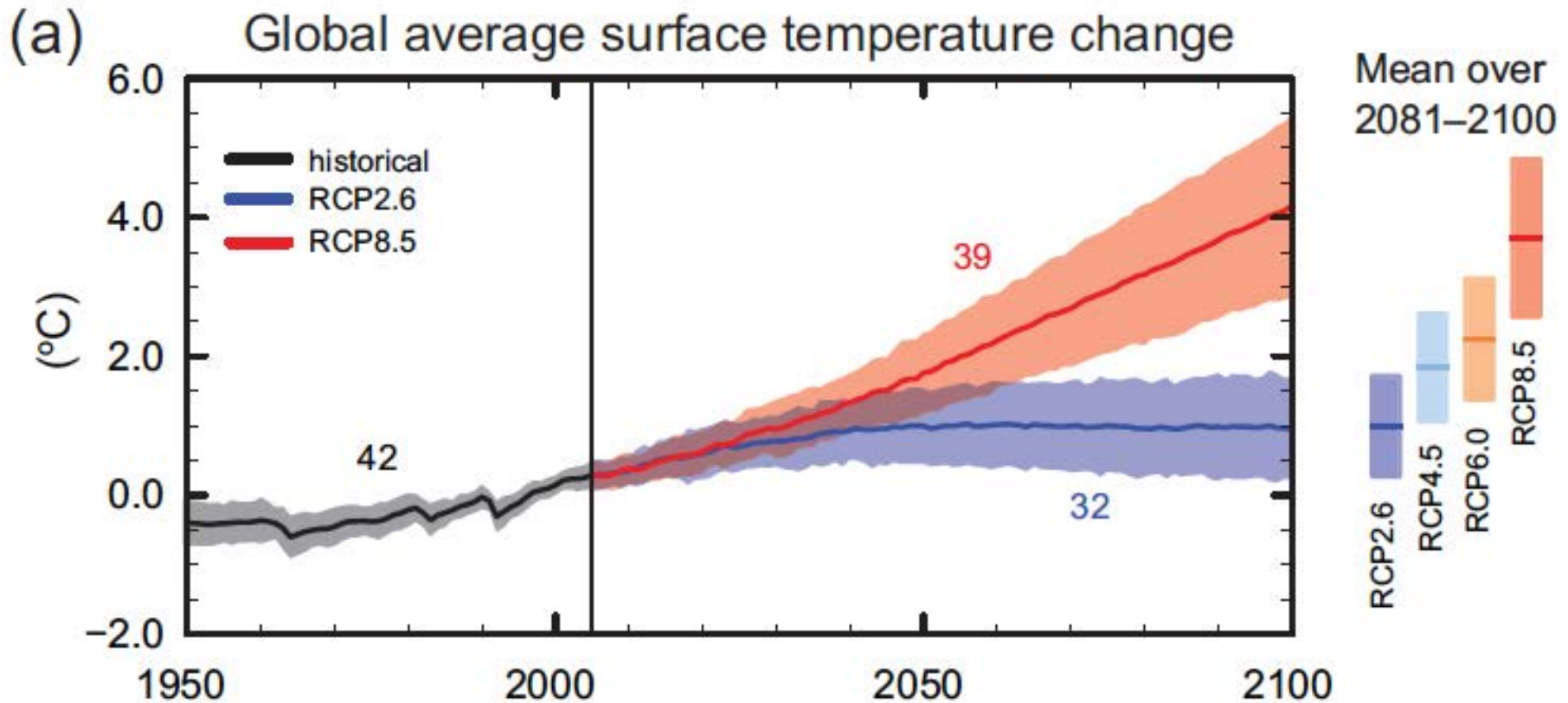
Only trends and statistical quantities
(PDFs) are meaningful

Climate ensemble predictions

Start from different initial conditions
and generate an ensemble of simulations
with the same model and same parameters/forcing
or with different models (“multimodel superensemble”)
and/or with different parameter choices

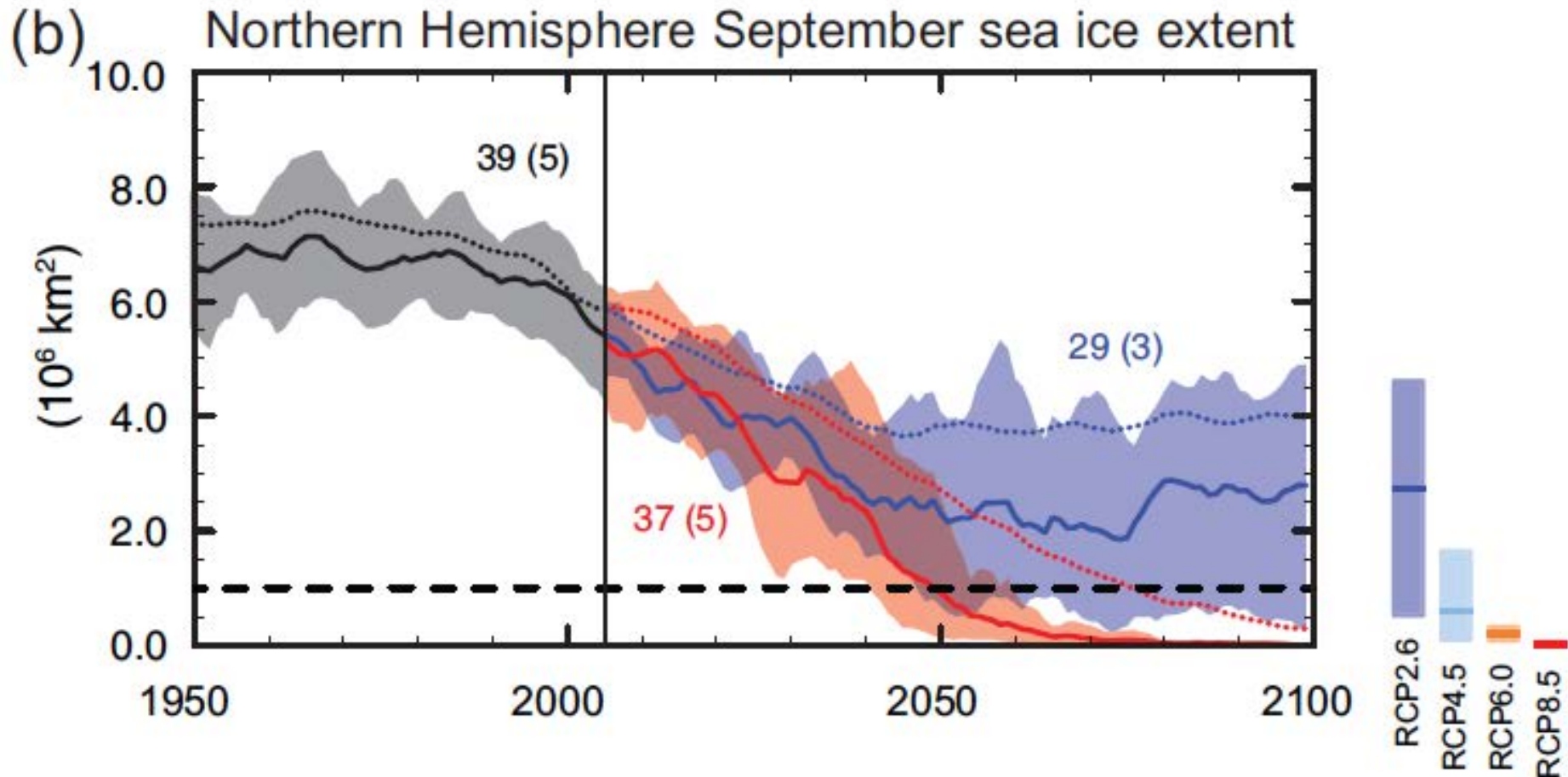


Simulations of future climate

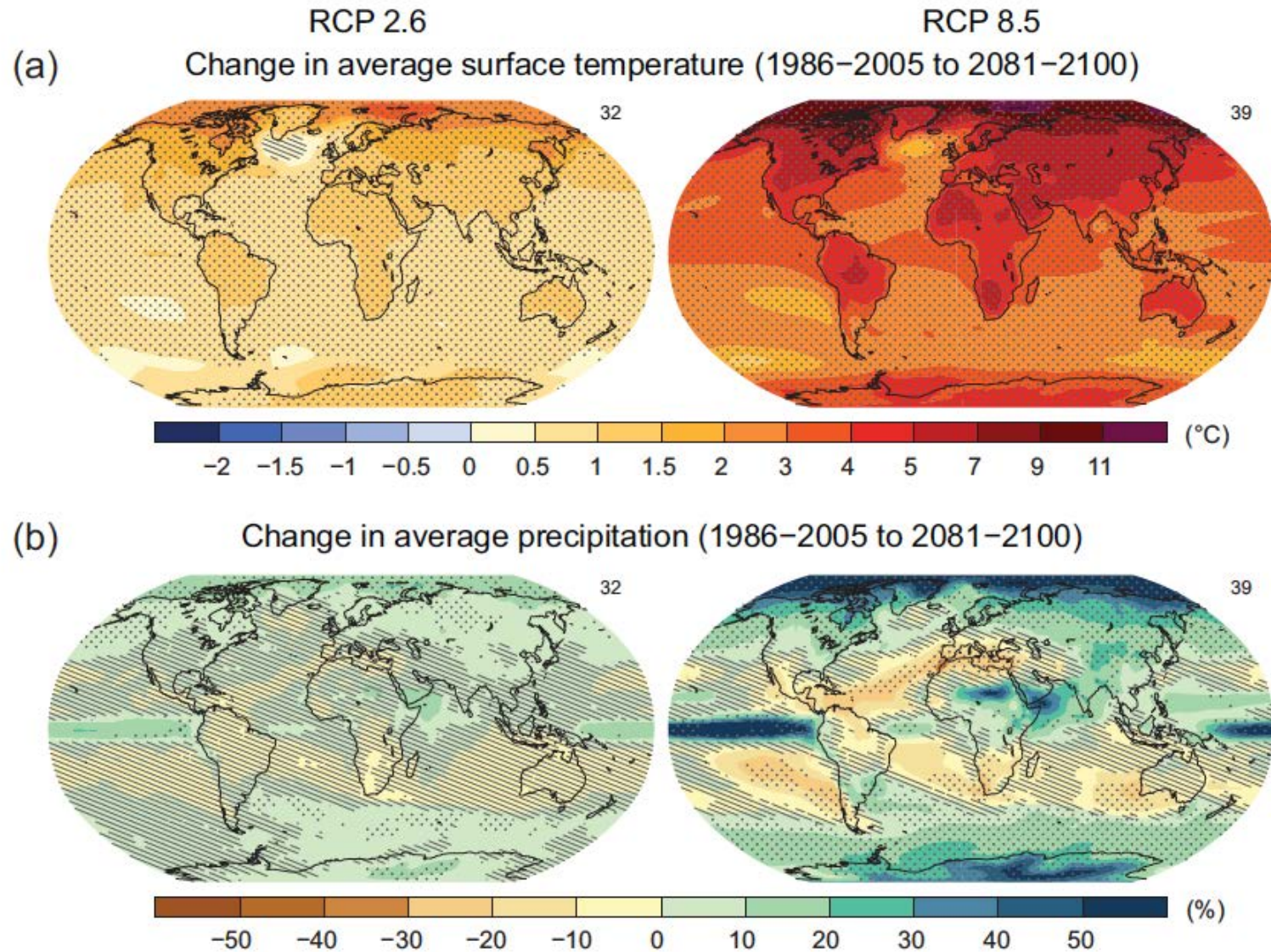


IPCC 2013

Simulations of future climate



Simulations of future climate



The concept of seamless predictions

- Weather and Climate: Same physical processes (but acting on different space and time scales)
- Initial conditions vs boundary conditions (predictability of the first or second kind)
- From weather → to seasonal → to decadal predictions
- Advantages: climate models profit from advances in NWP and vice-versa

Ref.: Hazeleger, W. et al., 2009. EC-Earth: A Seamless Earth System Prediction Approach in Action. *Bull. Amer. Meteor. Soc.*, in press.



