

# Climate Modeling and Downscaling

# Types of climate-change experiments: a preview

- 1) “What-if” sensitivity experiments – increase the optically active gases and aerosols according to an assumed scenario, and compare the model solution with that from a no-change experiment. (IPCC-type experiments)
- 2) Deterministic initial-value forecasts
- 3) Feedback sensitivity studies
- 4) Anthropogenic landscape-changes – impacts on climate
- 5) Downscaling from AOGCMs, using regional models

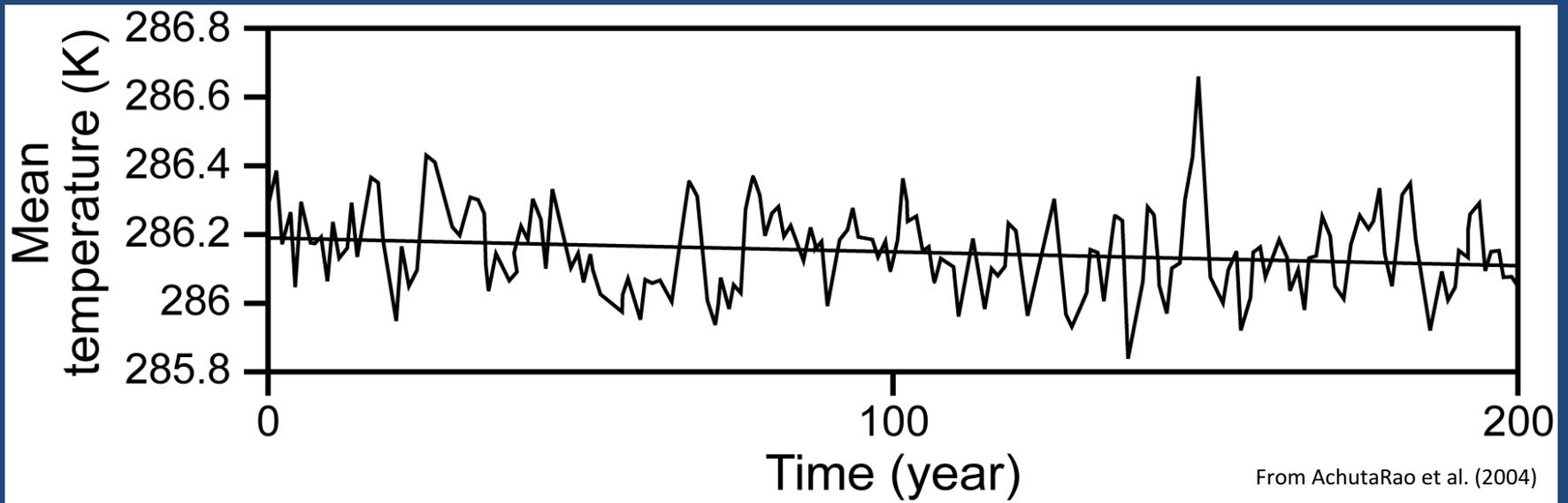
# Type 1: IPCC-type Greenhouse-gas-impact studies

- Objective - define the change associated with anthropogenic forcings. Somehow, the natural “internal variability” of the atmosphere should be filtered from the solution.

# The concept

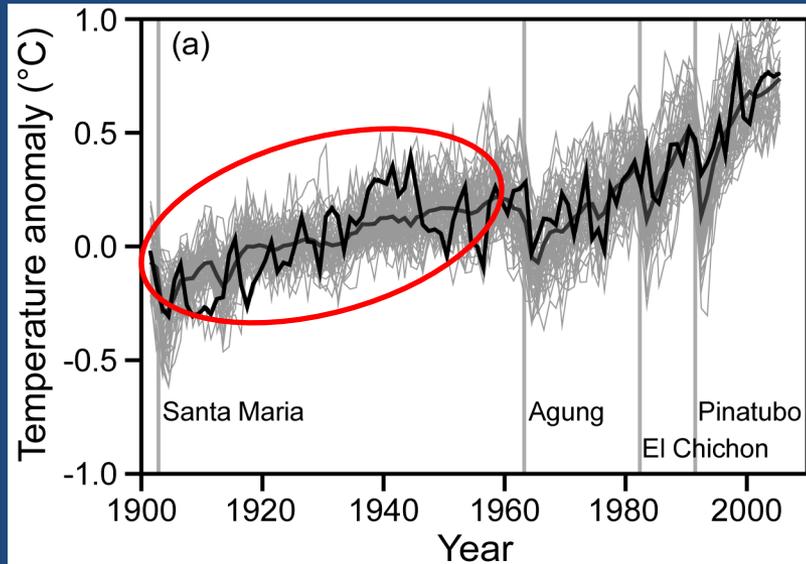
- The change in the global-average (or regional) weather conditions is a sum of
  - the natural (internal) variation of the climate system, that would have occurred without human impacts
  - the climate change that results from the human impact (greenhouse gases, land-surface modification)
- Need to isolate the human impact

# Example 1 of internal variability

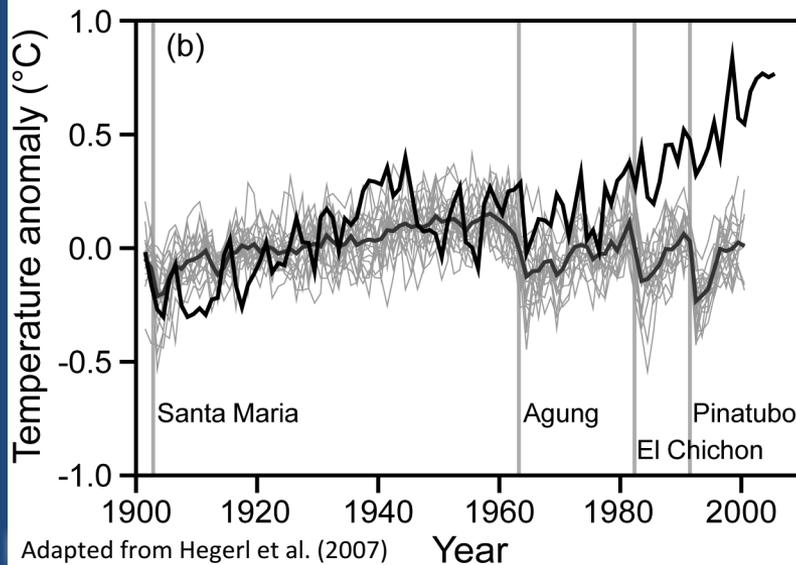


Global-mean surface air temperature from a simulation by an AOGCM that has no anthropogenic forcing.

# Example 2 of internal variability



The solid black line is the observed global average surface air temperature



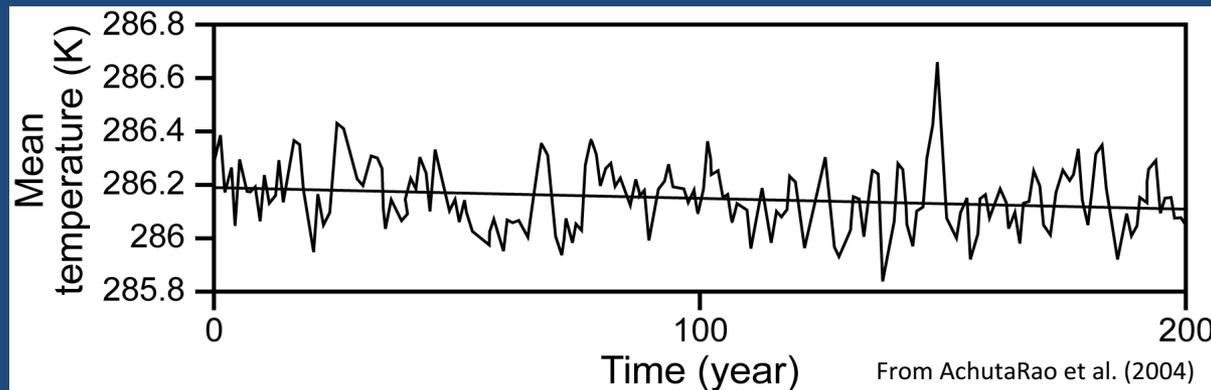
What is the cause of  
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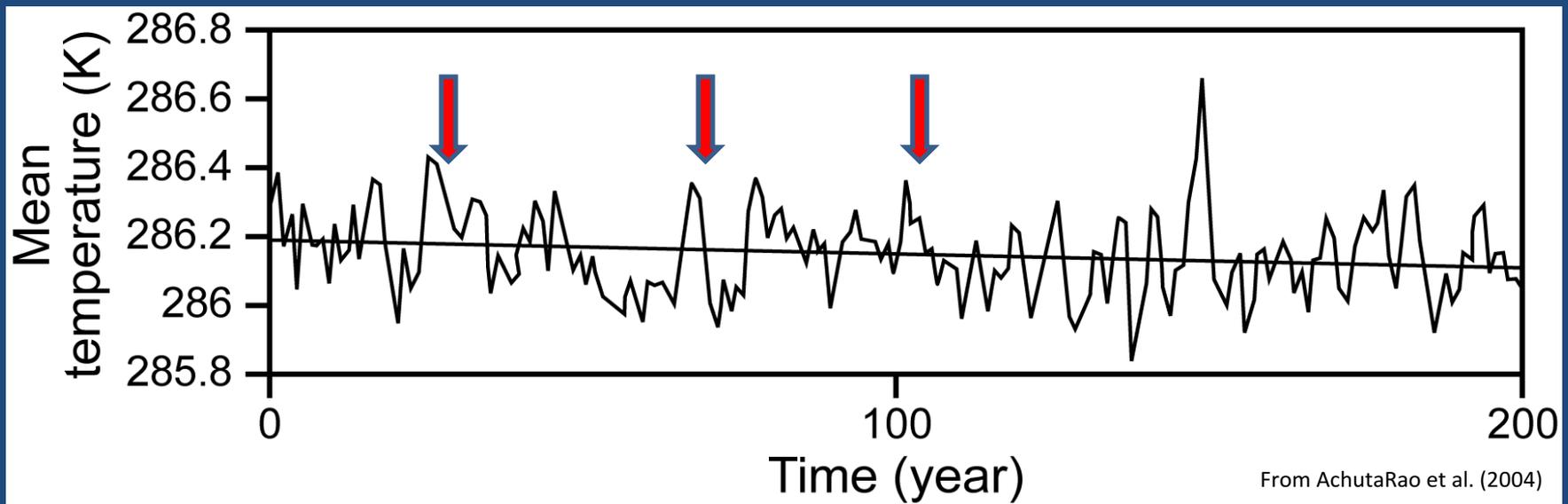
- Natural long-time-scale physical processes
  - Deep ocean circulations
  - Land and sea ice
  - Land surface

# Experimental design for the IPCC-type human-impact studies

- Generate a control run for the present climate
  - allow possibly thousands of years to spin up, and then run for a couple of hundred years of the present climate (no change in CO<sub>2</sub> or aerosol forcing).

