

CORDEX and NARCCAP: Foundations in Reanalyses

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**Coordinated Regional Downscaling
Experiment (CORDEX)
North American Regional Climate Change
Assessment Program
(NARCCAP)**

1. Overview of NARCCAP & CORDEX
2. General results – multiple regions
3. Sensitivity to reanalysis
4. Summary

NARCCAP



MM5

Iowa State/
PNNL

RegCM3

UC Santa Cruz
ICTP

CRCM

Quebec,
Ouranos

HADRM3

Hadley Centre

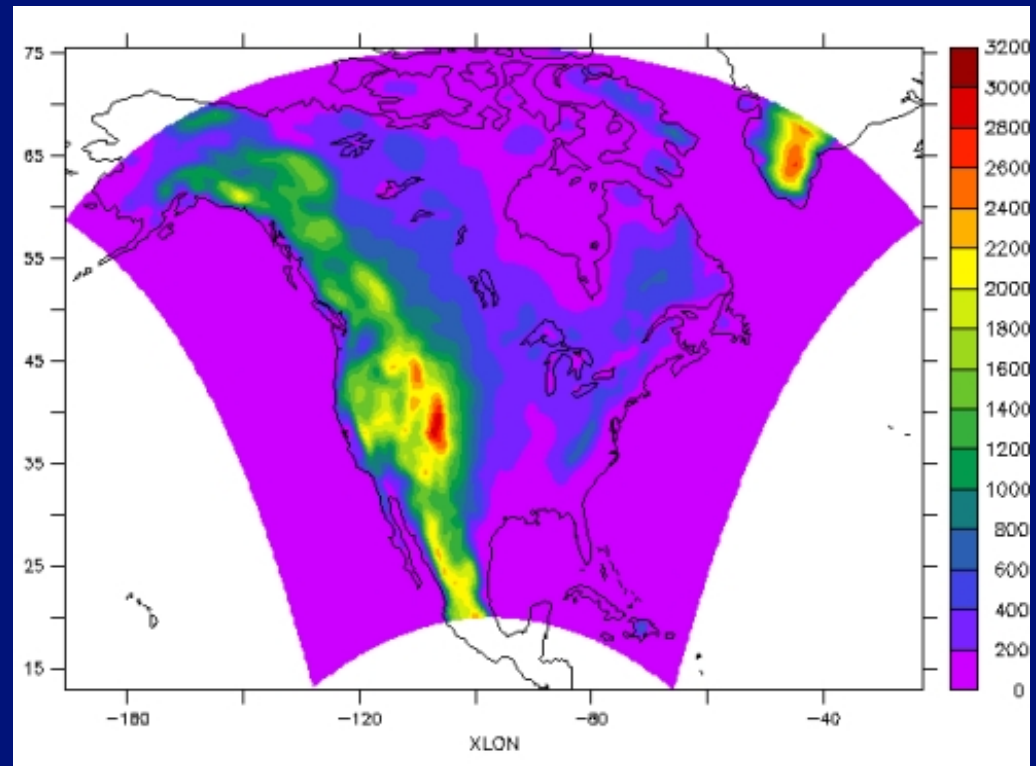
RSM

Scripps

WRF

NCAR/
PNNL

- Domain
 - Most of North America
- Period
 - 1980-2004
- Boundary Conditions
 - NCEP/DOE reanalysis
- Resolution
 - 50 km