A European Earth-System-Model for climate studies

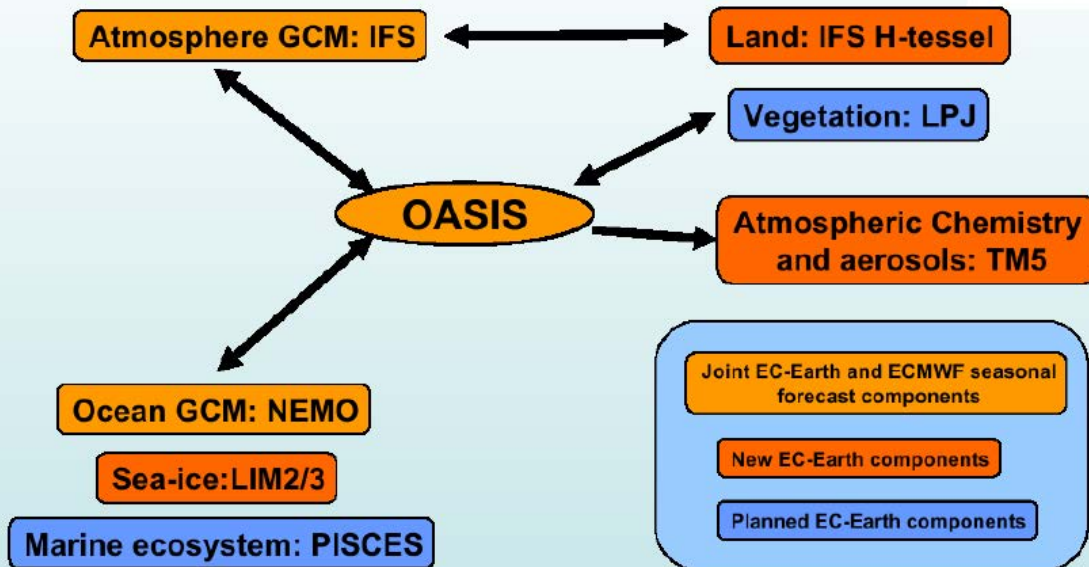


The EC-Earth Model

Based on the idea of “seamless predictions”

ECMWF IFS atmosphere (31r1 - T159L62/N80)+ Land/veg module
+ NEMO2 ocean (OPA/ORCA1) (1° L32)
+ TM5 chemistry/aerosols ($6^\circ \times 4^\circ$ / $3^\circ \times 2^\circ$)

EC-EARTH components



Integrated Forecast System
ECMWF



Nucleus for European
Modelling of the Ocean

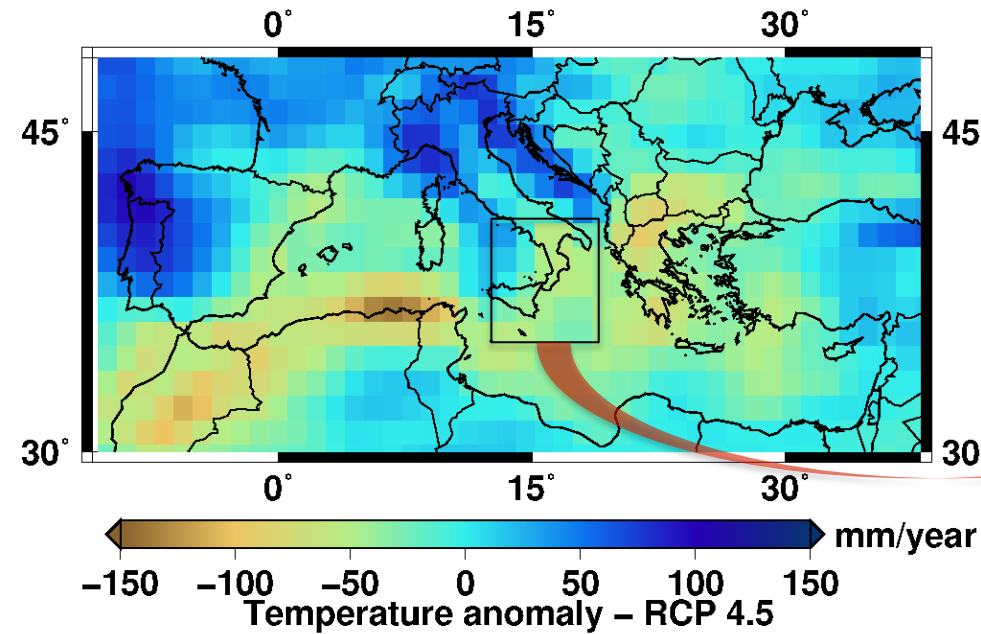


TM5 atmospheric chemistry
and transport model

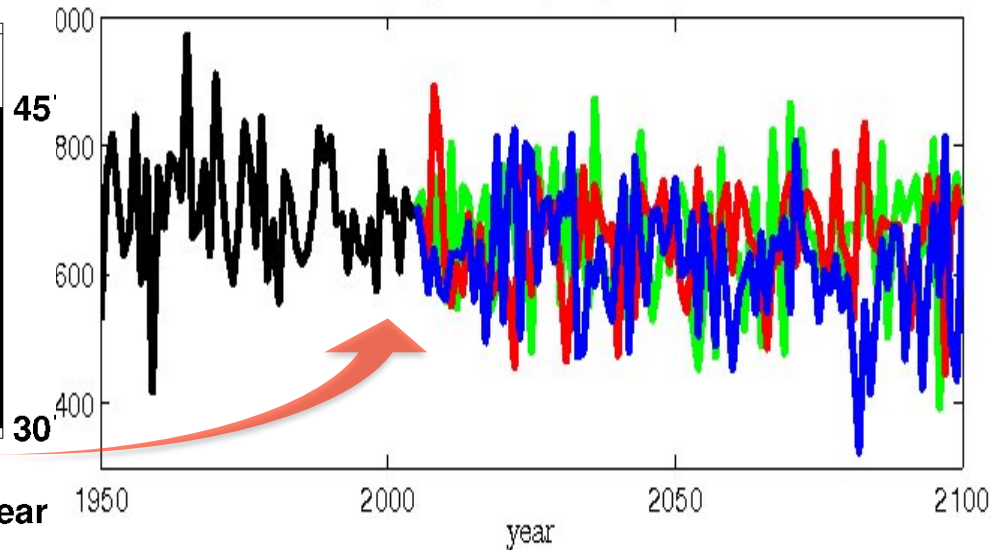
Ref.: Hazeleger, W. et al., 2009. EC-Earth: A Seamless Earth System Prediction Approach in Action. *Bull. Amer. Meteor. Soc.*, in press.

EC-Earth Model, RCP 4.5: [2041-2060] – [1986-2005]

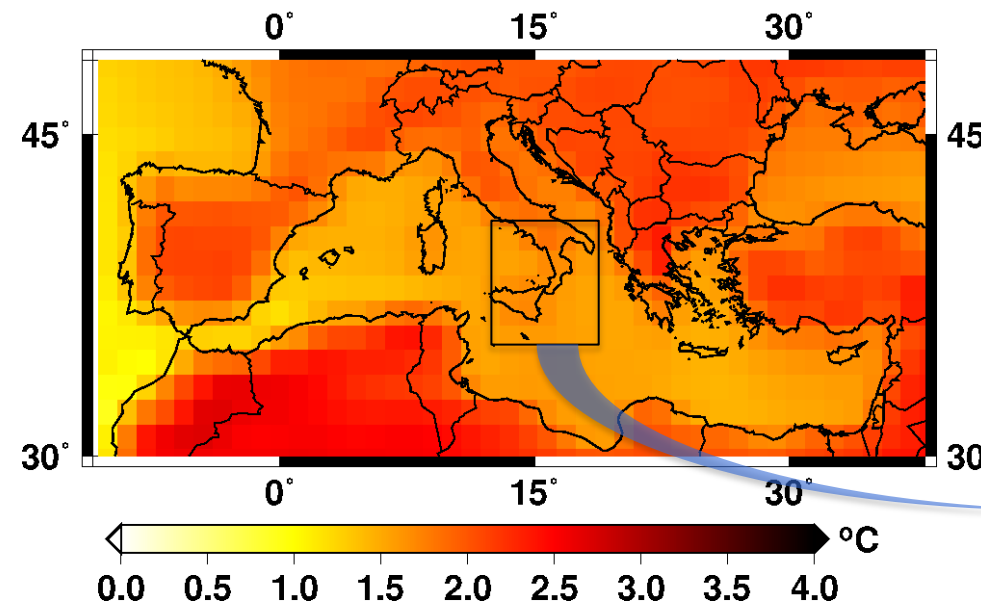
Precipitation anomaly – RCP 4.5



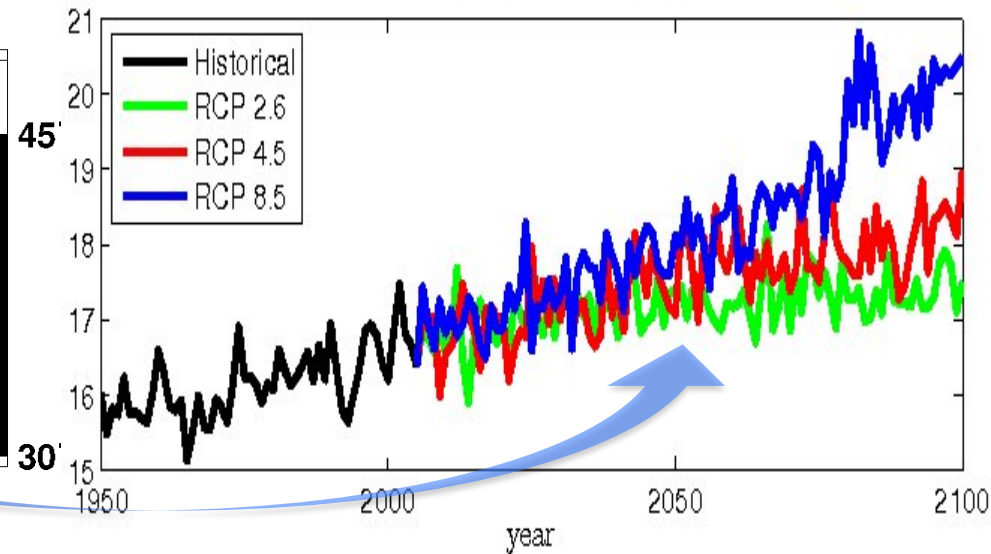
Average annual precipitation

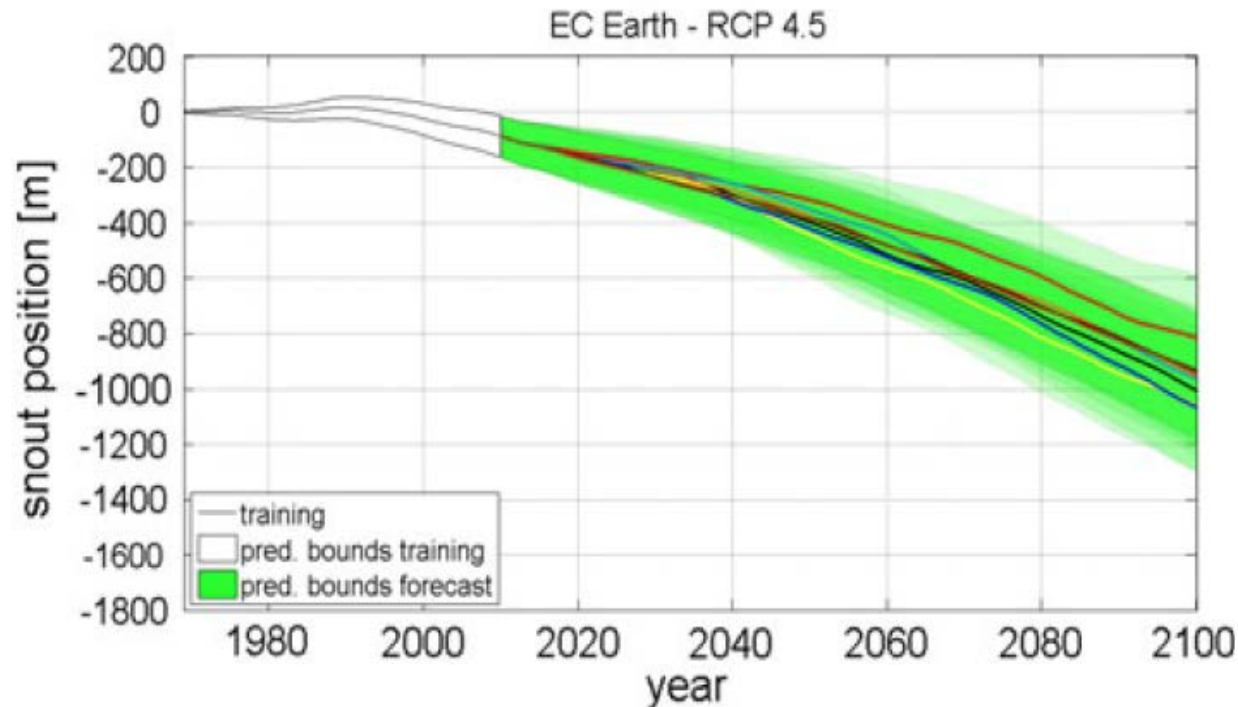
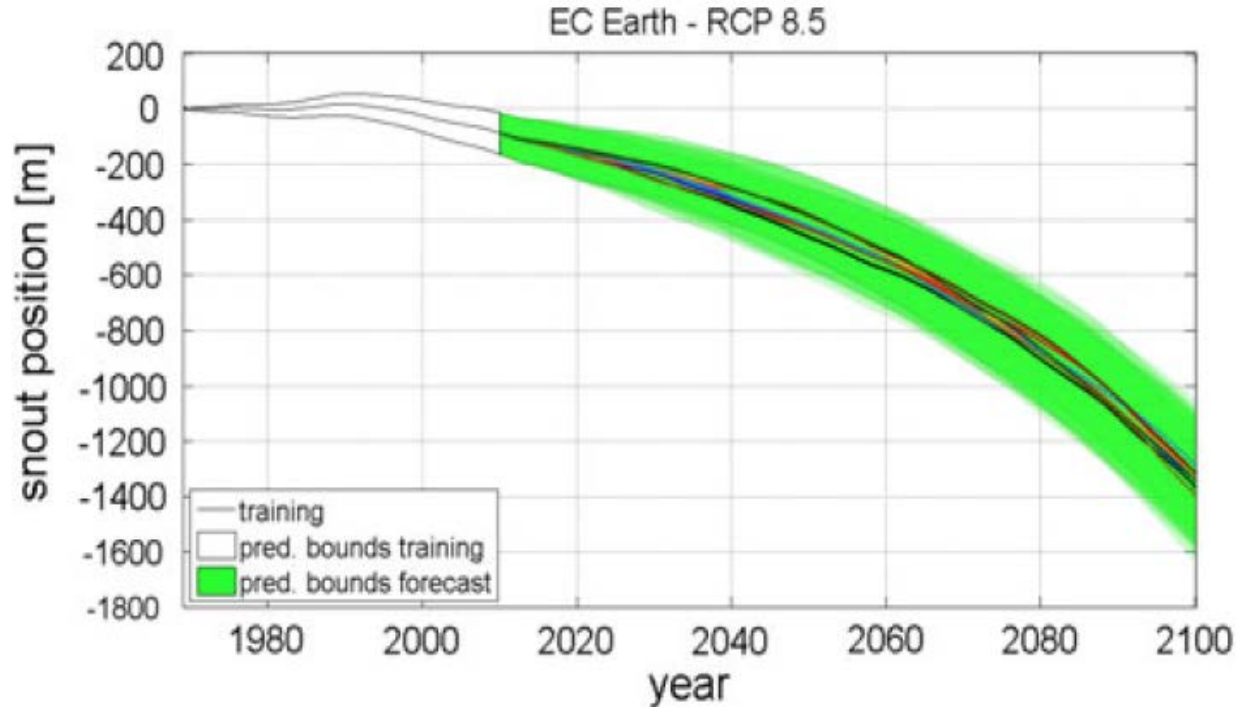