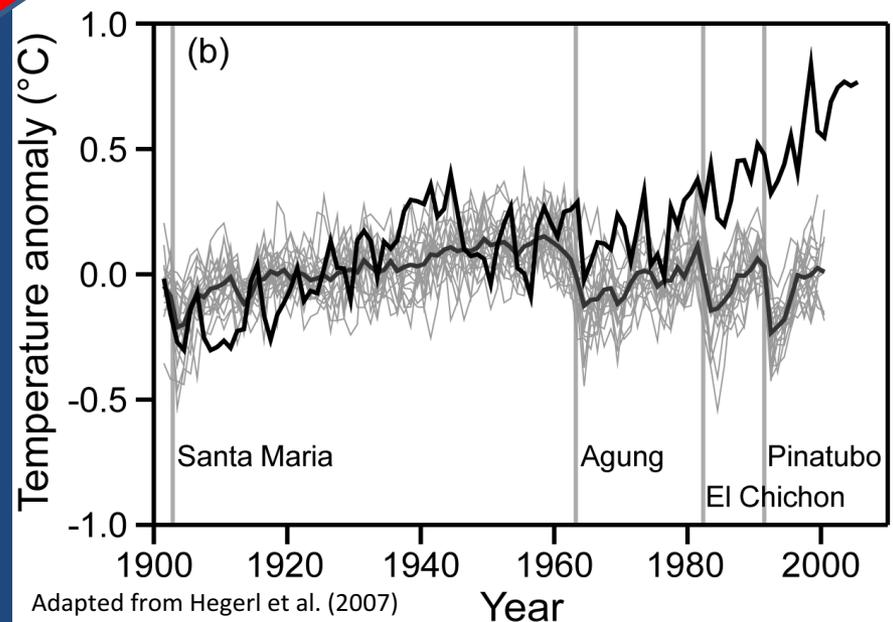
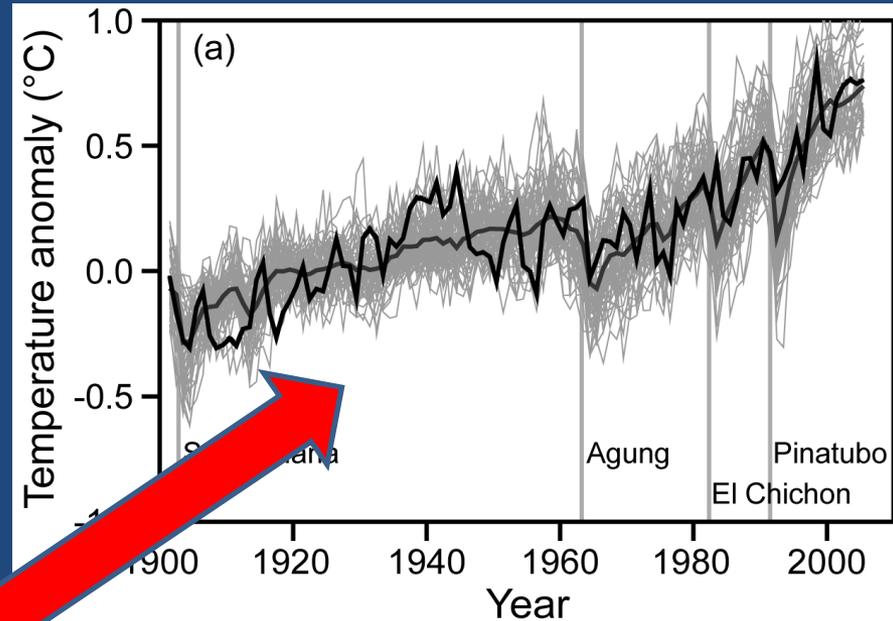


- Begin a future-climate run at any arbitrary time by imposing a ramp up, as a function of time, in CO<sub>2</sub> or aerosol forcing – there are many accepted scenarios.



- Where you start in the internal cycle determines the change.

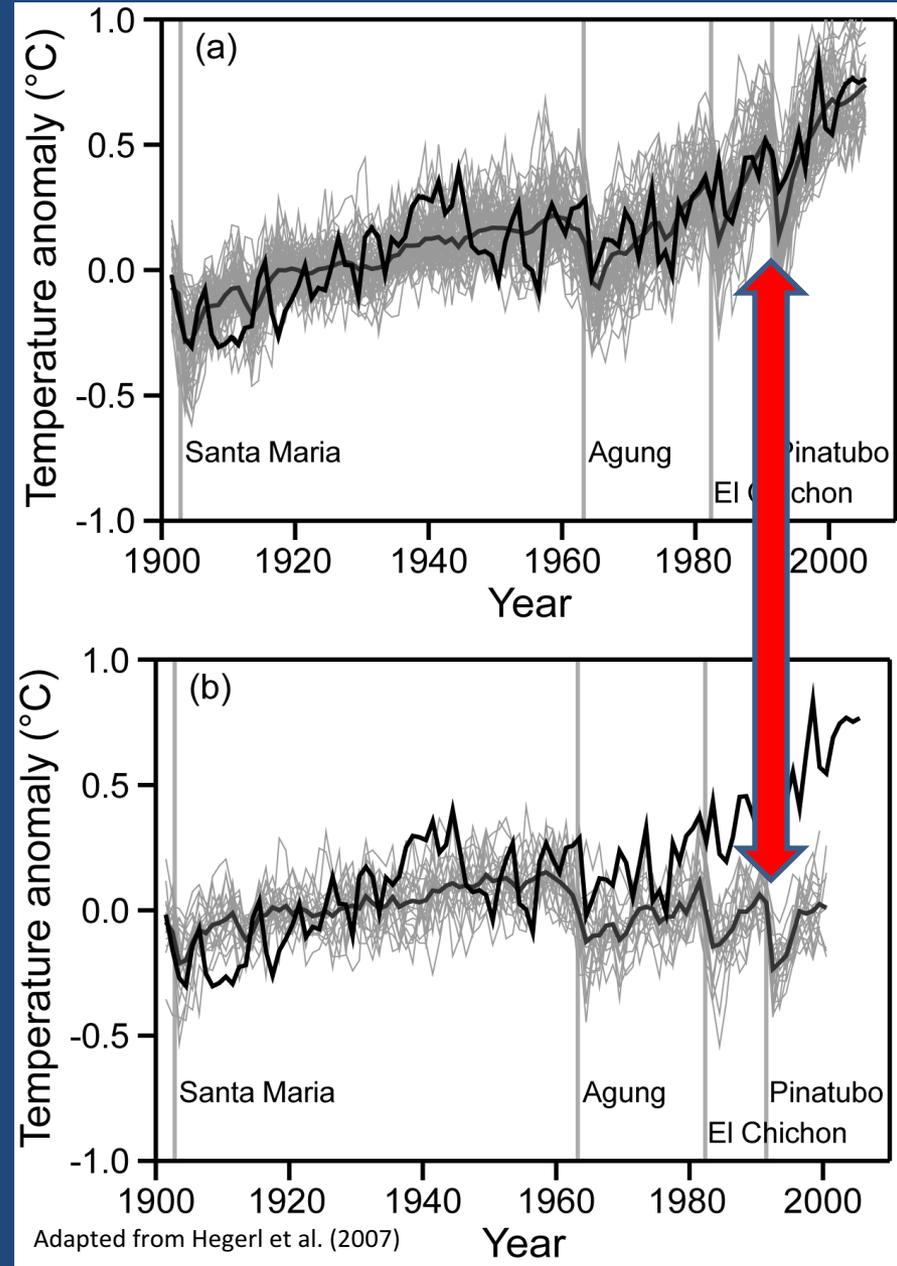
- Solution - Run an ensemble of simulations using the same model with different start times, and average them – removes most of the effect of the internal variability.
- Or, create an ensemble by running many different AOGCMs (from different organizations and countries) for the same scenario, and average them.



Adapted from Hegerl et al. (2007)

- These are NOT deterministic, initial-value predictions in the sense that we start from an observed state of the physical system – rather we have sensitivity experiments.

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## Type 2) Deterministic initial-value forecasts

- Can be used for seasonal to decadal prediction.
- Must initialize all components of the physical system – deep ocean circulations to deep-soil conditions. (How do we do this?)
- This is being done operationally for inter-seasonal prediction – e.g., the NCEP Climate Forecast System (CFS).
- For longer periods (decadal) – major research area.

## Type 3) Feedback sensitivity studies

- Experiments to define the strengths and sign of feedbacks between radiative forcing and climate
- For example, one standard procedure defines the “equilibrium climate sensitivity”
  - Double the CO<sub>2</sub> instantaneously and let the model run to an equilibrium near-surface temperature.
  - The change in the temperature reflects the sensitivity.