CORDEX Phase I experiment design

Model Evaluation Framework

Climate Projection Framework

Multiple regions (Initial focus on Africa)
50 km grid spacing

ERA-Interim BC
1989-2007

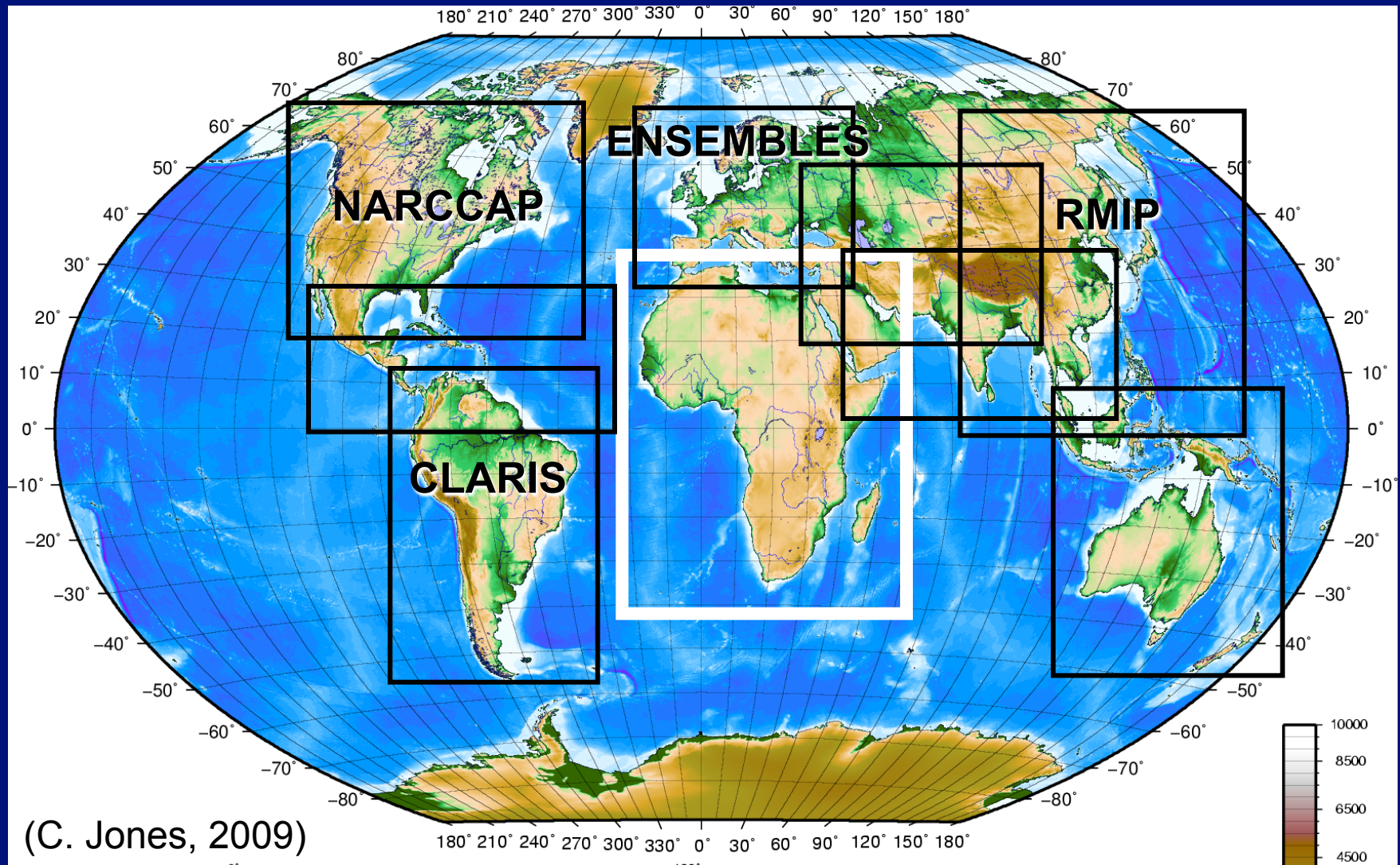
RCP4.5, RCP8.5

Multiple AOGCMs

Regional Analysis
Regional Databanks

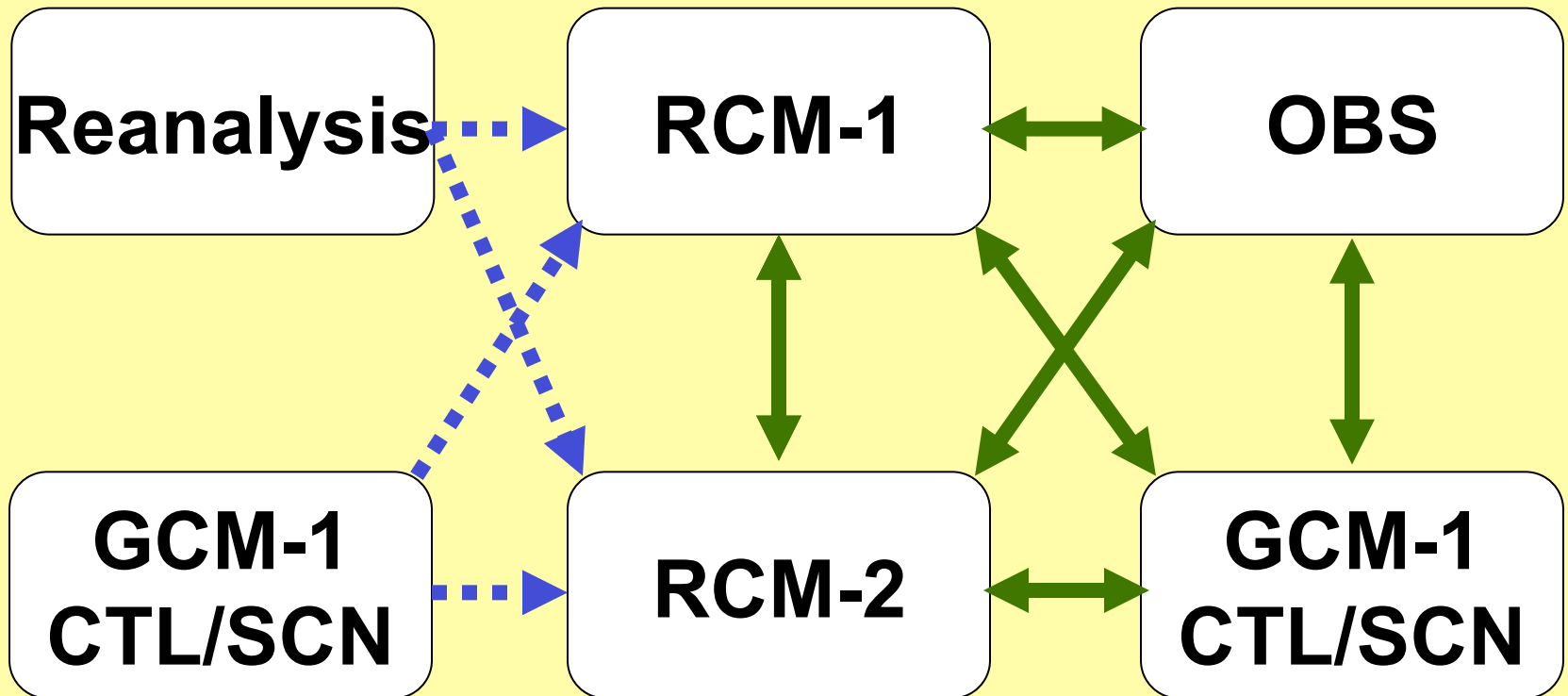
1951-2100
1981-2010, 2041-2070, 2011-2040

Coordinated Regional Downscaling Experiment (CORDEX) ~ Regions ~



Possible Comparisons?

(Pan et al. 2001)



---> Driving

↔ Differences

- A simple spatial ANOVA model:

$$\begin{array}{l} \text{NCEP } \mathbf{X}_{0k} = \boldsymbol{\mu} + \boldsymbol{\epsilon}_{0k} \\ \text{RCM}_i \mathbf{X}_{ik} = \boldsymbol{\mu} + \boldsymbol{\alpha}_i + \boldsymbol{\epsilon}_{ik} \end{array} \quad i = 1, \dots, 5, k = 1, \dots, 20$$

- $\boldsymbol{\mu}$ represents common effect
- $\{\boldsymbol{\alpha}_i\}$ represents model-specific effects
- Spatial priors for $\boldsymbol{\mu}$, $\{\boldsymbol{\alpha}_i\}$, $\{\boldsymbol{\epsilon}_{0k}\}$, $\{\boldsymbol{\epsilon}_{ik}\}$
- Inference focuses on posterior distribution of $\boldsymbol{\mu}$ and $\{\boldsymbol{\alpha}_i\}$.

(S. Sain, 2008, 2012)

South America

7 RCMs using ERA-Interim boundary conditions
(1990-2008)

Temperature Bias ($^{\circ}\text{C}$)