Temperature anomaly – RCP 4.5



Average annual temperature

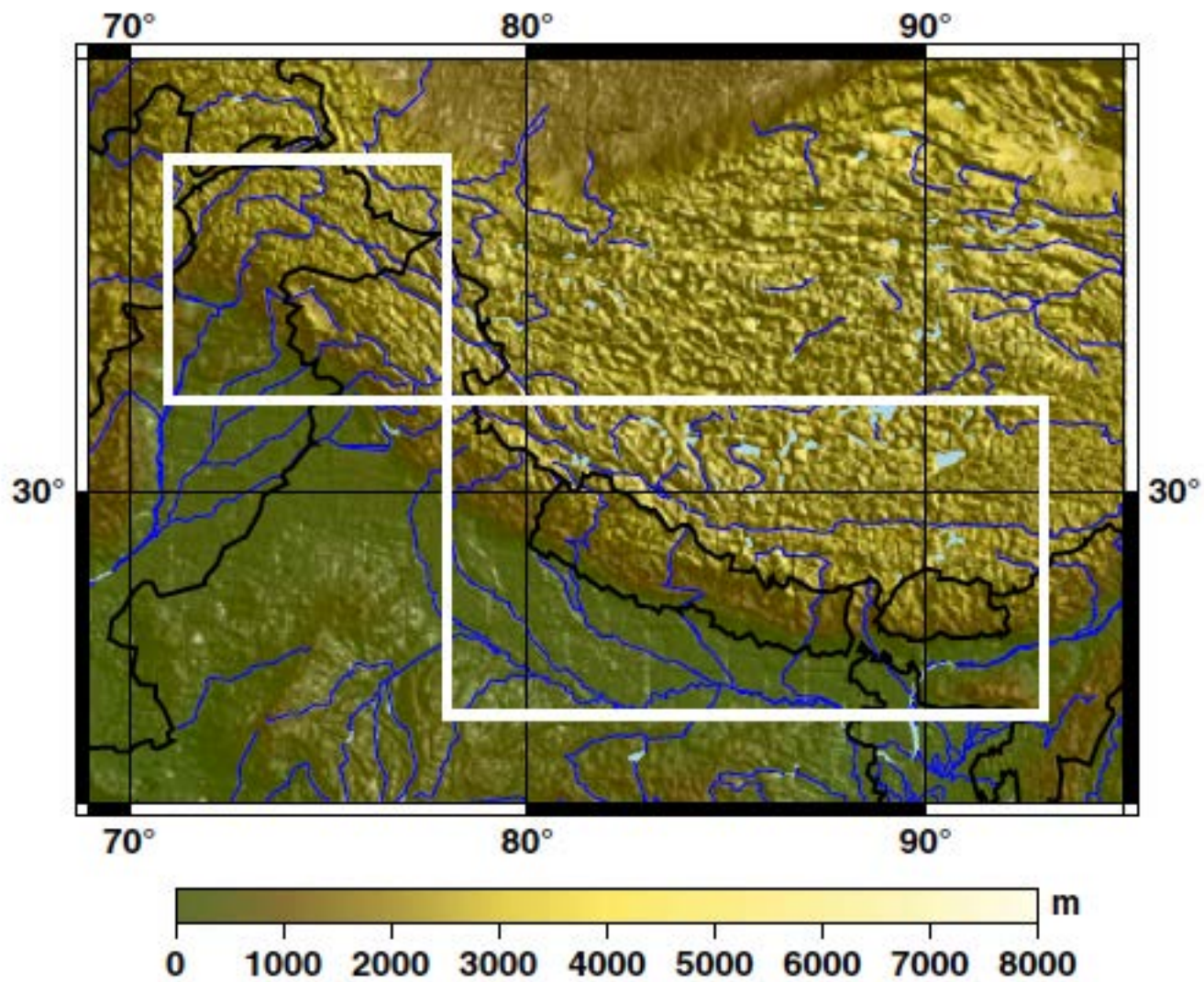


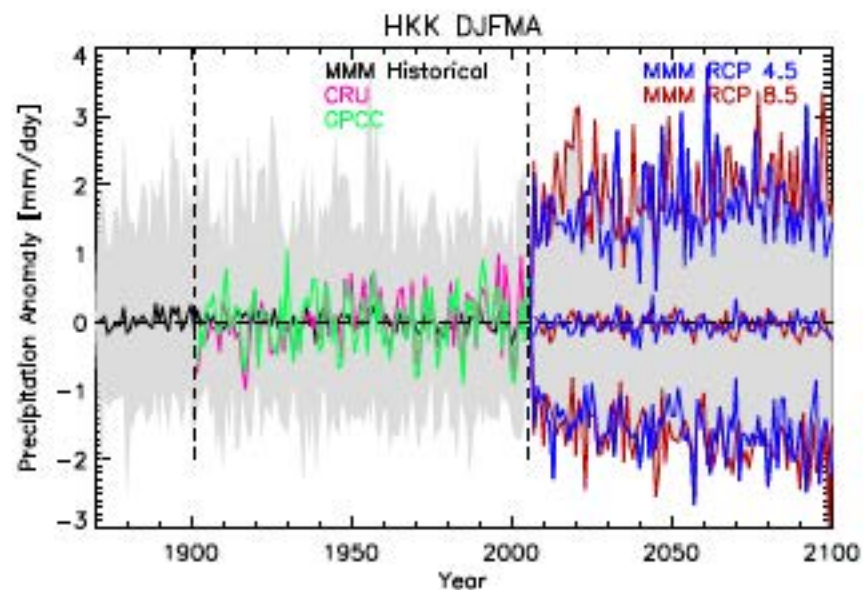
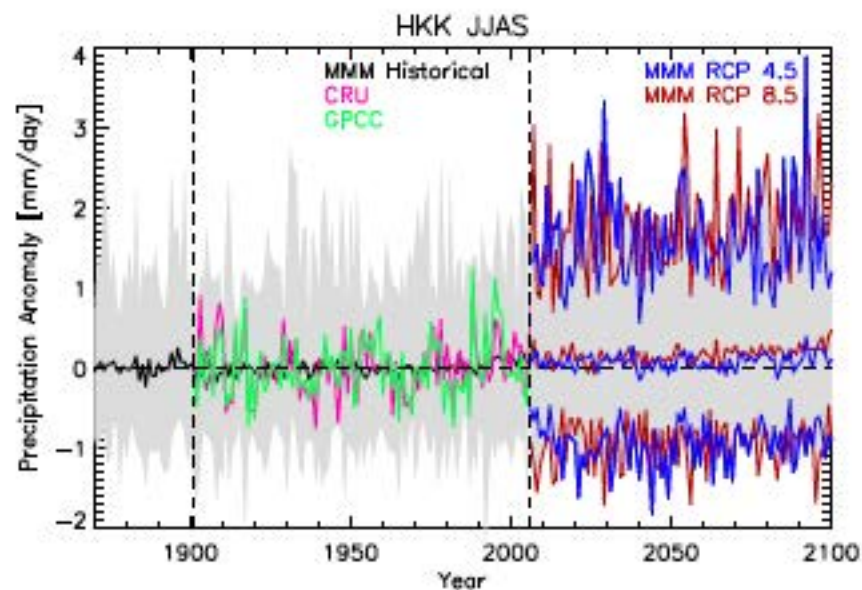
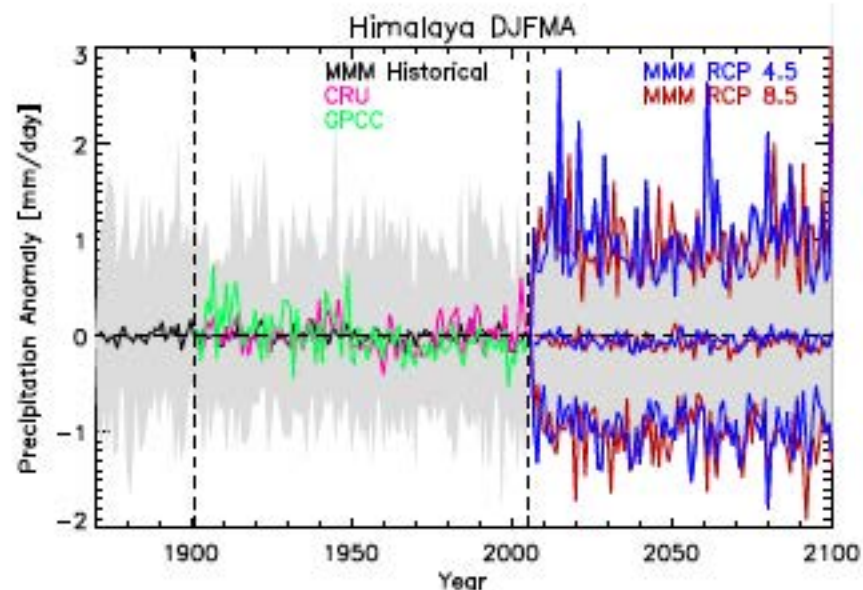
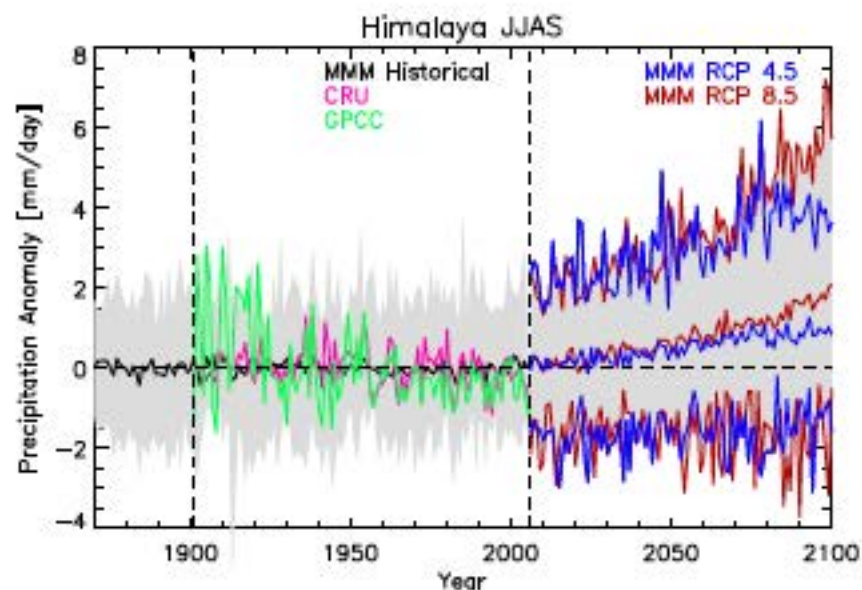


Alpine glacier dynamics (glacier ensembles)

Glacier retreat

Bonanno et al. 2012





CLIMATE SERVICES: ERA-NET ERA4CS

“Supporting research for developing better tools, methods and standards on how to produce, transfer, communicate and use reliable climate information to cope with current and future climate variability”

**Focus on seasonal to multiannual time scales:
Importance of initial conditions**

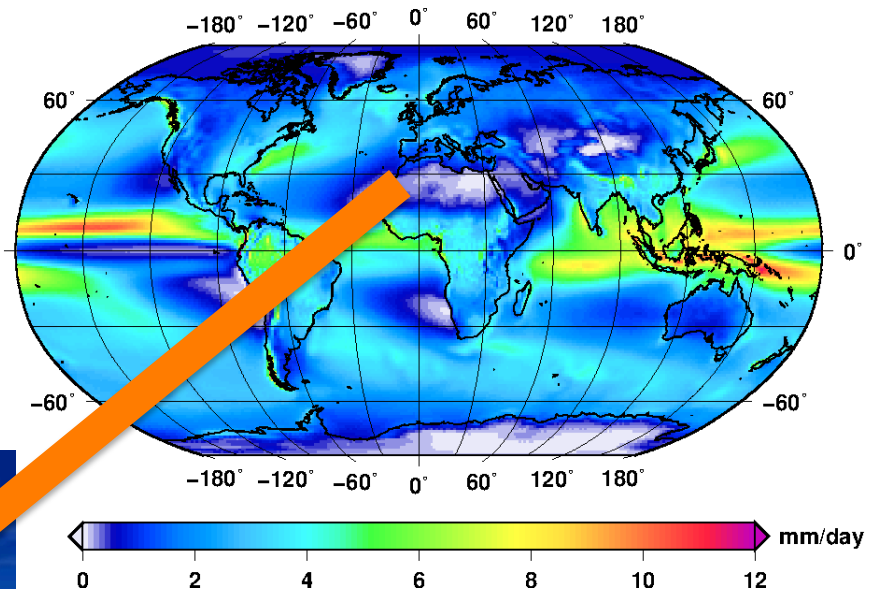
<http://www.jpi-climate.eu/ERA4CS>

To estimate future impacts and risks, we need climate and impact models

Global Climate Models: The most advanced tools that are currently available for simulating the global climate system and its response to anthropogenic and natural forcings.