## Type 4) Anthropogenic landscape changes: impacts on climate

- Can use a global or regional model.
- These are sensitivity studies of the sort that we discussed in the chapter on experimental designs for research – conduct a control simulation and then one with an existing or future anthropogenic change.
- These sensitivity simulations will be longer in duration than the simple ones discussed before

# Type 5) Downscaling from AOGCMs, using regional models

- Motivation – response to climate change is often at local level, so information is needed at high spatial resolution.
- Two approaches:
  - **Dynamical downscaling** – Force the LBCs of a mesoscale model with output from an AOGCM. Do “time slices” to make this computationally tractable. (**Risks of time slices?**)
  - **Statistical downscaling** – Statistically relate the small-scale response to the large (AOGCM) scale processes.

# Special requirements for AOGCMs used for climate studies/prediction

- **Land-surface modeling**
  - Carbon sources and sinks
  - Dynamic vegetation models – response to CO<sub>2</sub>, T, precipitation
  - Soil layers that extend through a deep layer
  - Plant root dynamics
  - Dust elevation
- **Ice modeling**
  - Snowpack modules
  - Motion and thickness of land and sea ice
  - Permafrost
  - Polar ice caps

- **Water bodies**

- Deep-ocean circulation – have different resolutions and time steps than the atmospheric model.
- Ocean chemistry
  - Salinity – affects density and saturation vapor pressure
  - CO<sub>2</sub> exchange with the atmosphere
  - Marine biosphere
  - Nutrient input from the atmosphere
- Wave intensity – affects albedo, evaporation rate, aerosol (salt) source, roughness.

- **Physical-process parameterization**

- Small errors may lead to a large cumulative effect; e.g., cloud-errors causing a drift in the temperature.
- Existing parameterizations are often tuned for the current climate.
- Different approaches may be needed versus what is used for weather modeling, because of coarser resolution.

- **Conservation properties of dynamical cores**
  - mass and energy, more-important for long integrations
  - “mass fixer” sometimes used to compensate for mass leaks or sources.
- **Initial conditions**
  - For IPCC-type GHG impact experiments, and for assessing the impact of projected landscape changes, any realization of the current climate will work.
  - For initial-value simulations, the state of the entire physical system must be initialized.
- **Flux corrections**
  - To compensate for small errors in the fluxes of heat, water vapor and momentum , artificial corrections have been added to some models.
  - These corrections are non physical.
  - Few contemporary AOGCMs are flux corrected.

# Verification of AOGCMs for past or present climates

- Why is success with the current climate not a guarantee that the model will be able to replicate a future climate?

# Verifying recent versus past climates: relative advantages

- Recent/present climate (e.g., the last 100 years)
  - Conventional observations are available for verification.
  - External forcings are known (volcanic eruptions, aerosols, orbital and solar properties)
  - Climate changes have been small, compared to what has been experienced over longer periods.

- Past (paleo) climates
  - Must rely on proxy information to define the climate conditions for verification, and for the external forcings.
  - Time periods are too long to simulate with full-physics models.
  - Significant changes took place.

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individual weather events

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- Why?

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- The climate signal results from the aggregate effect of individual weather events and their simulated properties – e.g., cyclone tracks and intensities.
- These individual events should be simulated well in order to get the climate right.
- Even if the current climate statistics look right, they may be “right” for the wrong reason...unless you look at individual events.

# Testing of climate models at the “individual-component” level

- Climate models are obviously much more complex than NWP models...with many more “components”.
- Thus, to the extent that individual components can be tested in isolation, the more confidence we can have in the entire system.

# Metrics for climate model verification

- Global-mean quantities – e.g., temperature
- Composite global indices, based on many variables
- Spatial patterns of variables (e.g., annual precip, SLP)
- Replication of specific features/processes – e.g., ENSO, diurnal cycle of precip
- Regional extremes

# Verification of global-average climate statistics

- Challenge – What variable(s) best represent climate, and can serve as metrics of the overall errors in the simulations?
- There are many possible variables associated with the atmosphere, hydrosphere, cryosphere, lithosphere and biosphere.
- Some studies simply use global-mean temperature, and others use a “composite index”.

# An example of a composite index from simulations of current climate

- Simulations were from three Climate Model Inter-comparison Projects (CMIP)
  - CMIP1 – mid 1990s
  - CMIP2 – early 2000s
  - CMIP3 – late 2000s (based on IPCC AR4)
- Verification was based on a synthesis of reanalyses and observations, to create annual-mean climatologies from 1979-1999.
- Calculating a model-performance index

$$e_{vm}^2 = \sum_n w_n (\bar{s}_{vmn} - \bar{o}_{vn})^2 / \sigma_{vn}^2$$

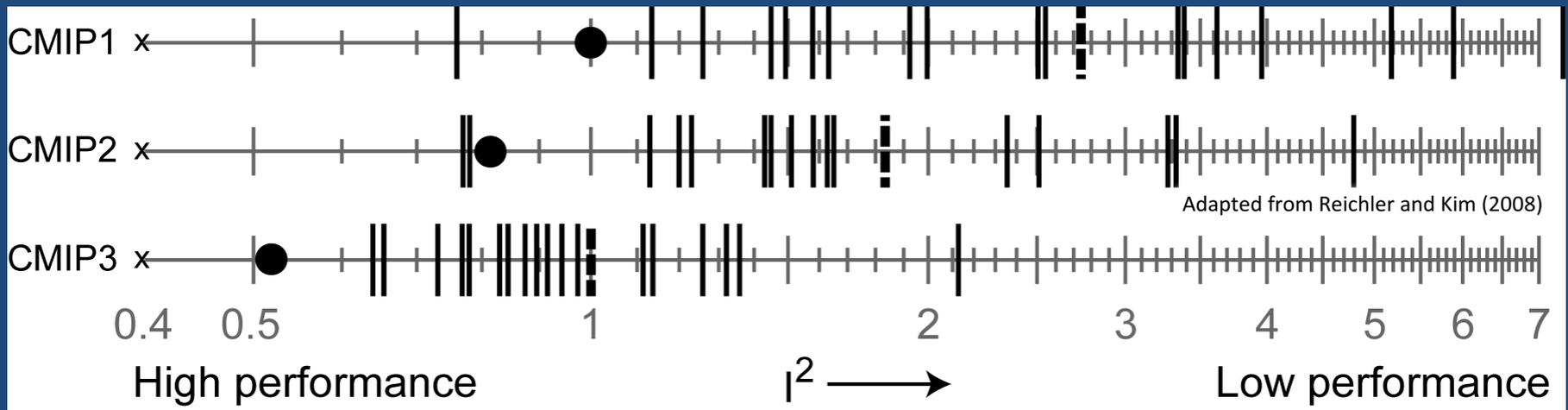
$$I_{vm}^2 = e_{vm}^2 / \overline{e_{vm}^2}^m$$

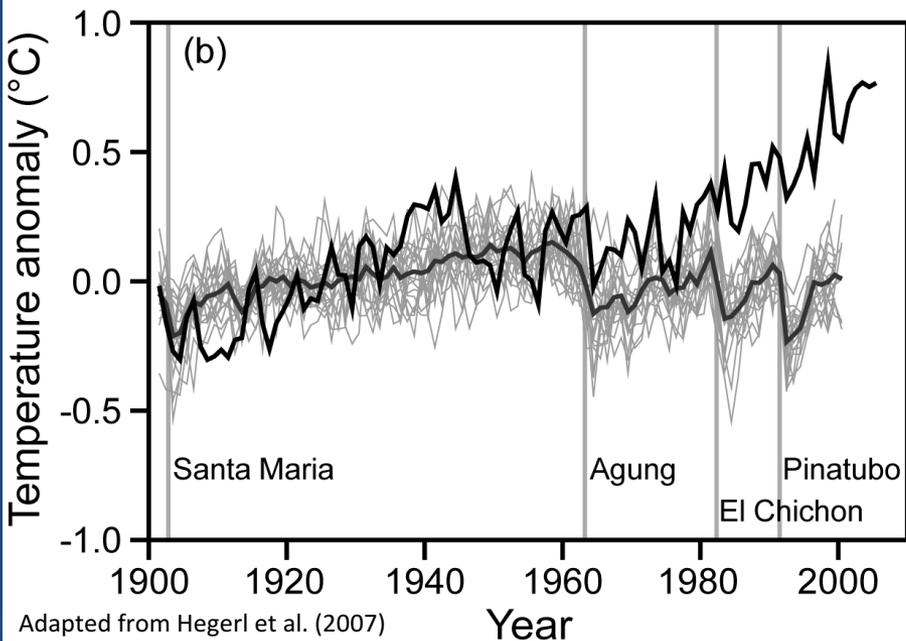
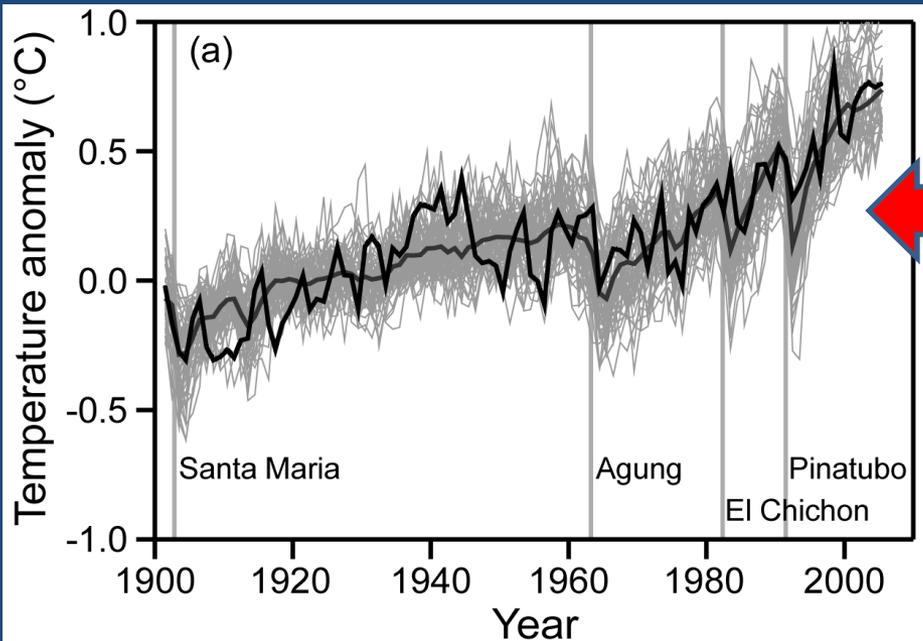
$$I_m^2 = \overline{I_{vm}^2}^v$$