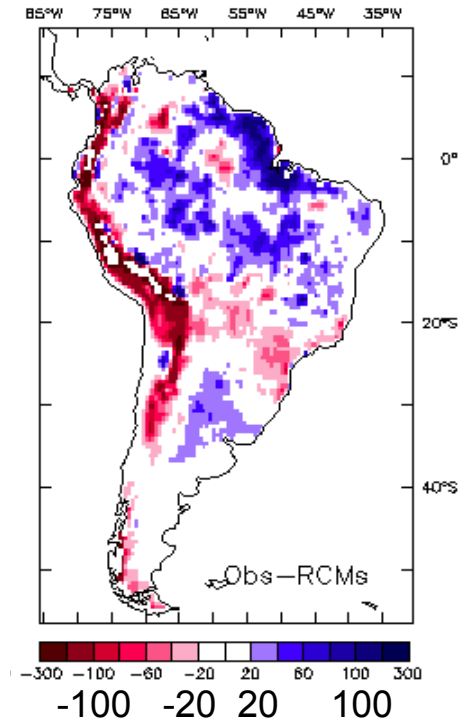
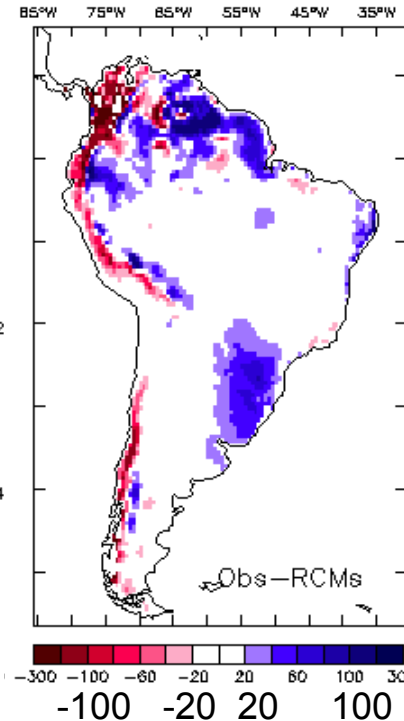
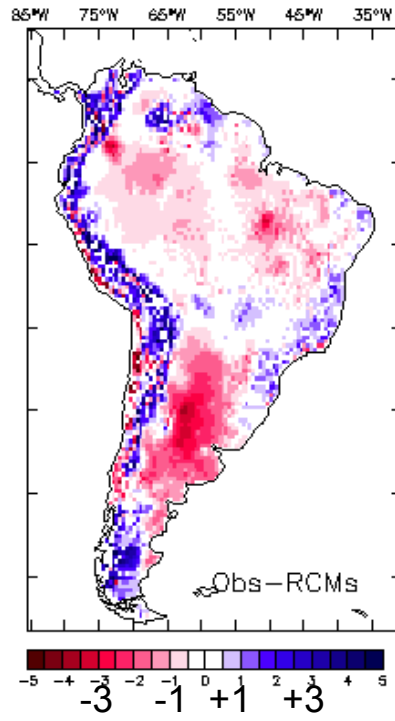
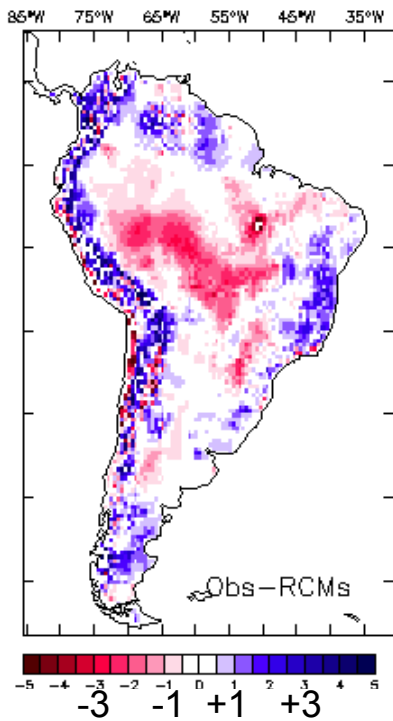
Precipitation Bias (mm/month)

JJA

DJF

JJA

DJF



Ensemble vs. (CRU & UDEL)

Ensemble vs. (GPCC, CRU, UDEL & CPC)

(Solman et al., 2012 - submitted)

South America

7 RCMs using ERA-Interim boundary conditions
(1990-2008)

Intermodel spread
Temperature (°C)