Total precipitation annual mean 1951–2007



Impact models:
Basin response
Ecosystems
Glaciers and snow
Agriculture, Land surface
Water resources

Problem:

**Most climate change impacts
take place at local scale**

**Global Climate Models
currently provide climate projections
spatial resolution between 40 and 100 km**

**So: scale mismatch and
need for climate downscaling**

Climate downscaling approaches:

Dynamical downscaling

Regional Climate Models

(eg RegCM, Protheus)

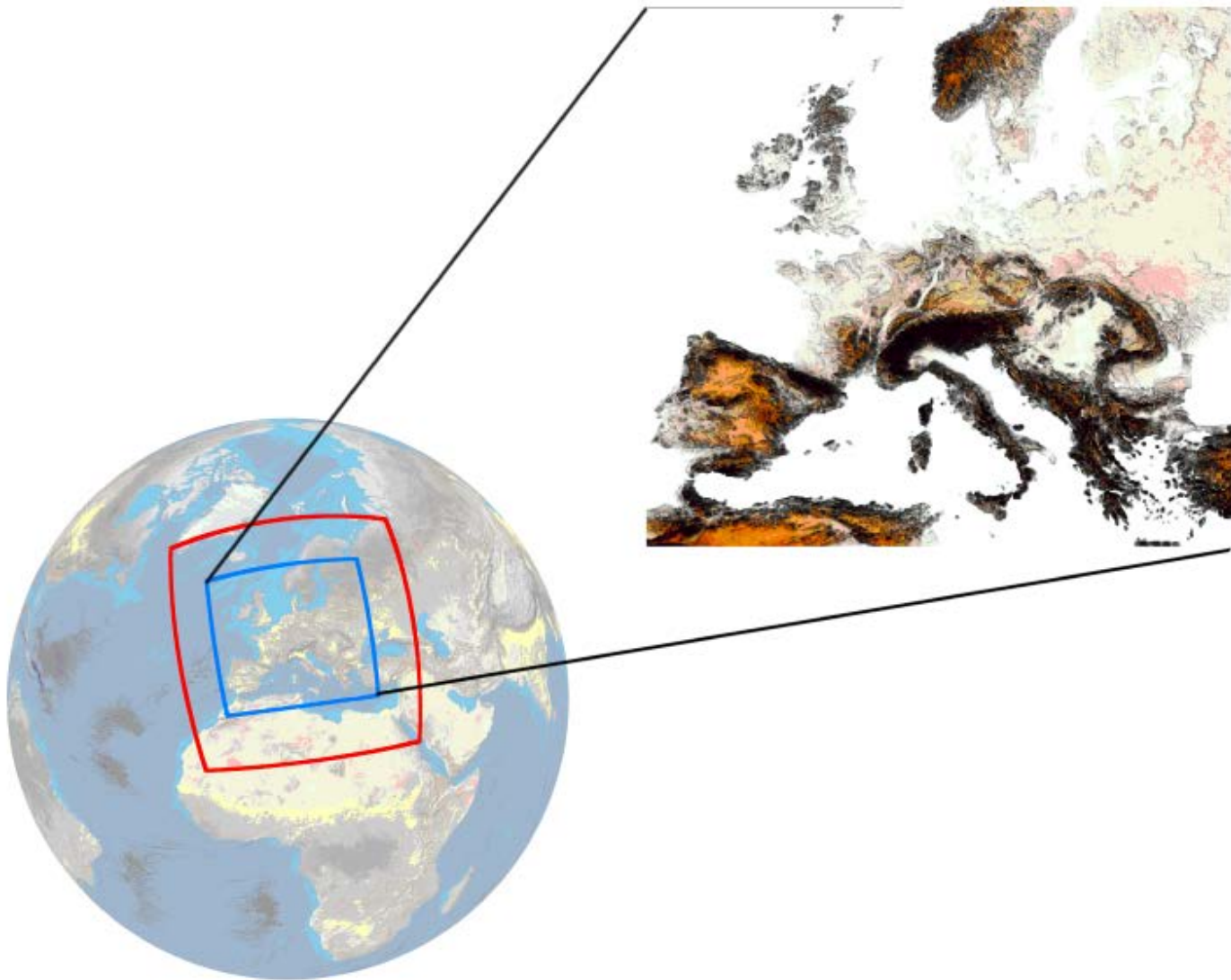
Non-hydrostatic models

(eg COSMO-CLM, WRF)

Statistical downscaling

Stochastic (rainfall) downscaling

Dynamical downscaling



Non-hydrostatic RCMs: simulations with WRF

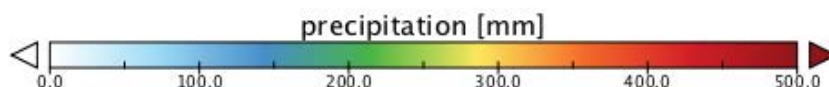
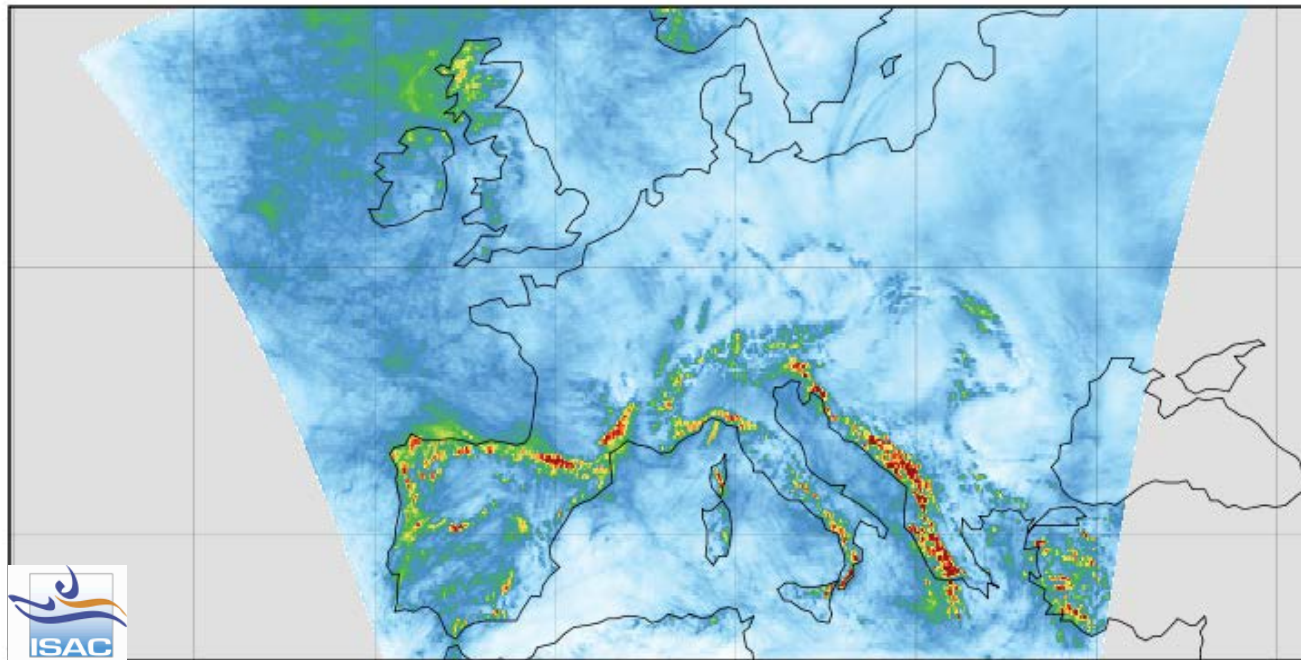


WRF - Weather Research & Forecasting Model

<http://www.wrf-model.org/index.php>

**Climate simulations (30 years) with WRF at high spatial resolution (0.11° and 0.04°)
nested into reanalyses (to be nested also into the EC-Earth GCM)**

Precipitation January 1979



Total precipitation

**from WRF climate
simulations**

at 4 km

January 1979

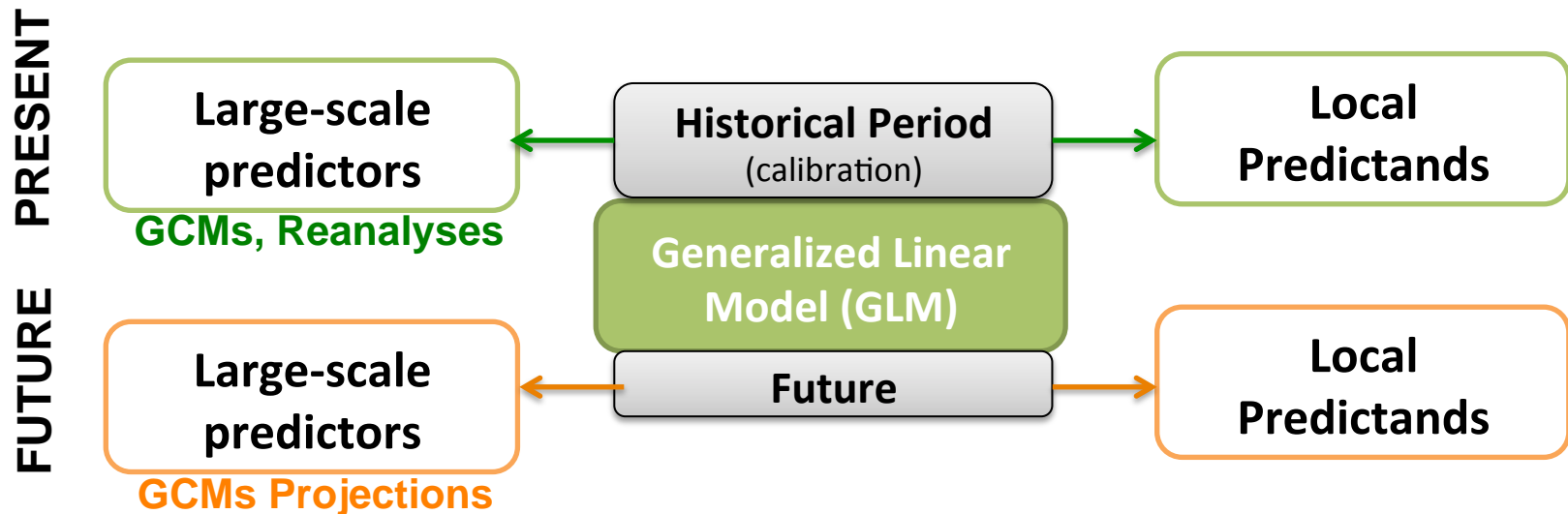
Simulations @ Leibniz-
Rechenzentrum (LRZ)/
SuperMUC, Munich

Pieri et al, JHM, 2015

Statistical downscaling

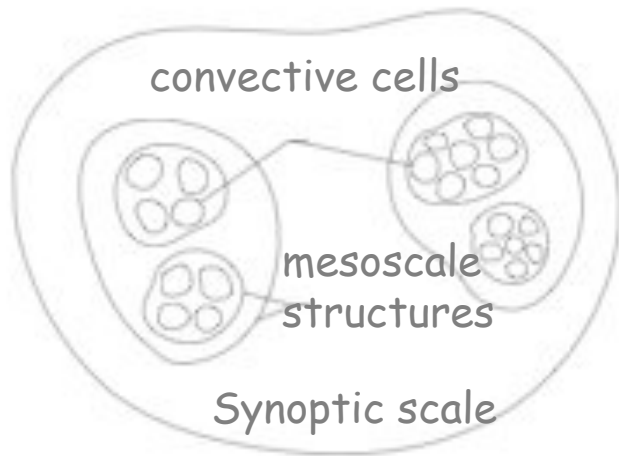
Find statistical relationships between large-scale climate features and fine-scale climate for a give region:

1. Find a large-scale predictor
2. Determine its relation with a predictand
3. Use the projected value of the predictor to estimate the future value of the predictand (**assuming stationarity**)



Stochastic downscaling

Highly intermittent fields such as rainfall can be difficult to handle with dynamical or statistical downscaling (no simple interpolation is possible).



- Highly **non-homogeneous phenomenon**
- Organized in **hierarchic structures (scaling property of rainfall)**
- **Highly intermittent in space and time** (alternating between dry and rainy periods).