- $e^2$ : normalized error variance
- $s$ : simulated annual climatological mean
- $o$ : observed annual climatological mean
- $w$ : weights
- $\sigma$ : interannual variance based on observations
- $I$ : overall performance index
- Subscripts
  - $v$ : variable
  - $m$ : model
  - $n$ : grid point

# The composite indices

- Vertical lines – individual model performance
- Dashed lines – average performance
- X – Performance of the NCEP-NCAR reanalysis
- Solid circles – Performance of the multi-model ensemble mean



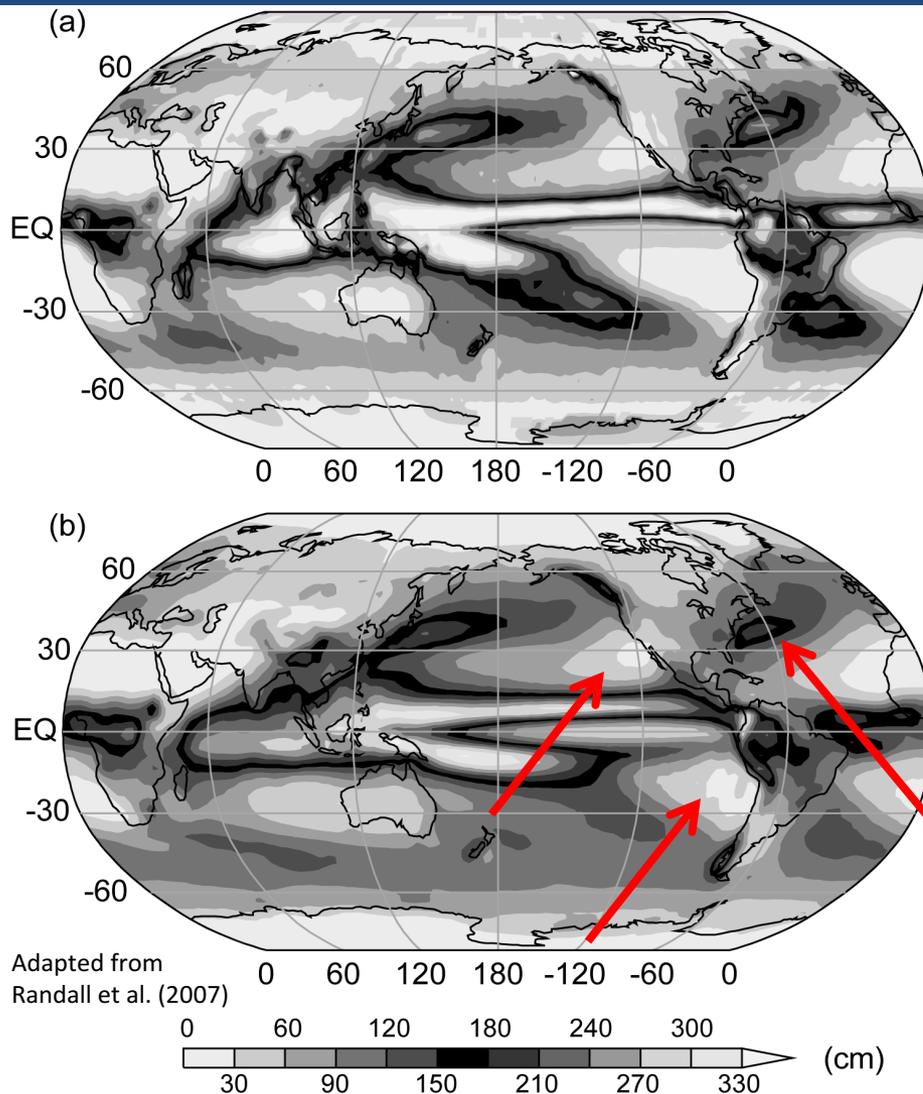


Adapted from Hegerl et al. (2007)

Another example –

Qualitative verification  
of ensemble-mean  
global-average near-  
surface temperature for  
the last century

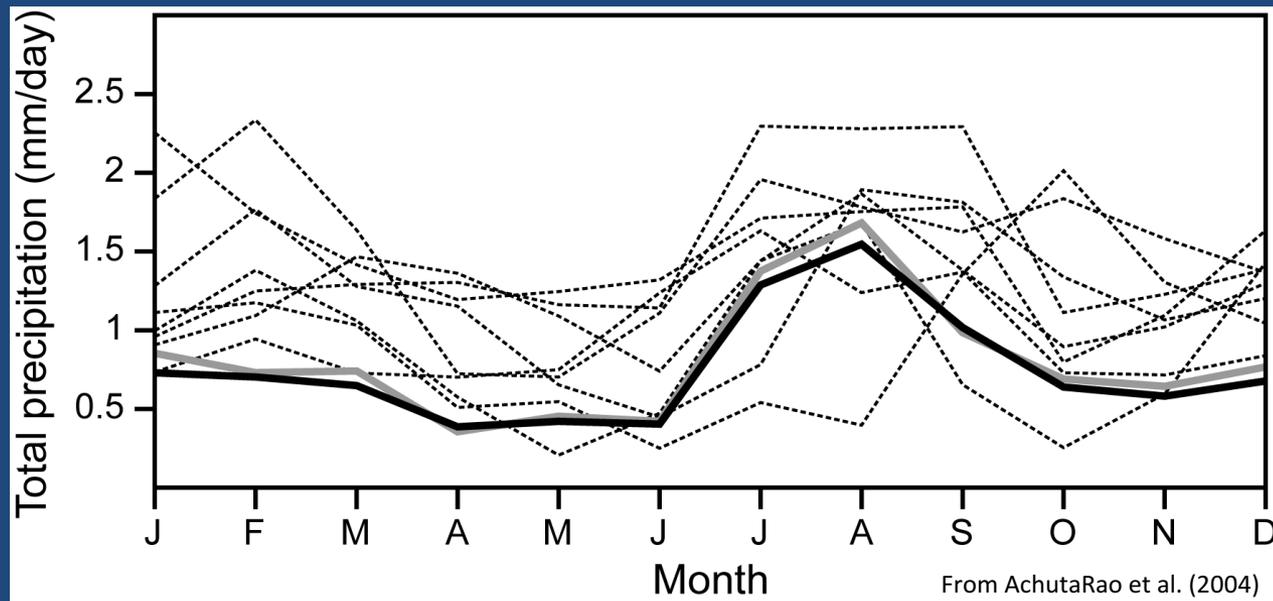
# Verification of specific processes and regional features – current climate



- Annual-mean precipitation
  - a) analysis of obs
  - b) CMIP3 simulation

# An example of a quantitative regional verification

- CMIP2 comparison
- Precipitation in the southwestern US
- Overall annual cycle is captured, but there is much variation from model-to-model.
- Heavy lines – analyses of precipitation



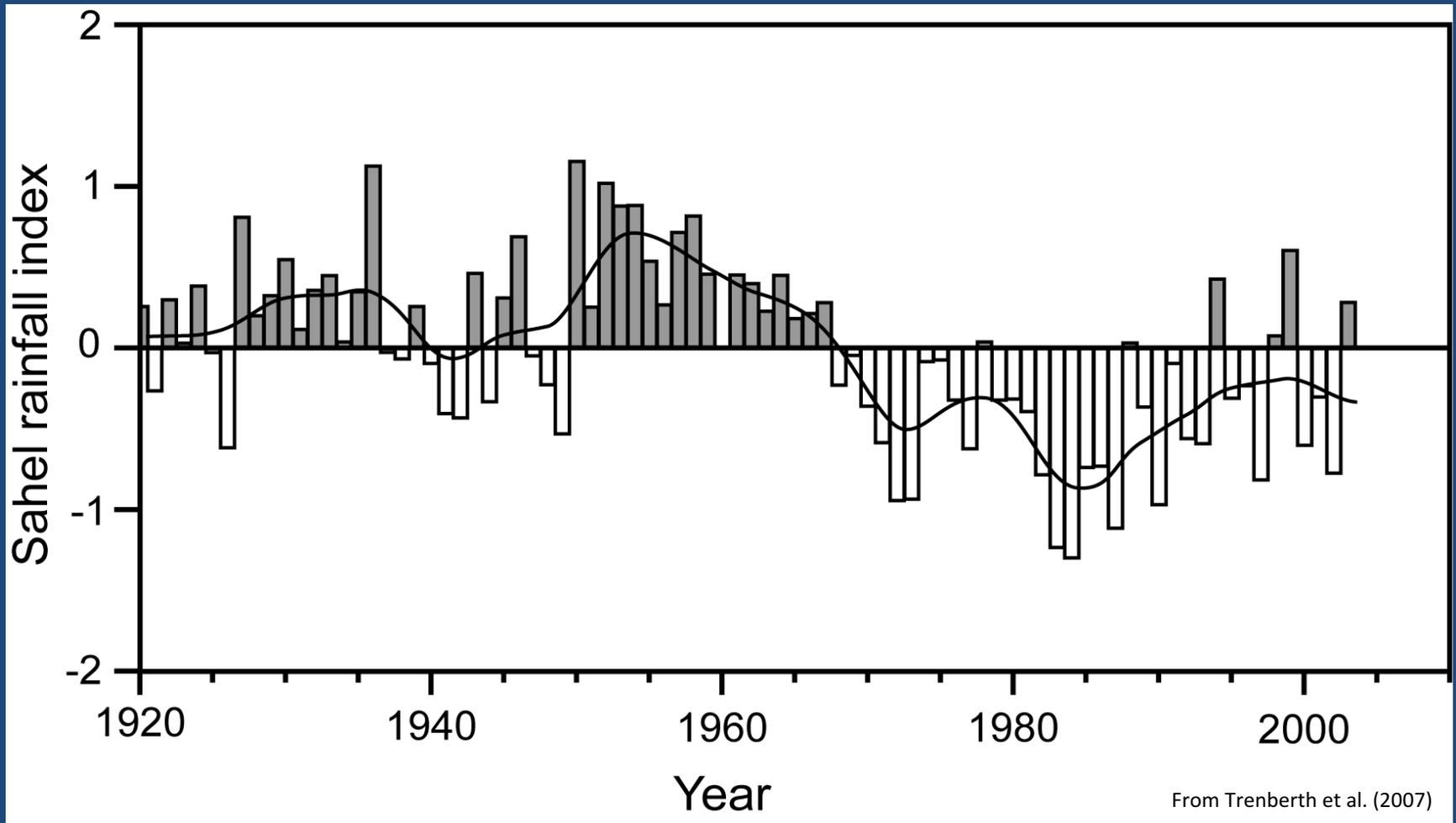
# Modeling extreme values in new climate regimes

- These are often more important than mean values.
- Heavy rains, temperature, wind, drought, stronger storms in general.
- European project – Modeling the Impact of Climate Extremes (MICE)

# Seasonal to multi-year initial-value predictions

- Must forecast anthropogenic impacts and internal variability.