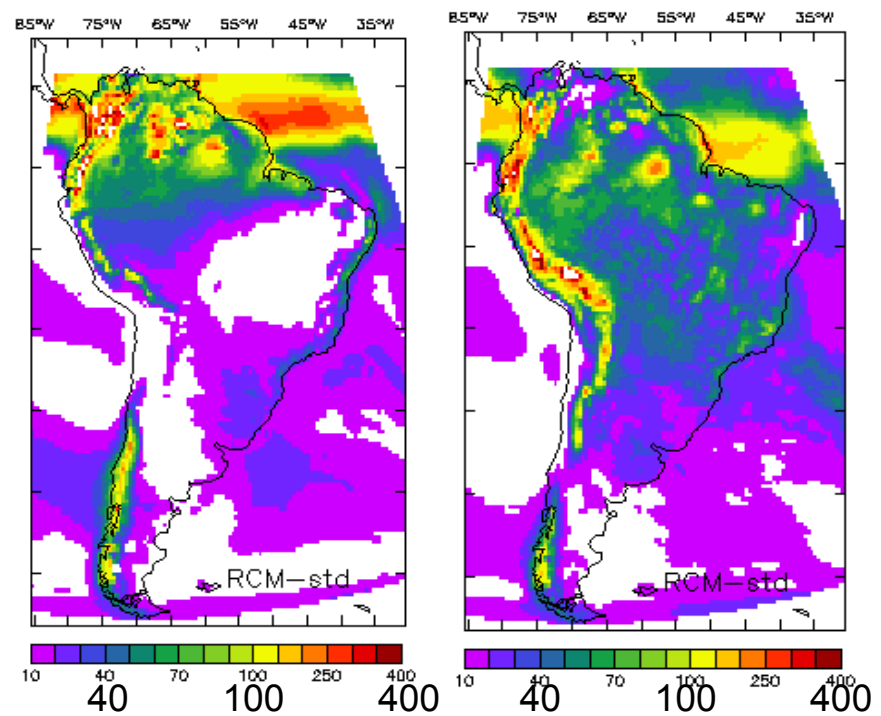
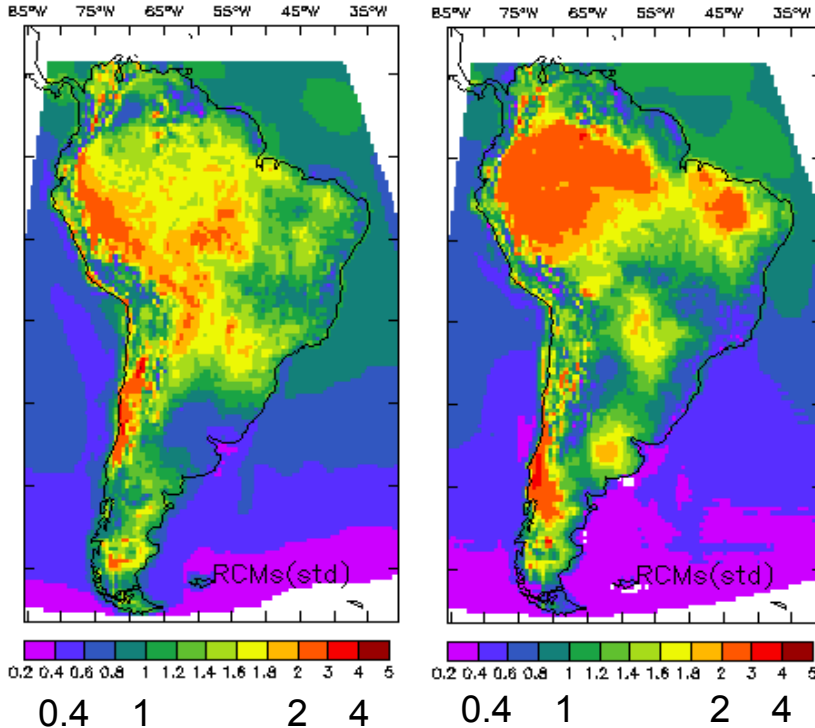
Intermodel spread
Precipitation (mm/month)

JJA

DJF

JJA

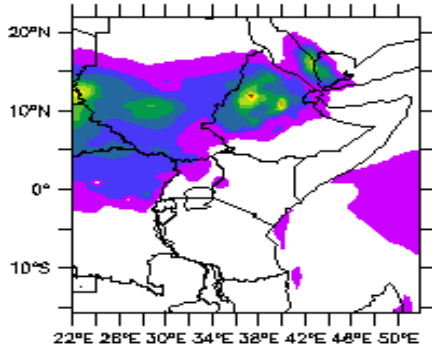
DJF



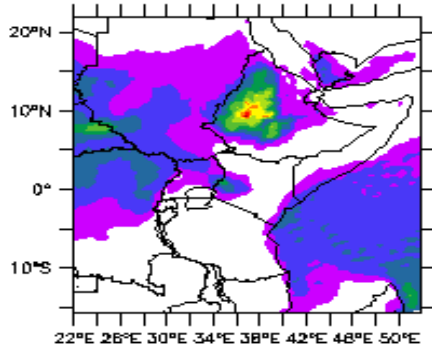
(Solman et al., 2012 - submitted)

CORDEX Africa : East Africa Rainfall (1990-2008 ; JAS)