An alternate approach is **stochastic downscaling** which leads to ensemble projections

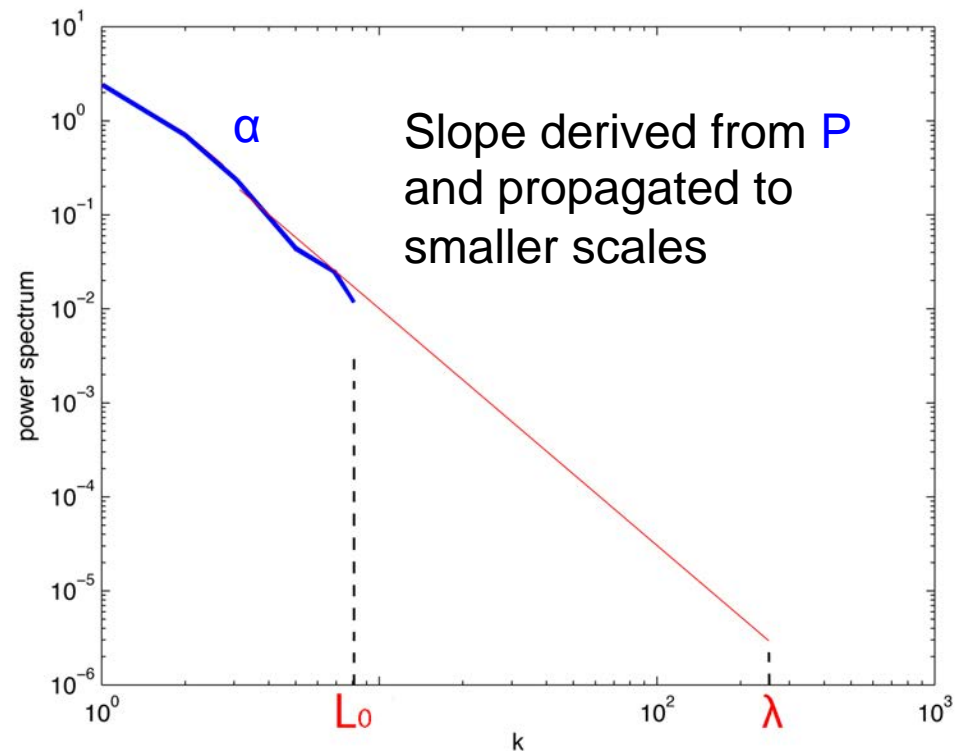
Stochastic downscaling

RainFARM (Rainfall Filtered Auto Regressive Model)

RainFARM uses simple statistical properties of large-scale rainfall fields, such as the **shape of the power spectrum**, and generates small-scale rainfall fields **propagating this information to smaller (unreliable/unresolved) scales**, provided that the input field shows a (approximate) scaling behaviour

— $P(X, Y, T)$, input field
 L_0, T_0 : reliability scales

SPATIAL Power spectrum of rainfall field



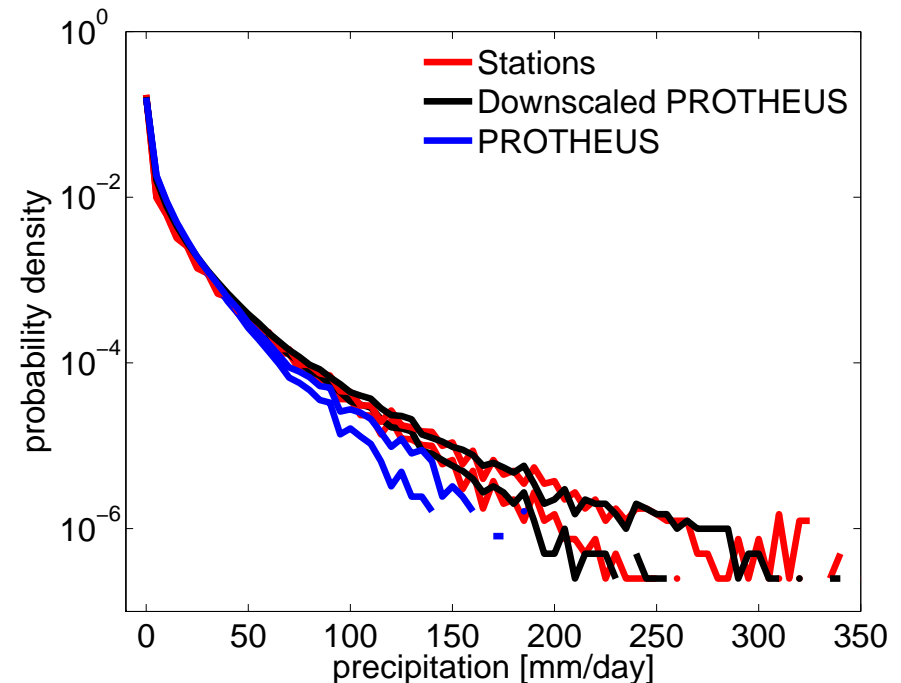
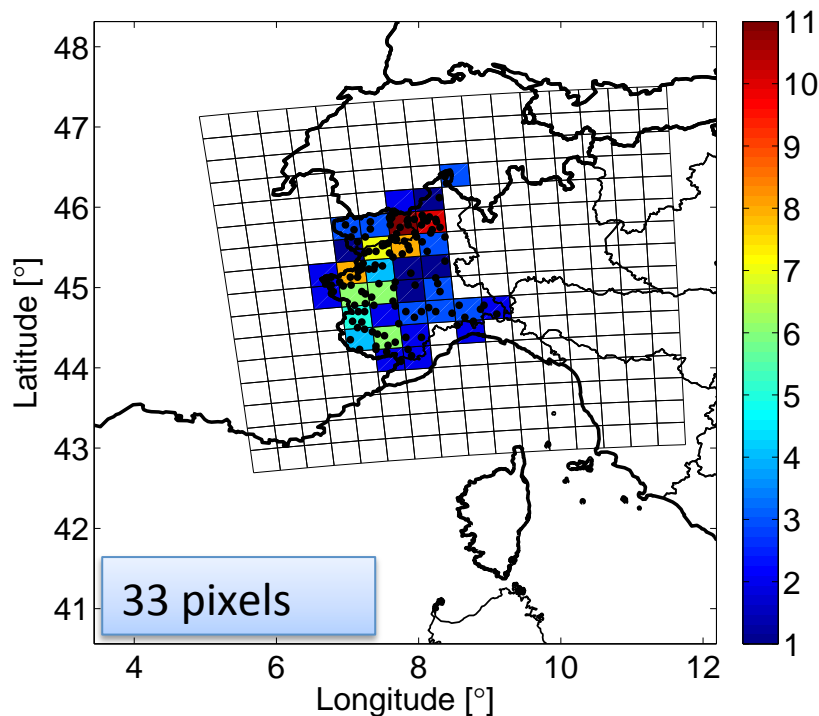
Stochastic downscaling

RainFARM (Rainfall Filtered Auto Regressive Model)

- 122 rain gauges
- 1958-2001
- Daily resolution
- Altitude max: 2526 m
- Altitude min: 127 m

D'Onofrio et al.,
J. Hydrometeor.
15, 830–843, 2014

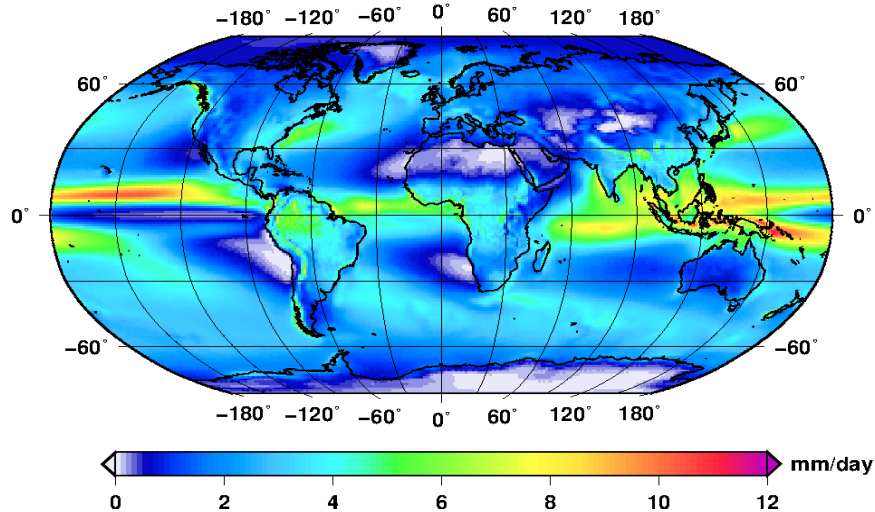
PROTHEUS: $\Delta x \approx 30\text{km}$



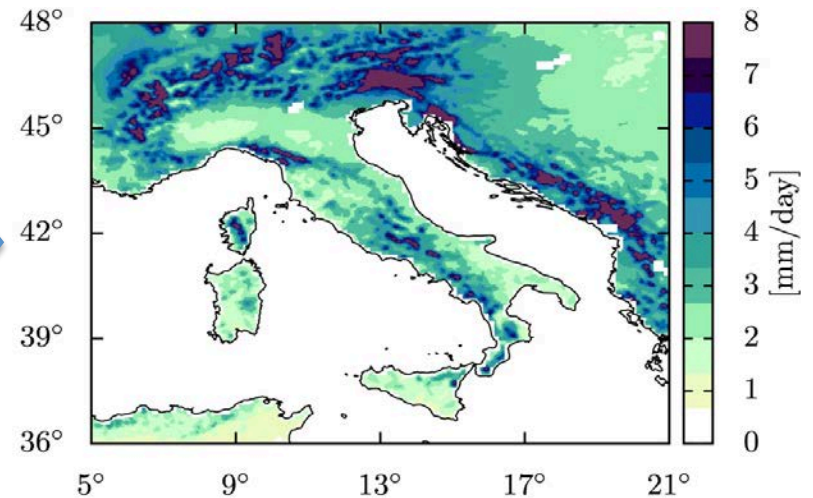
The downscaling-impact chain

Global climate model

Total precipitation annual mean 1951–2007



Regional climate model



Impact on
eco-hydrological processes

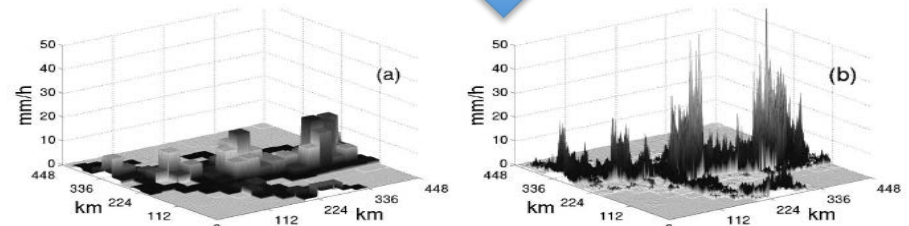


FIG. 10. (a) A snapshot of the forecasted rain field obtained from the LAM forecast and (b) one example of a downscaled field obtained by application of the RainFARM. The vertical scale indicates precipitation intensity (mm h^{-1}) and it is the same for the two fields.

Statistical/stochastic
downscaling

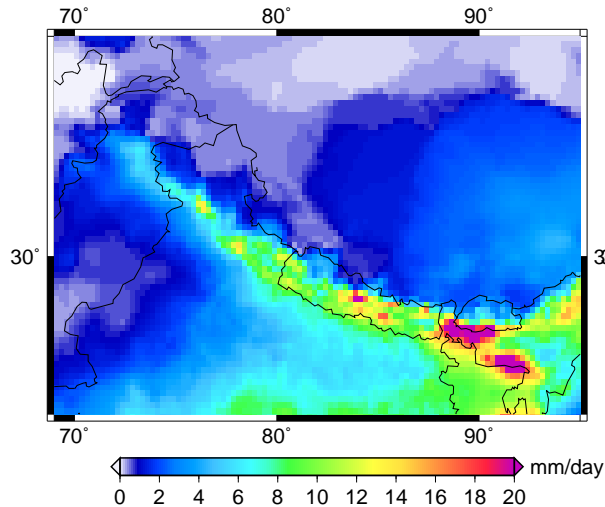


Troubles, oh troubles

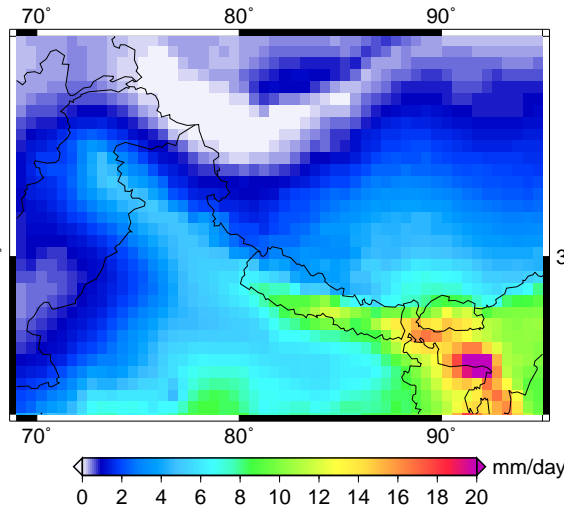
The chain of uncertainties: (1) data for model validation