# Internal variability in the Sahel

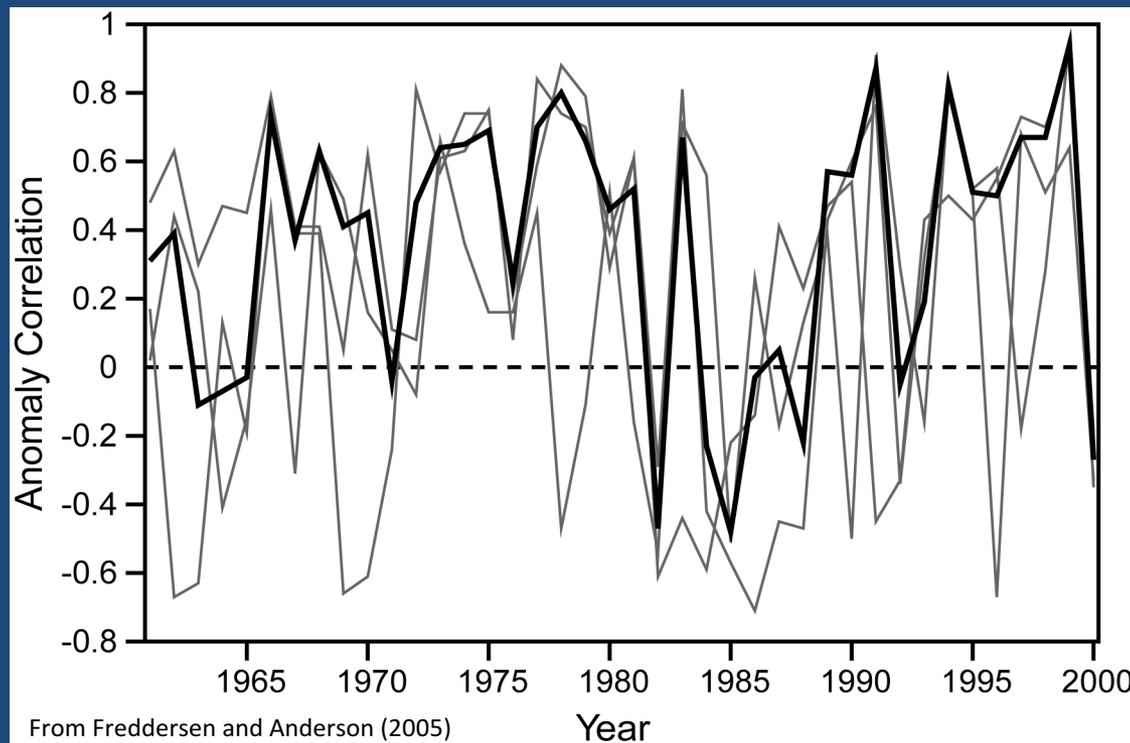


# Seasonal to multi-year initial-value predictions

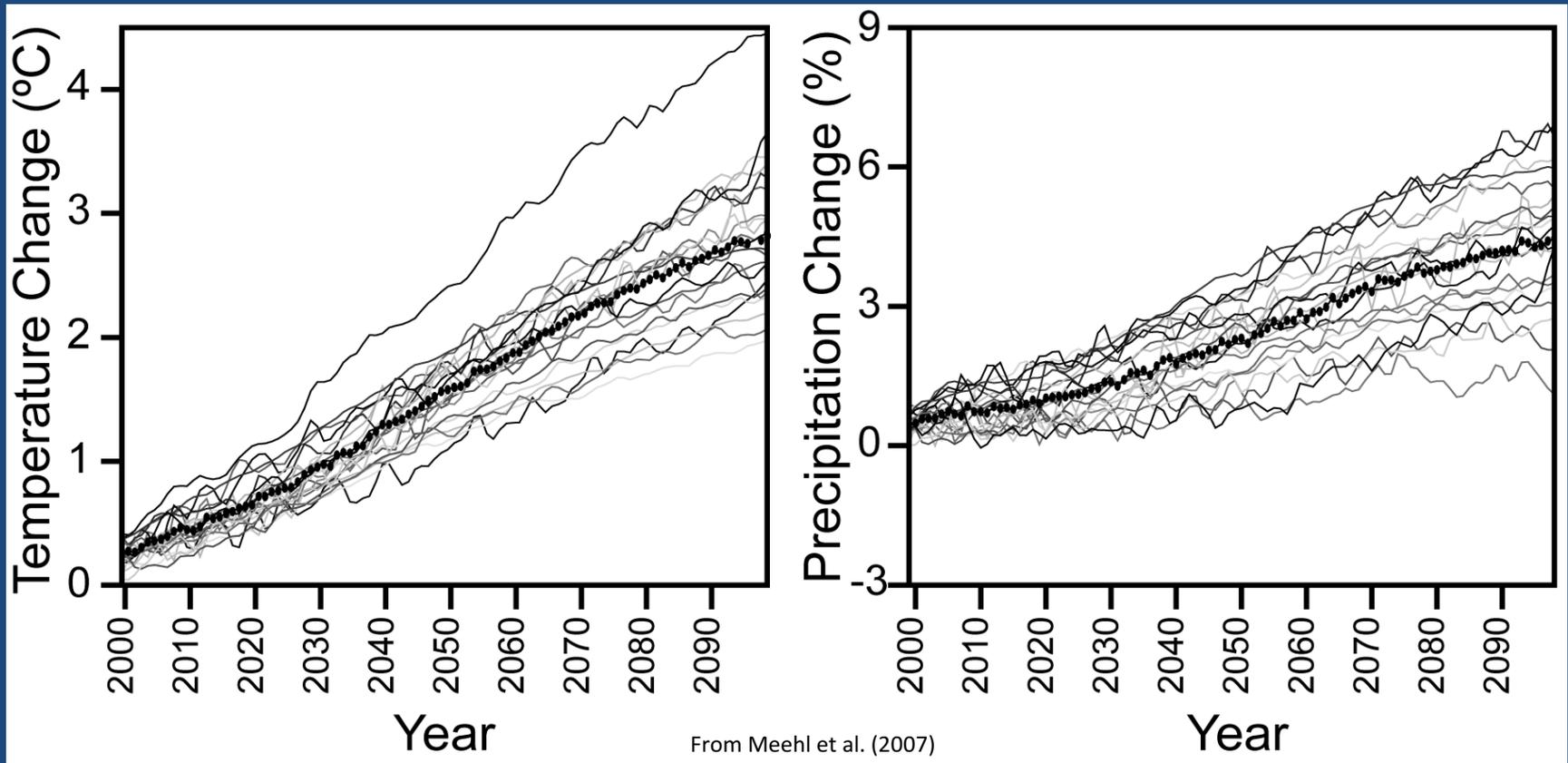
- Must forecast anthropogenic impacts and internal variability.
- Major modeling requirements
  - Accurate models of the entire physical system: atmosphere, ocean, ice, land
  - Initial conditions for the atmosphere, ocean, ice, land (using data assimilation)
  - Method for ensemble prediction
  - Method for correcting for the systematic error