Summer precipitation (JJAS), Multiannual average 1998-2007

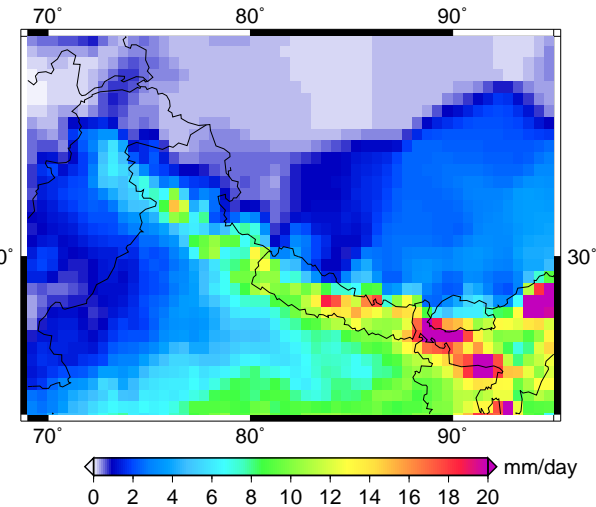
Aphrodite JJAS



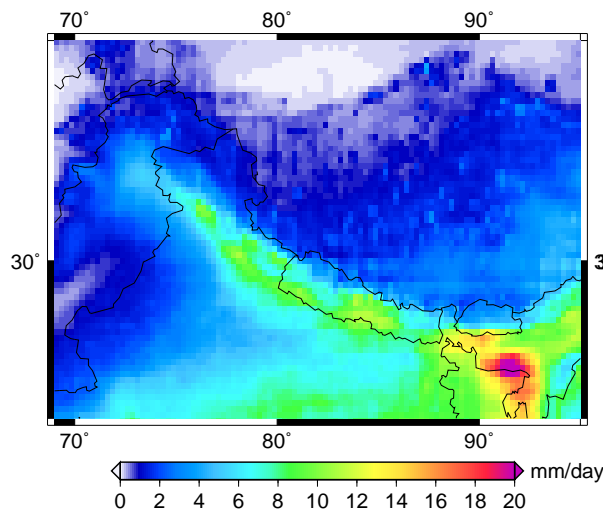
CRU JJAS



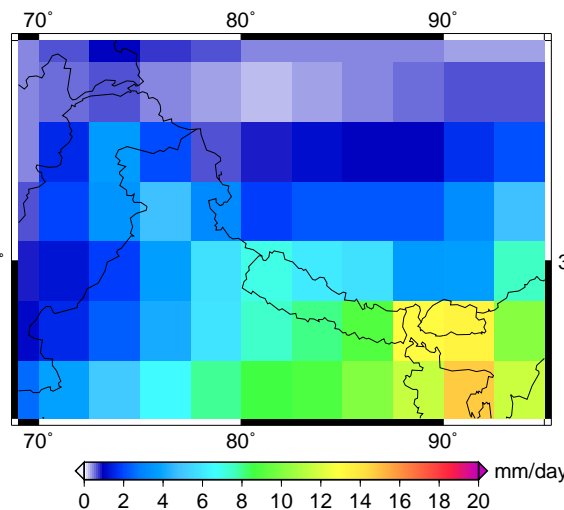
GPCC JJAS



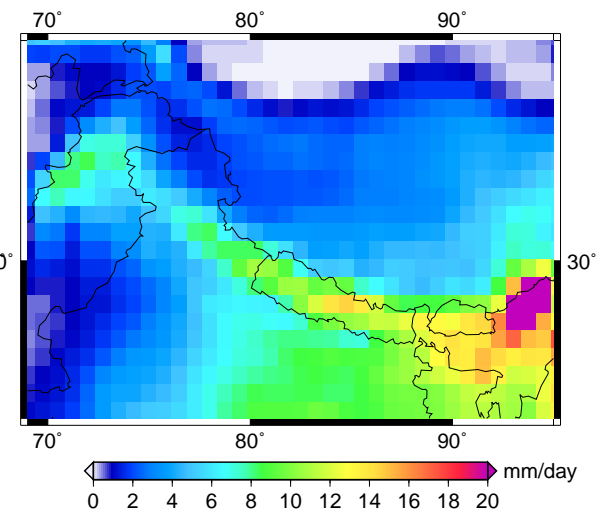
TRMM JJAS



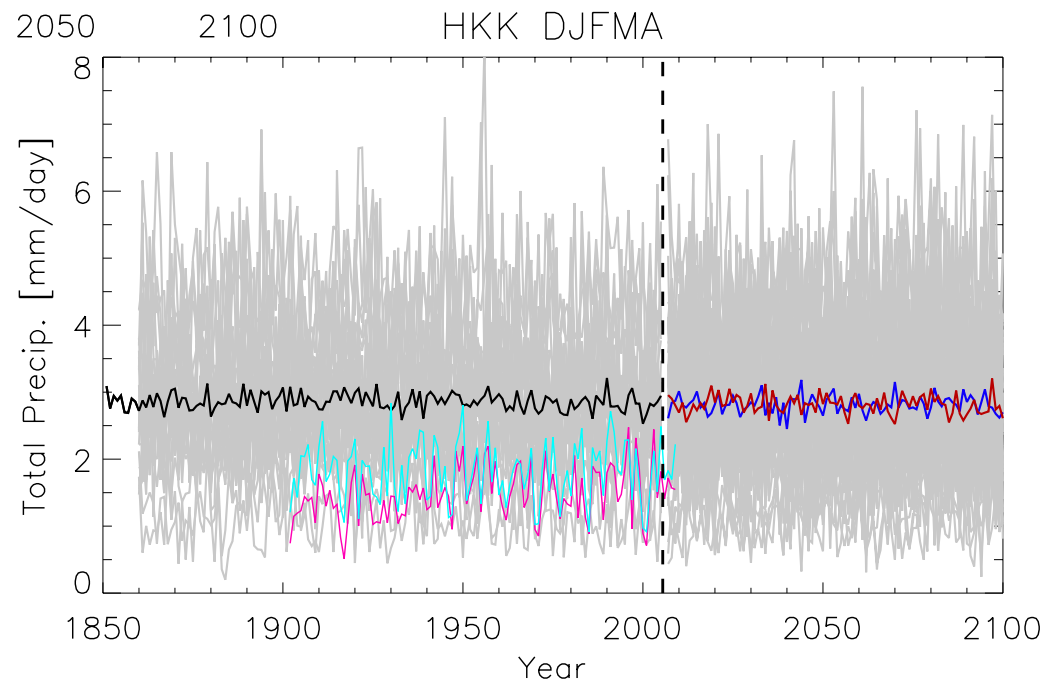
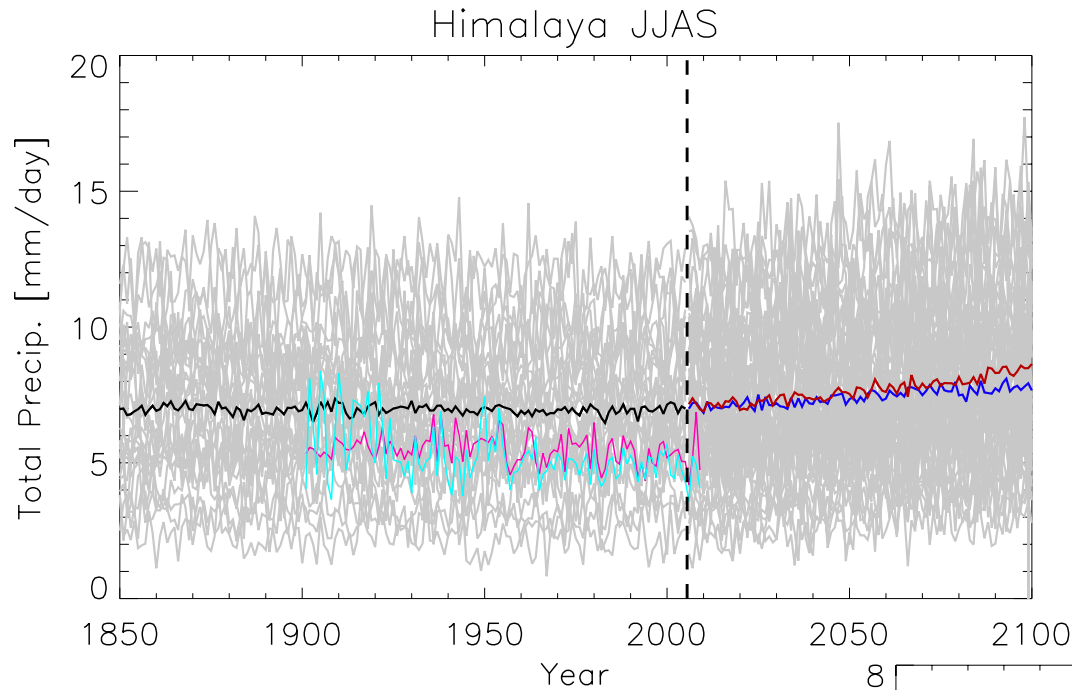
GPCP JJAS



ERA-Interim JJAS

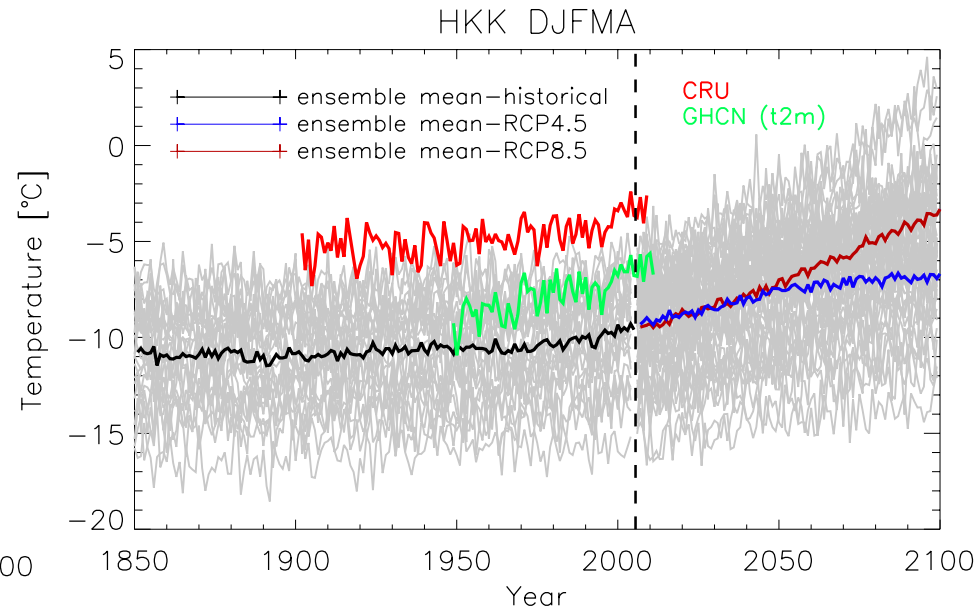
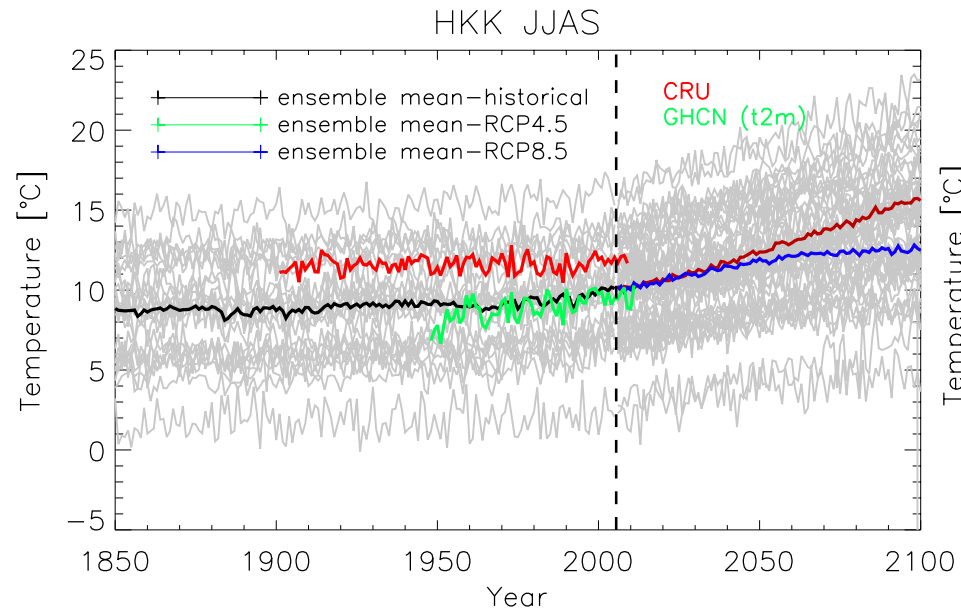
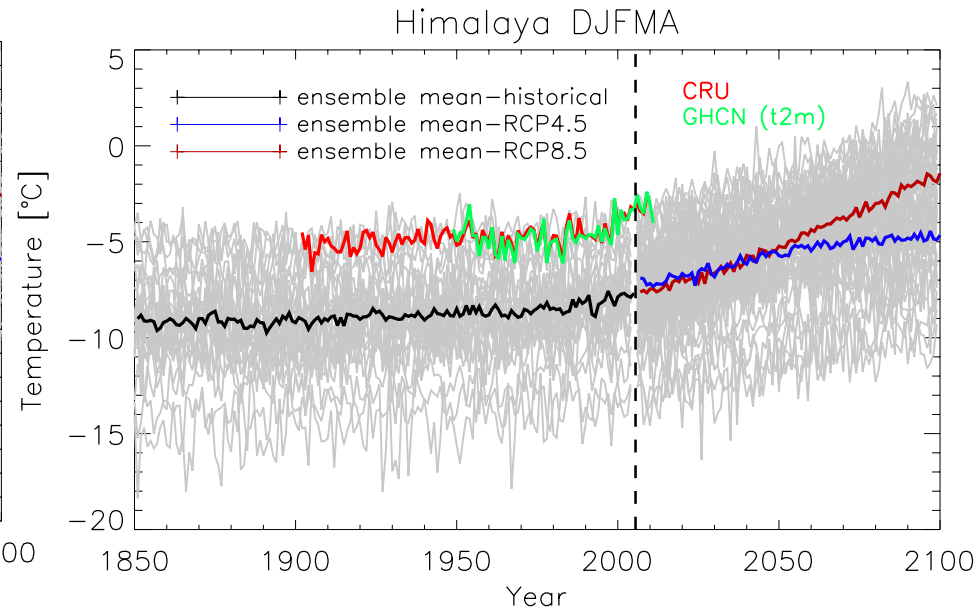
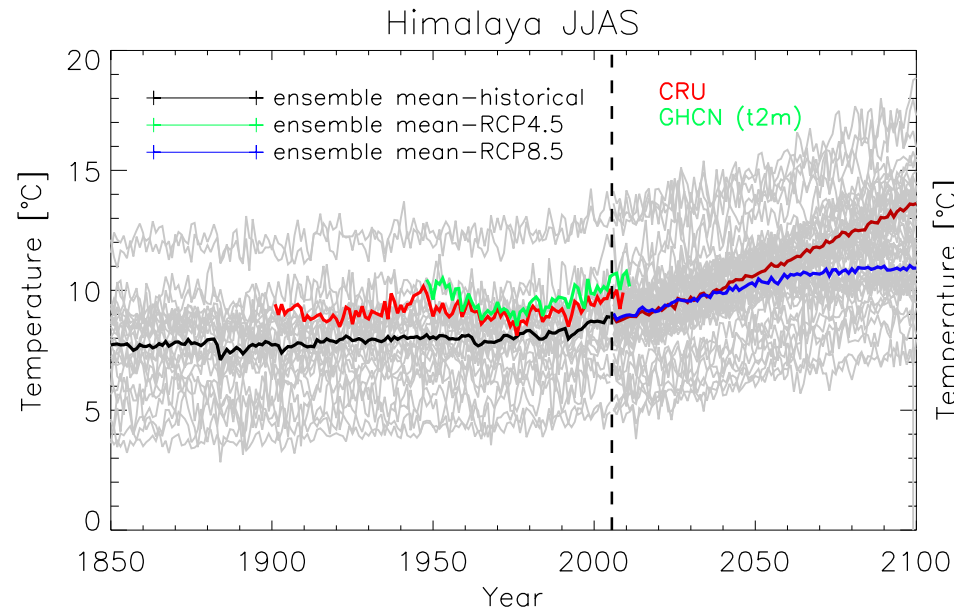


The chain of uncertainties: (2) spread between CMIP5 models

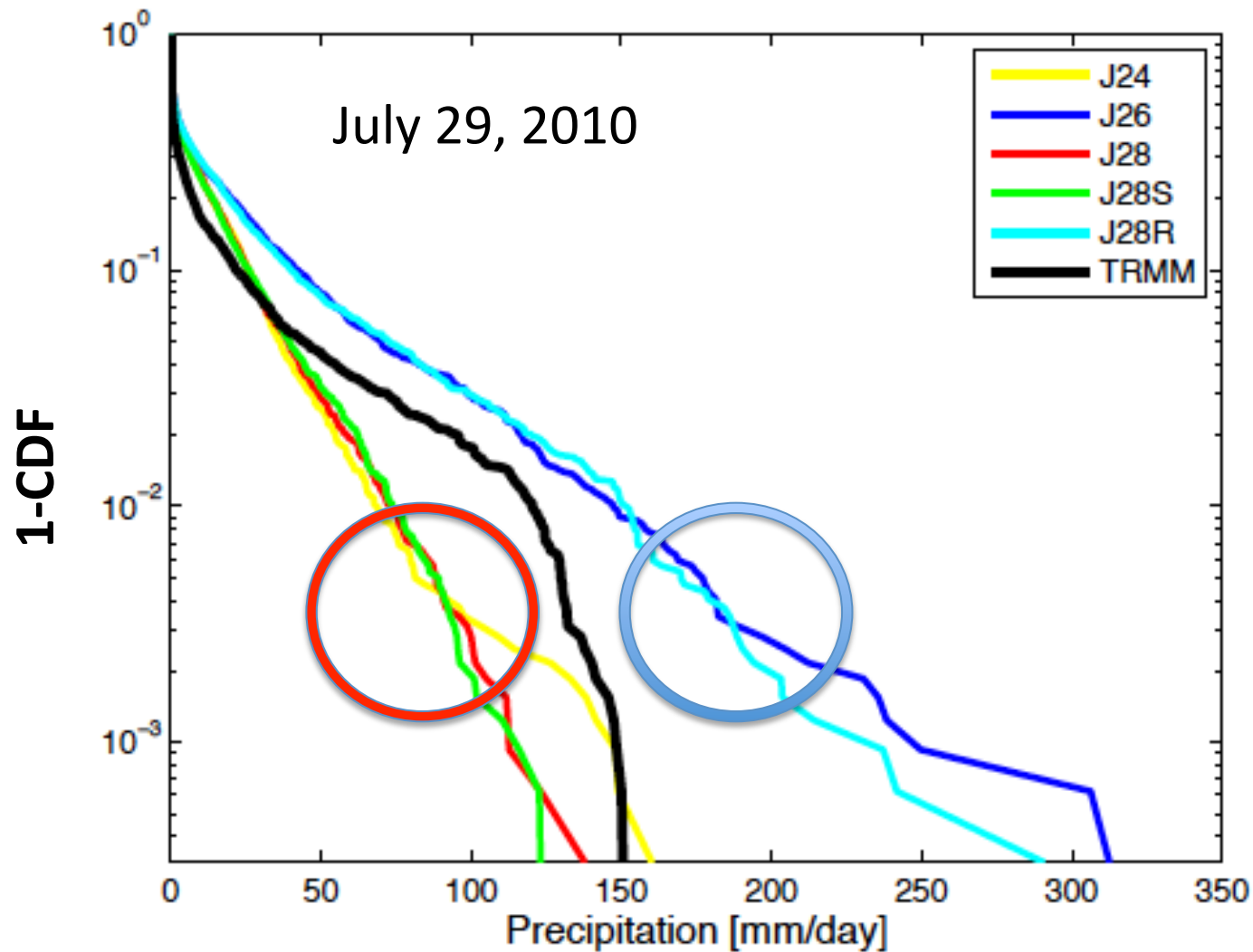


*Palazzi E., von Hardenberg J.,
Terzago S., Provenzale A.:
Precipitation in the Karakoram-Himalaya:
A CMIP5 view, Climate Dynamics, 2014*

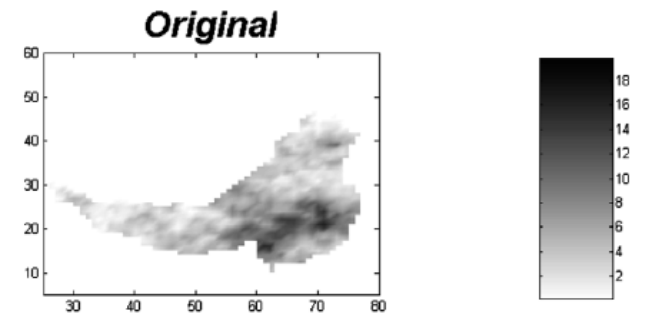
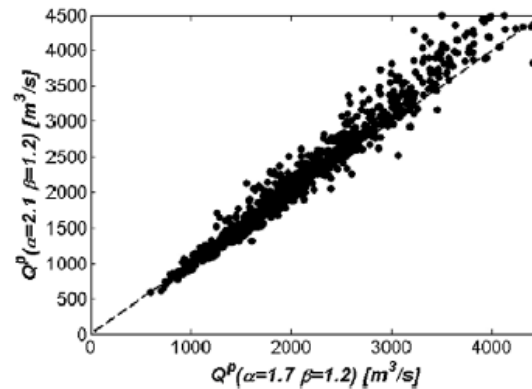
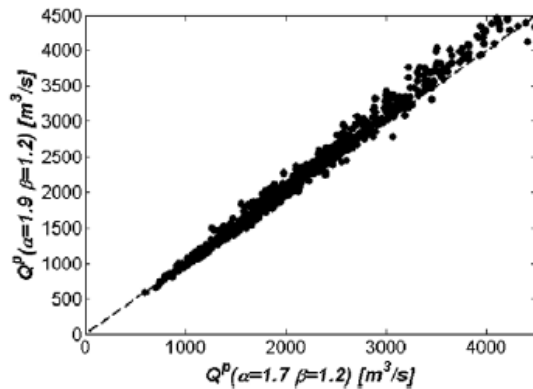
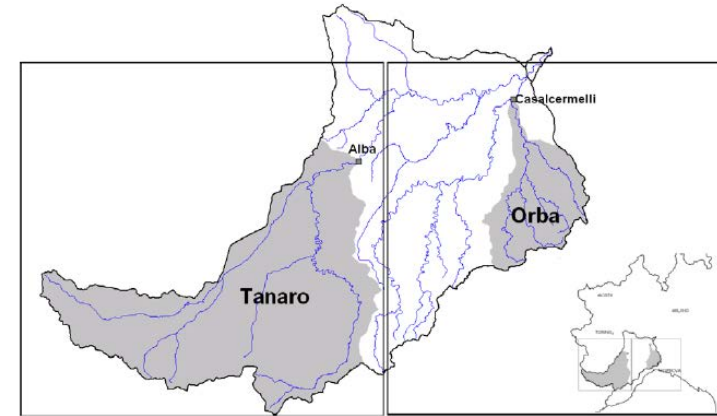
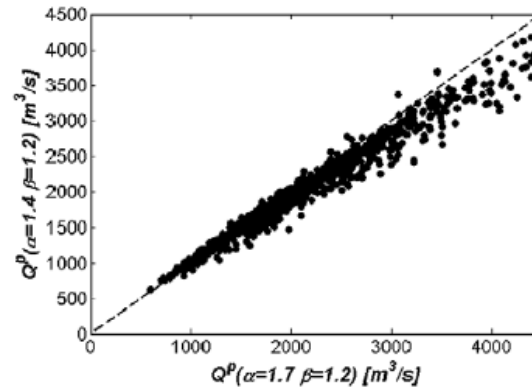
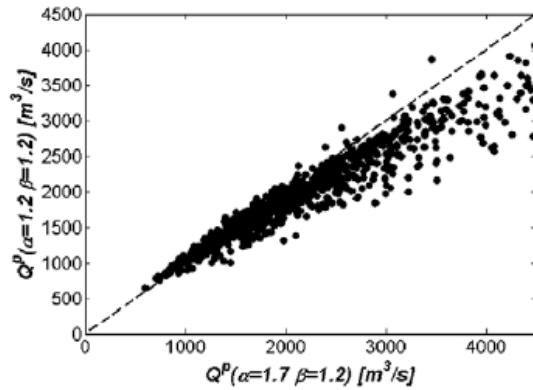
And the spread of CMIP5 temperatures



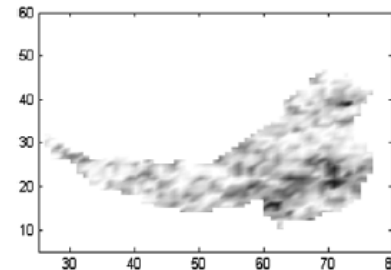
Precipitation statistics from WRF (Pakistan Flood 2010)



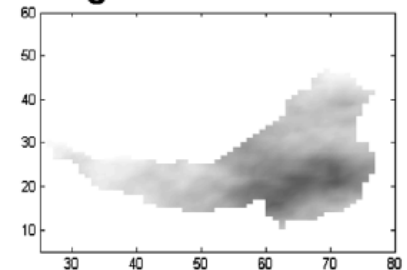
The chain of uncertainties: (3) downscaling