# Ensemble climate simulation - example

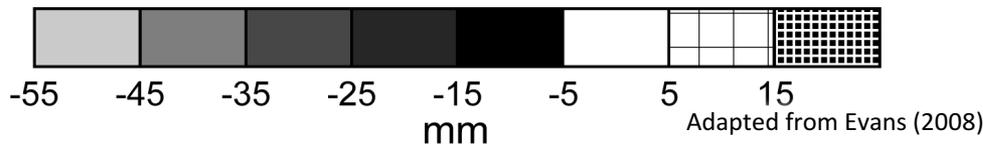
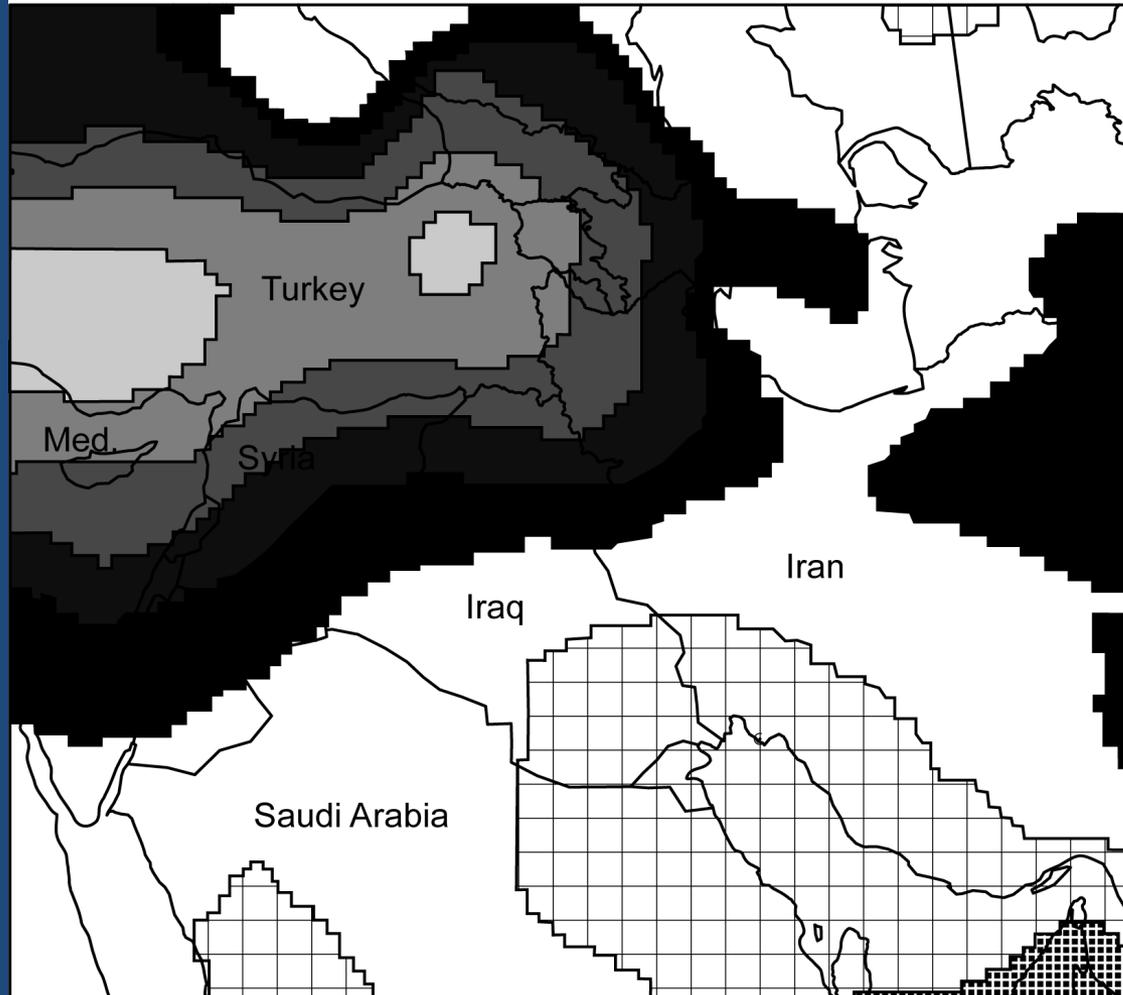
- 2-month predictions
- 40-years of forecasts
- Europe: July, August, September
- Three models – gray lines



# IPCC AR4 ensemble



2050 - 2005



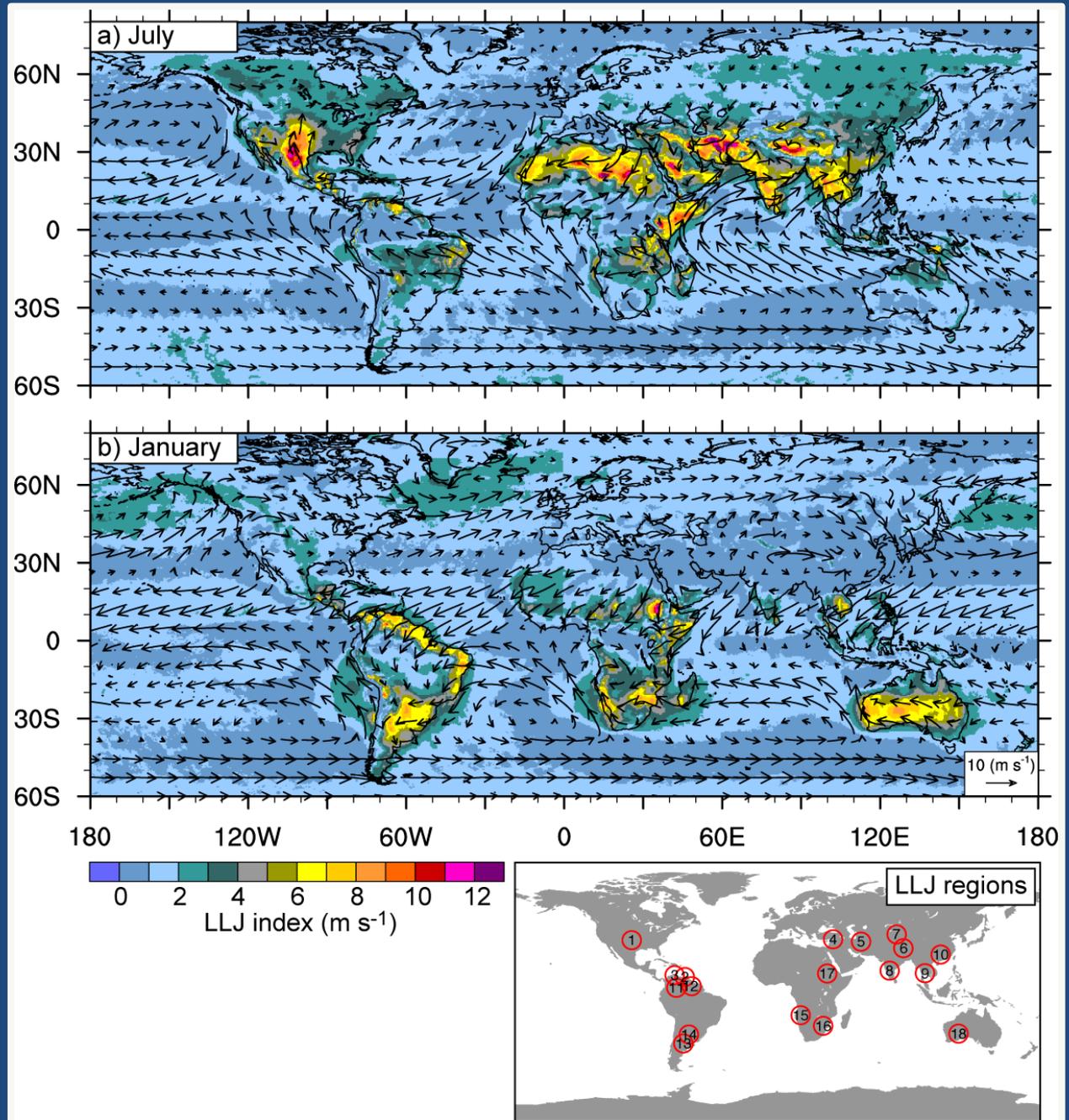
Global AR4  
ensemble  
prediction –  
local focus

# Reanalyses of the current global climate

- Generated and archived by ECMWF, NCEP, etc.
- Specialized reanalyses – for example

# High-temporal-frequency output

Arrows = Mean 500- m-AGL winds at local midnight, plotted every 20<sup>th</sup> grid point.

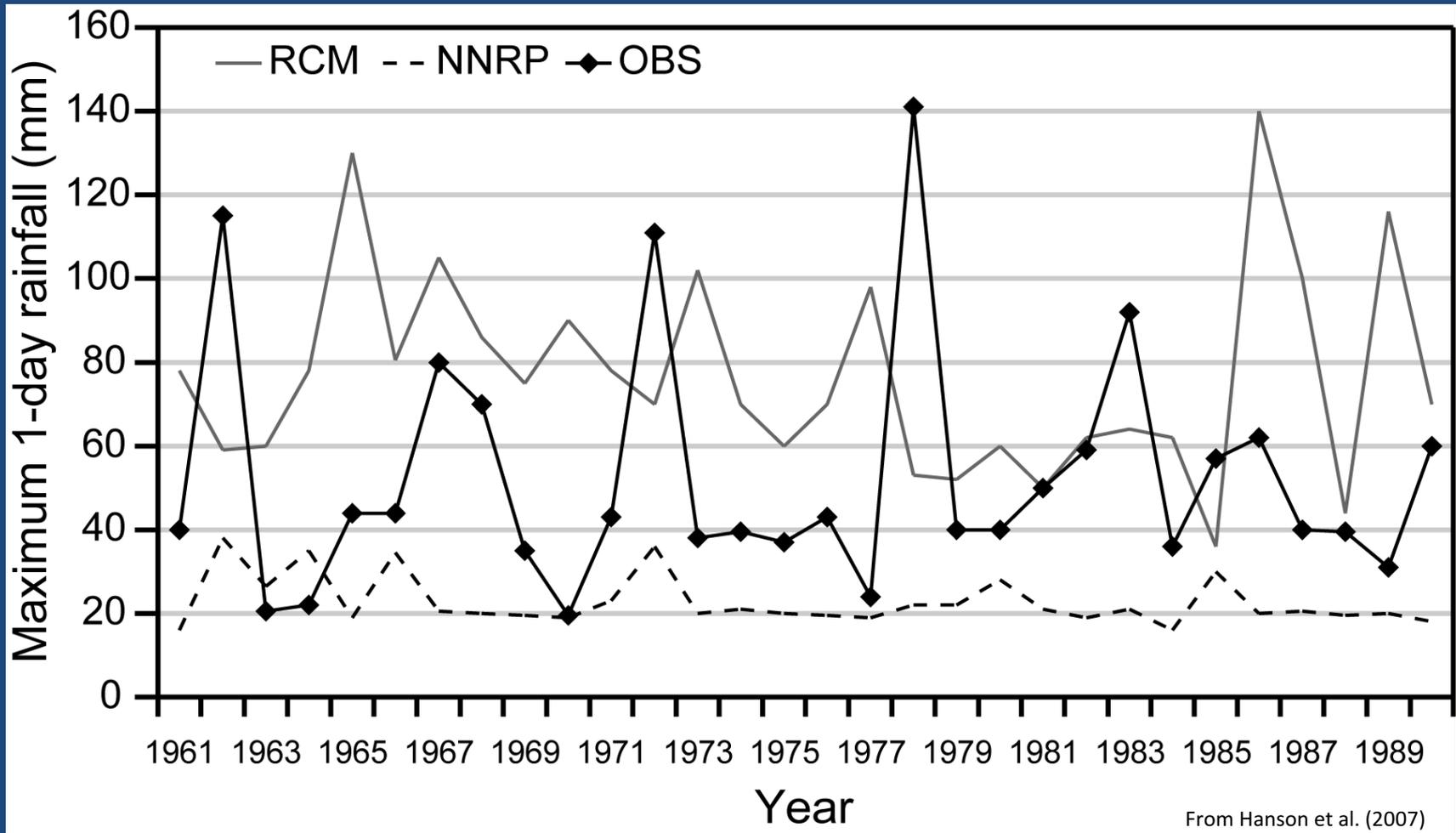


Rife et al. (2010)  
Monaghan et al. (2010)