Lower correlations

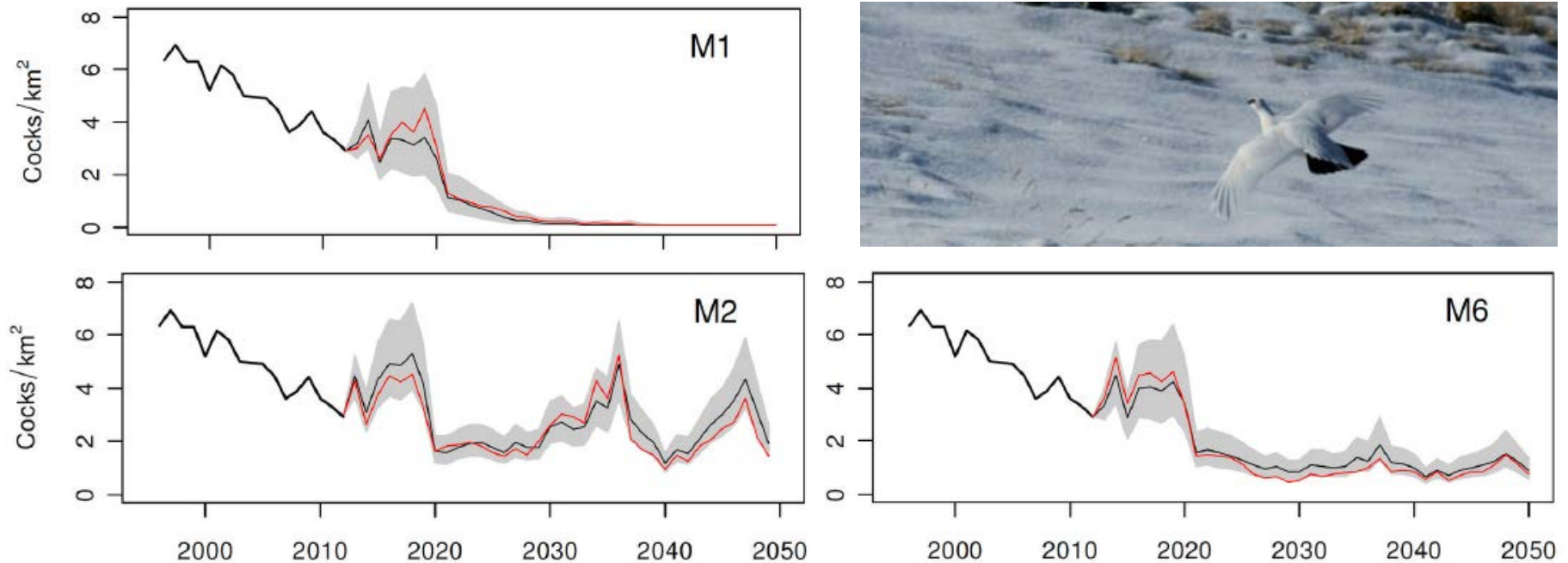


Higher correlations



Gabellani, Boni, Ferraris,
von Hardenberg, Provenzale
Adv. Water Res. 2007

The chain of uncertainties: (4) local impact models



Model	Intercept	$\ln N_{t-1}$	$\ln N_{t-2}$	SE_{t-1}	SS_{t-1}	SP_t	$T(\text{July})_{t-1}$	$P(\text{July})_{t-1}$	$T(\text{Jan-Mar})_t$	$T(\text{Apr-May})_t$	var.	R^2	AICc
M1	-0.07 ± 0.04			-0.19 ± 0.04	-0.18 ± 0.04						2	0.78	-50.53
M2	0.34 ± 0.24		-0.25 ± 0.14	-0.19 ± 0.04	-0.19 ± 0.04						3	0.83	-50.20
M3	-0.07 ± 0.04			-0.19 ± 0.04	-0.18 ± 0.04			0.05 ± 0.03			3	0.82	-49.28
M4	-0.07 ± 0.04			-0.19 ± 0.04	-0.17 ± 0.04		-0.05 ± 0.04				3	0.81	-48.51
M5	-0.07 ± 0.04			-0.20 ± 0.04	-0.18 ± 0.04				-0.03 ± 0.04		3	0.79	-47.28
M6	0.08 ± 0.26	-0.10 ± 0.16		-0.18 ± 0.04	-0.17 ± 0.04						3	0.78	-46.98

Simona Imperio, Radames Bionda, Ramona Viterbi, Antonello Provenziale,
Alpine rock ptarmigan, PLOS One, 2013

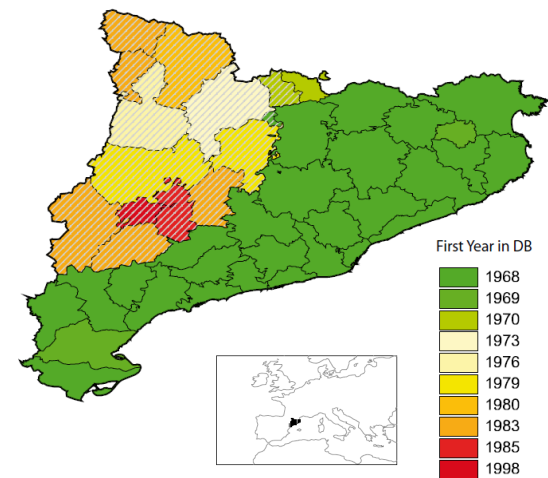
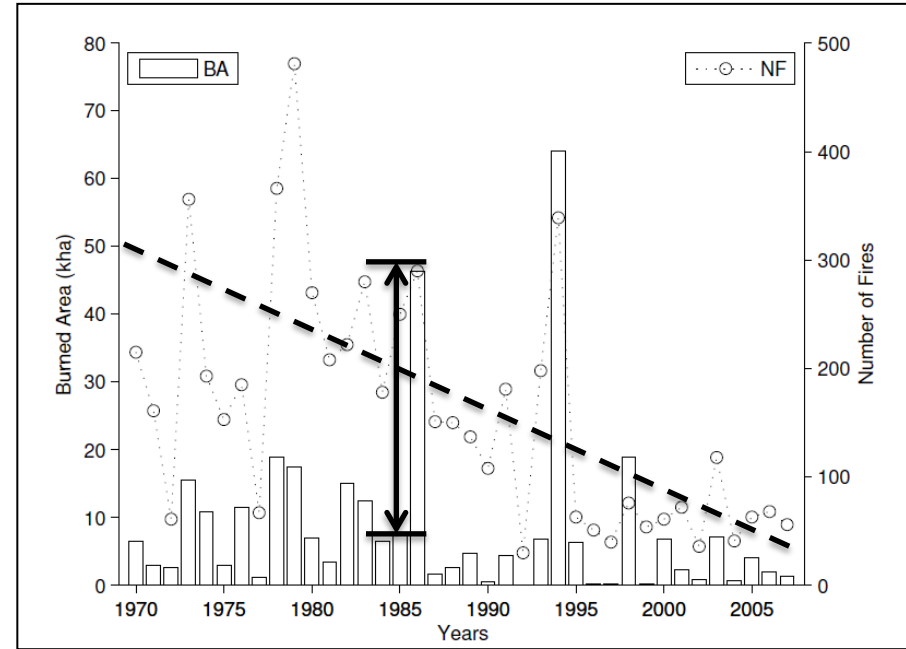
The chain of uncertainties: (4) local impact models

Climate change and forest fires

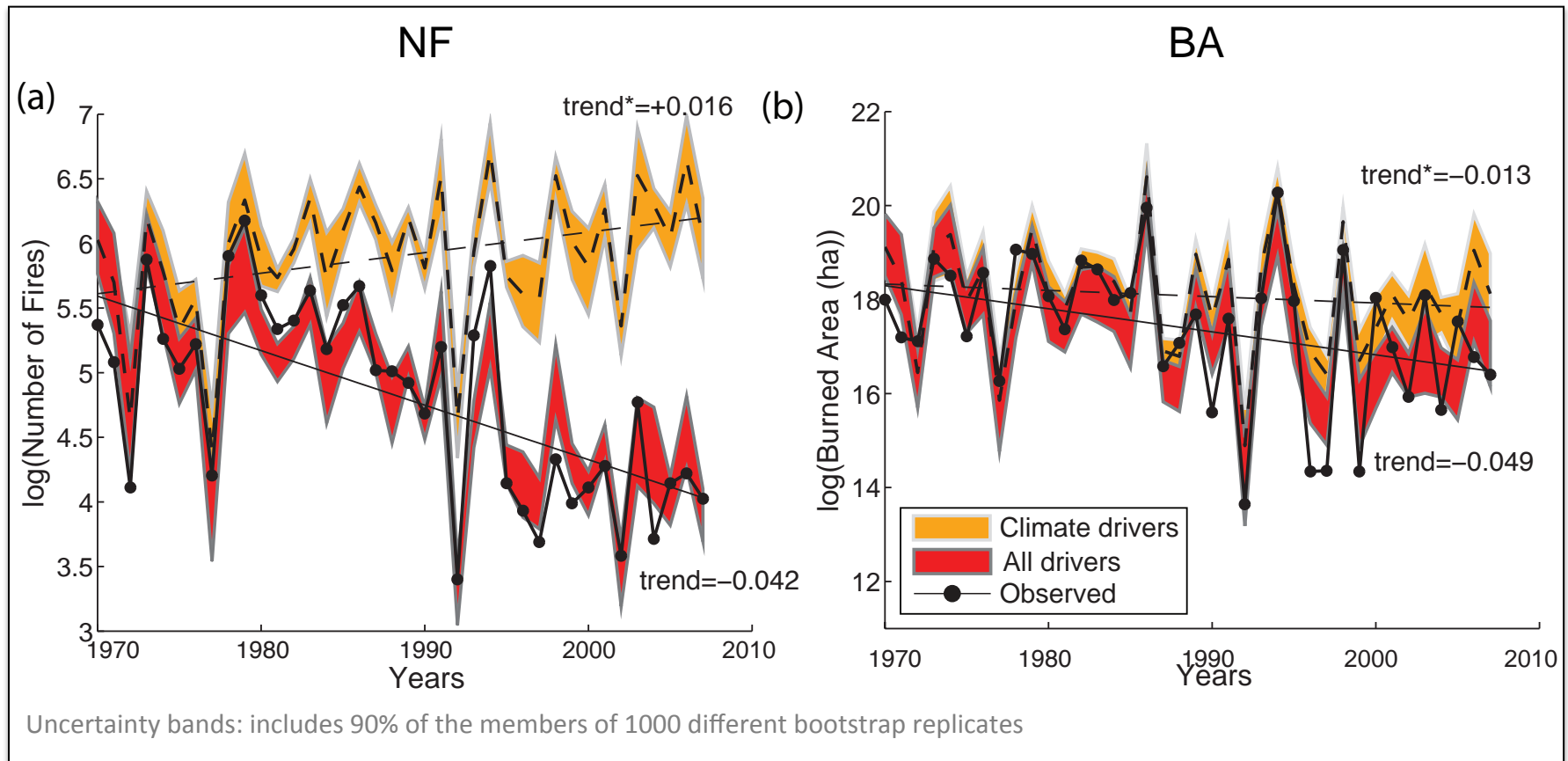
Long-term changes → human activities, climate trends.

The year-to-year changes in NF and BA are mainly related to **climate variability**.

The climate acts mainly on two aspects:
(i) **antecedent climate** → fuel to burn; (ii) **coincident climate** → fuel flammability.

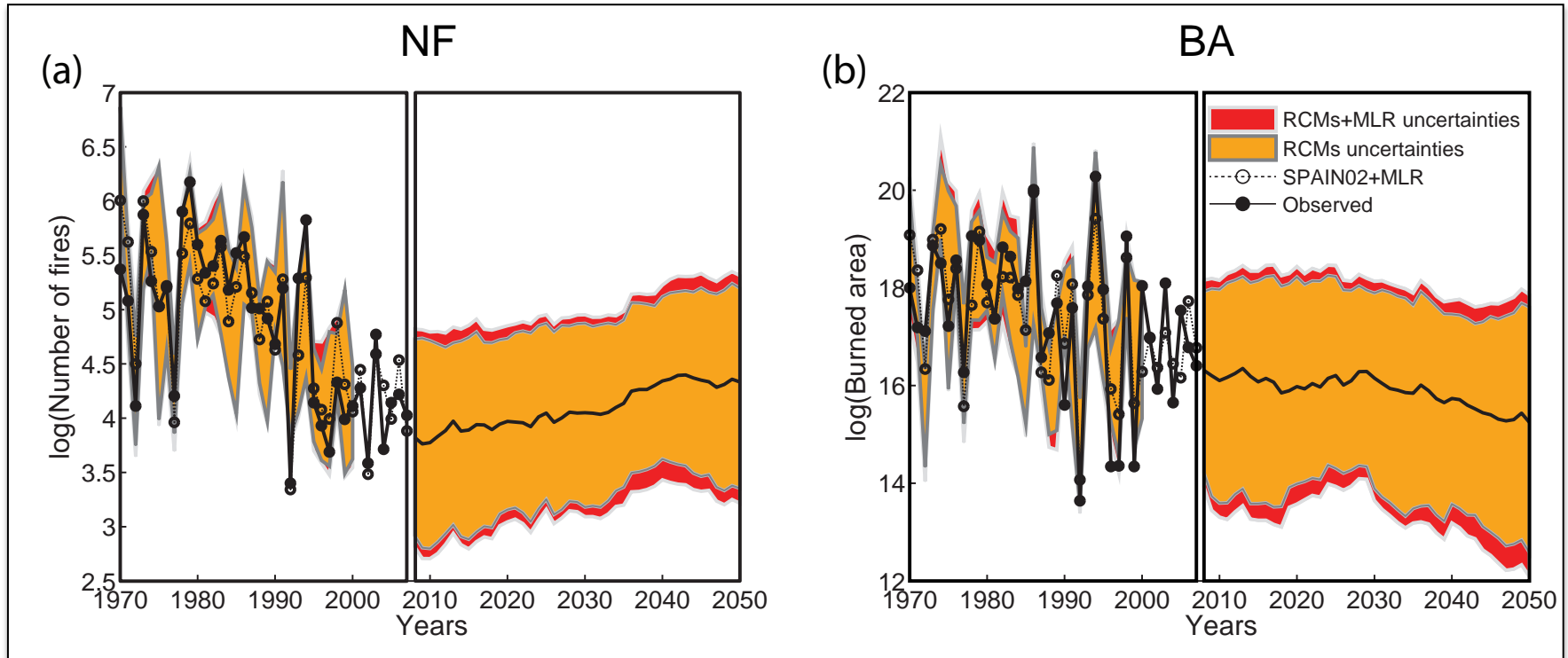


Fire response to climate trends



Climate drivers = both interannual variability and trends are driven by climate
All drivers = MLR considers the year-to-year climate variation + overall trend

Impact of future climate change on wildfires



- Future response depends on management strategies
- Uncertainty in RCM scenarios is larger than impact model uncertainties for forest fires

Conclusions (from climate to impacts)

Scale mismatch between climate models (and drivers) and land surface response: need for **climate downscaling**

Huge uncertainties in data, climate models, downscaling procedures, impact models: need for **ensemble approaches**, need for **uncertainty estimates**, need for **caution** in providing and interpreting results.

**Find the best strategy
without ideological constraints**