# Climate downscaling

- Approaches
  - Dynamical and statistical
  - Spatial and temporal
  - Current climate and future climate

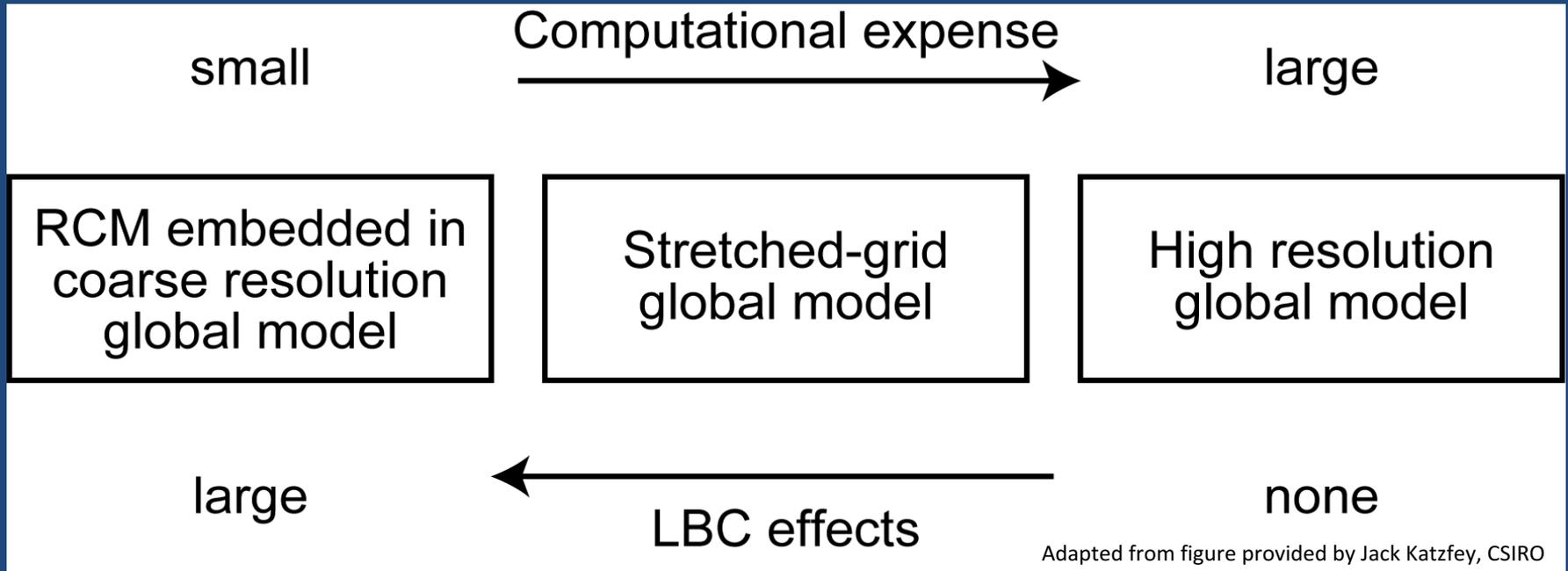
# Example of the benefit of downscaling: Current-climate, spatial, dynamic



# Dynamical climate downscaling methods

- Limited-area models (Regional Climate Models – RCMs) – define LBCs from AOGCM forecasts, or from reanalyses.
- Global stretched-grid AGCMs (shown before)
- Uniformly high-resolution AGCMs

# LBC effects for the different approaches

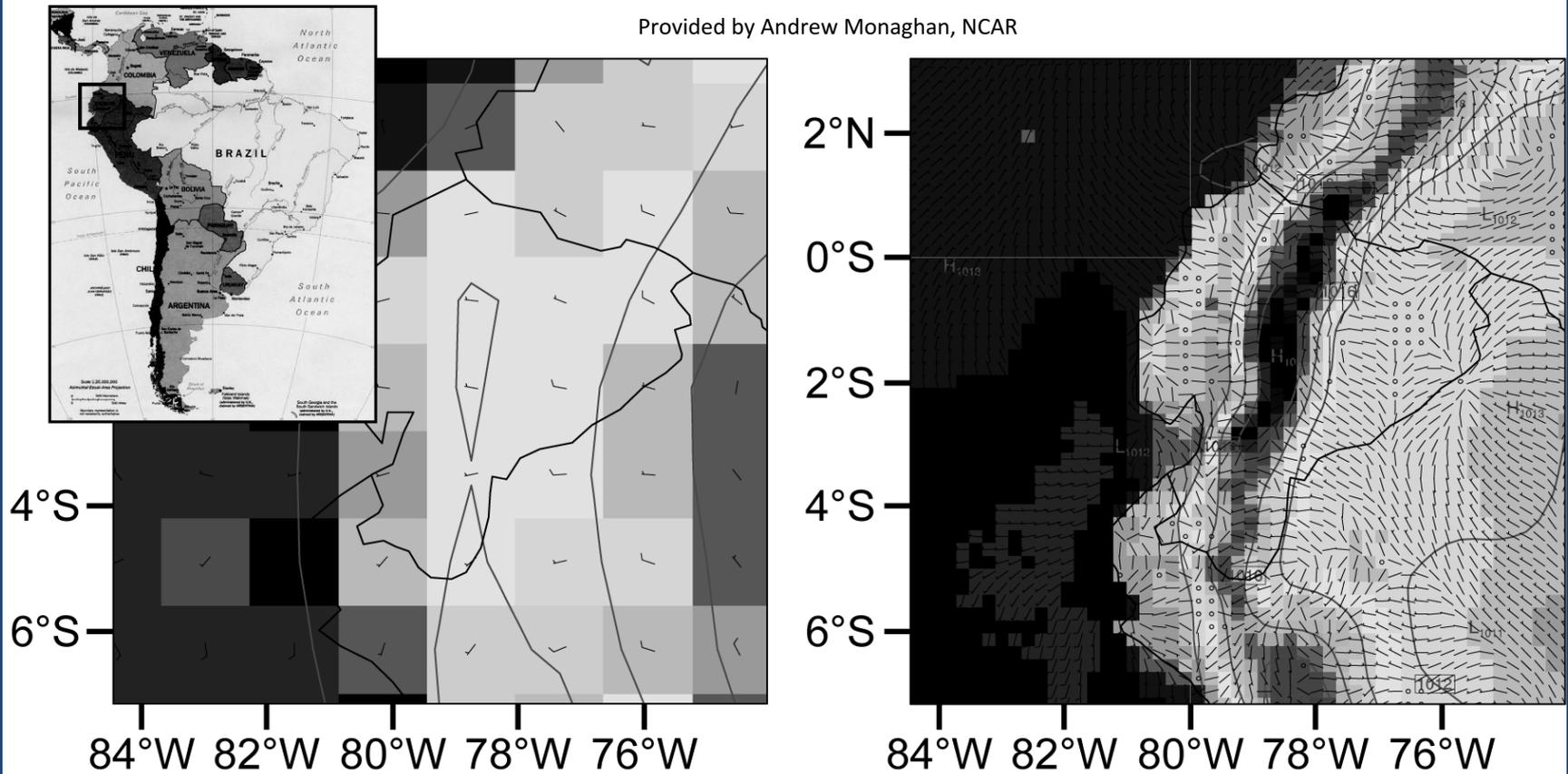


# Example of future-climate downscaling with RCM

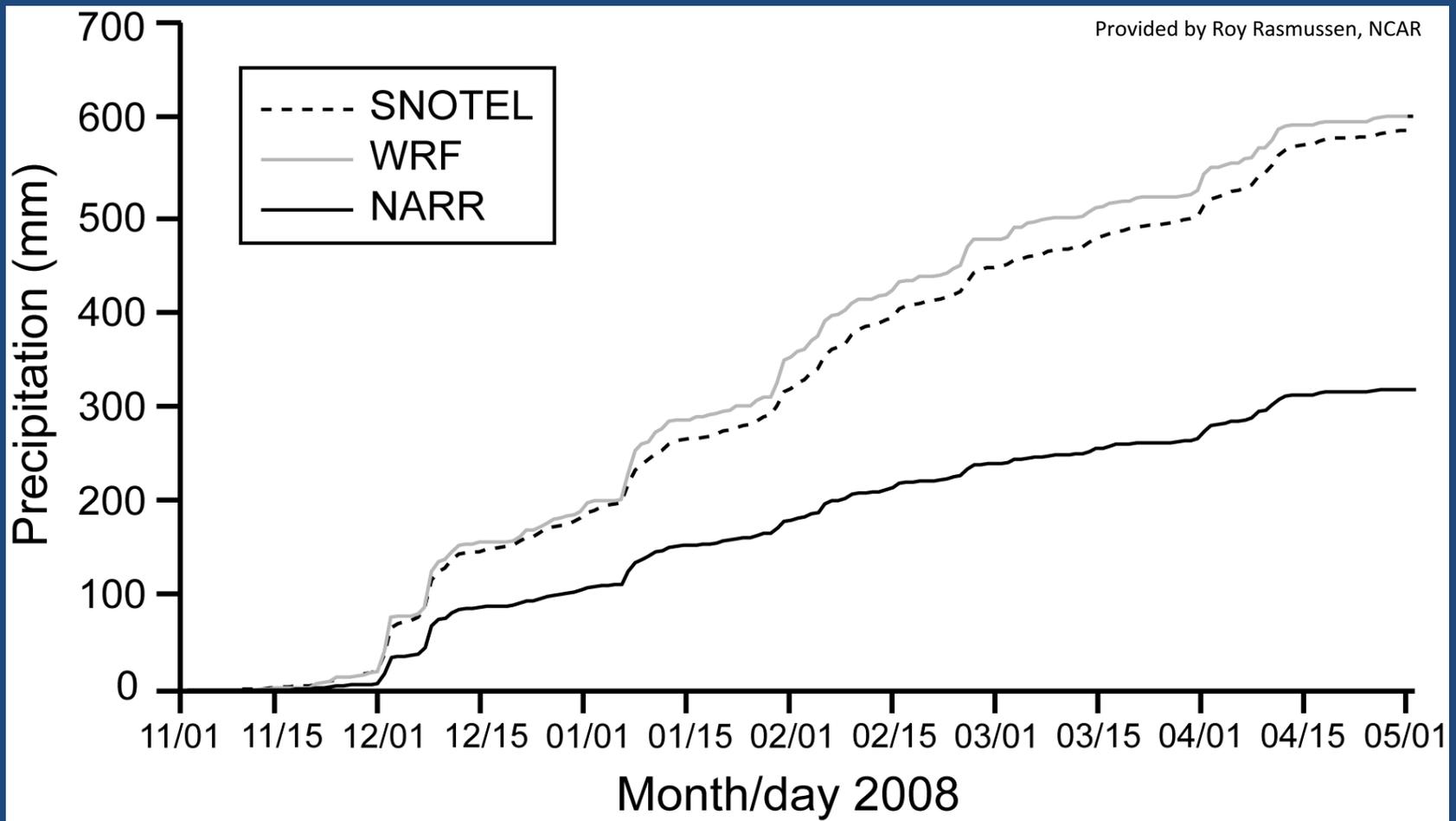
CCSM3

WRF

Provided by Andrew Monaghan, NCAR

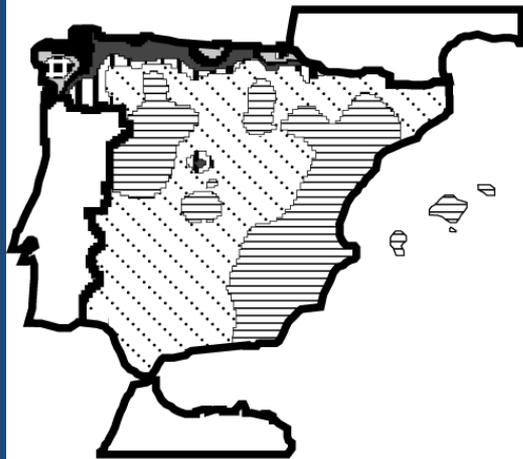


# Example of current-climate downscaling with a LAM

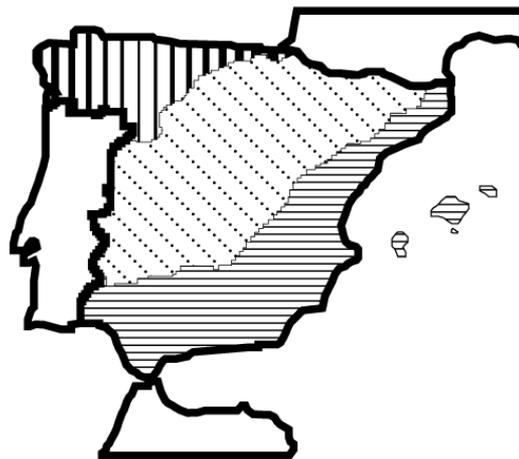


# An example of the benefit of statistical downscaling for February-April precipitation

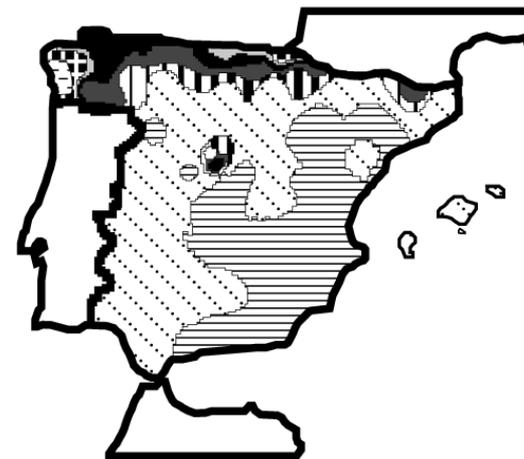
(a) Observations



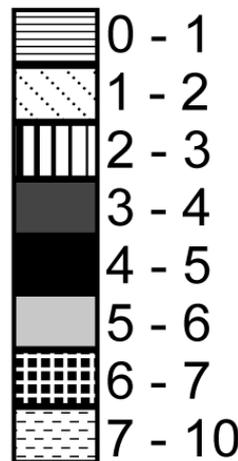
(b) Global ensemble



(c) Analog-based downscaling



mm/day



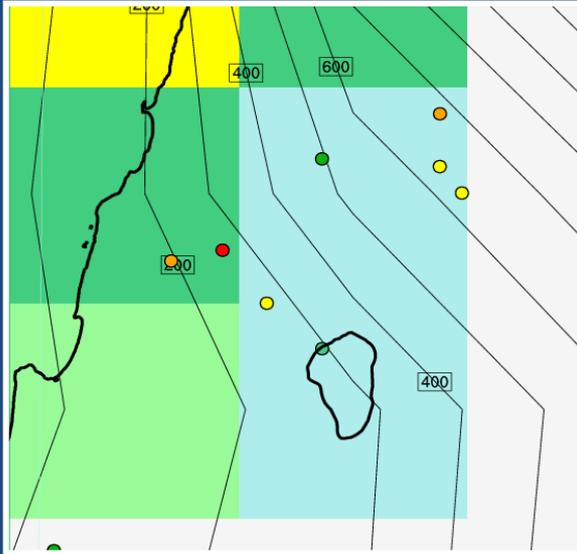
Adapted from Palmer et al. (2004)

# Benefit of high resolution for current-climate downscaling

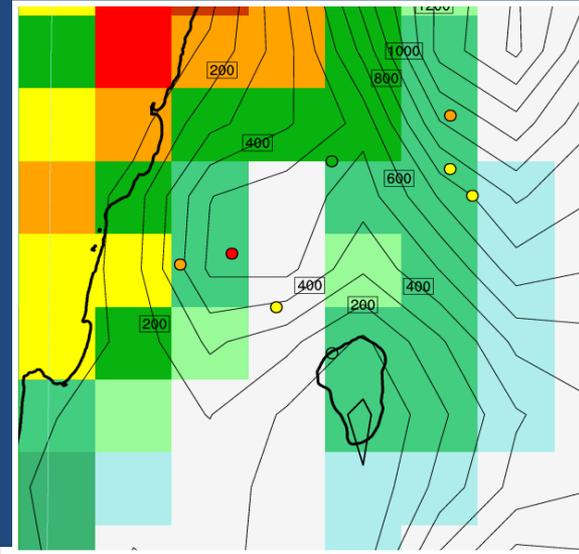
(Total accumulation of 10 Nov 2008 – 22 Mar 2009)

Grid 1  
DX:  
40.5km

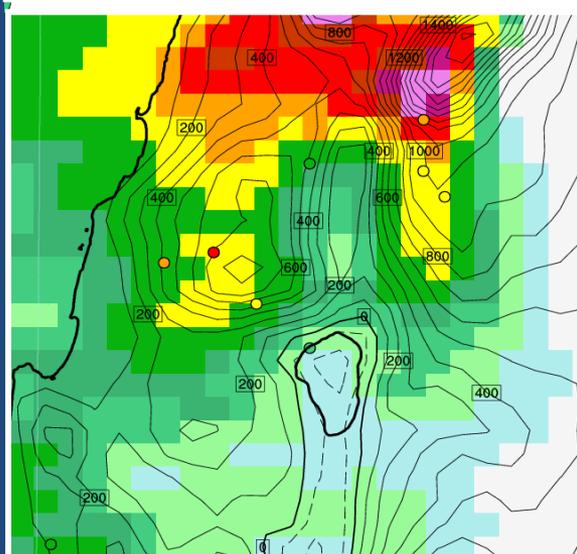
100x100  
km<sup>2</sup>



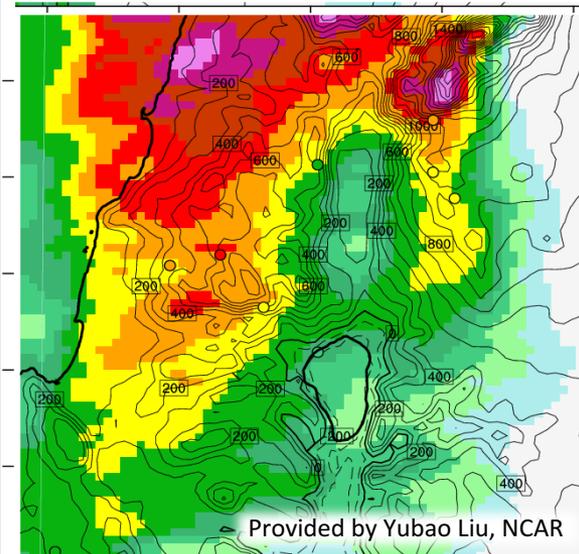
Grid 2  
DX:  
13.5km



Grid 3  
DX:4.5km



Grid 4  
DX:1.5km



Accumulated rainfall (mm)



Provided by Yubao Liu, NCAR

# Modeling the climate impacts of anthropogenic landscape changes – sensitivity studies

- Urbanization
- Deforestation and advancement of agriculture
- Irrigation
- Drying of lakes