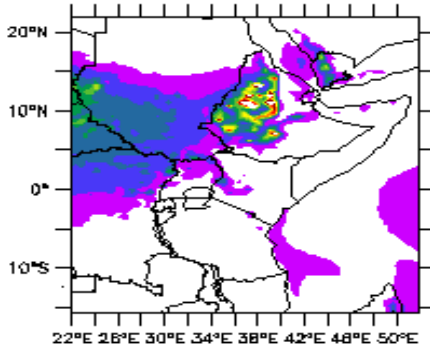
CCLMcom—CCLM



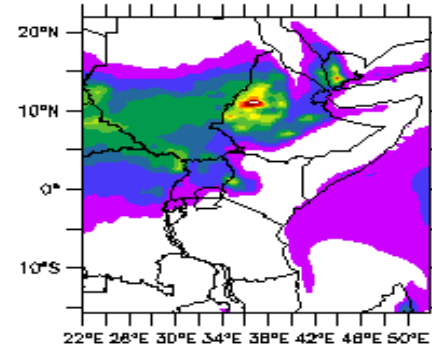
CNRM—ARPEGE



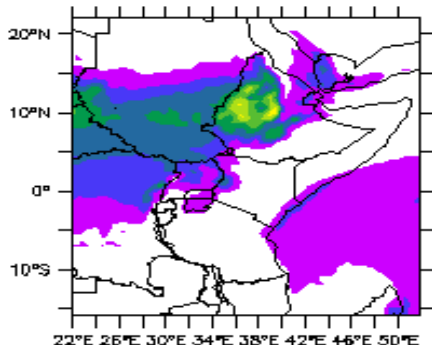
DMI—HIRHAM



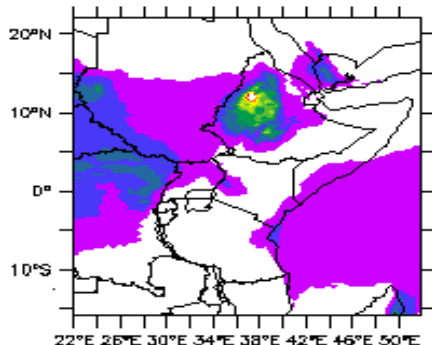
ICTP—RegCM3



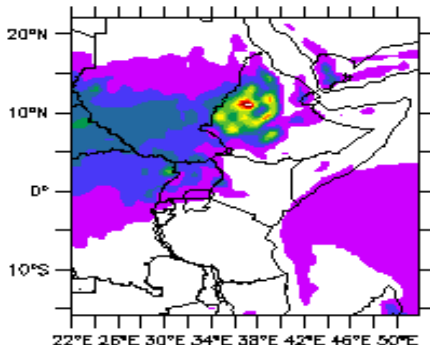
KNMI—RACMO



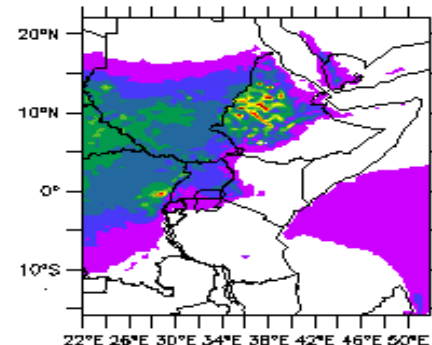
MPI—REMO



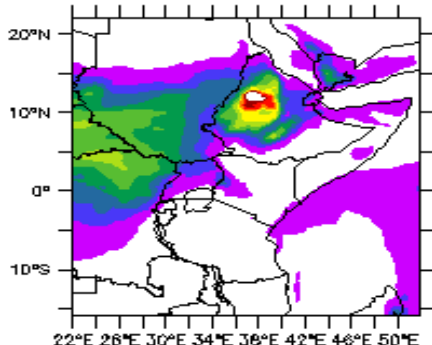
SMHI—RCA35



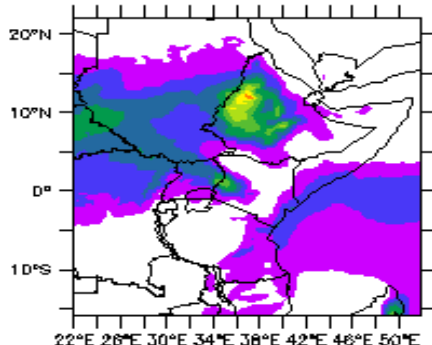
UCT—PRECIS



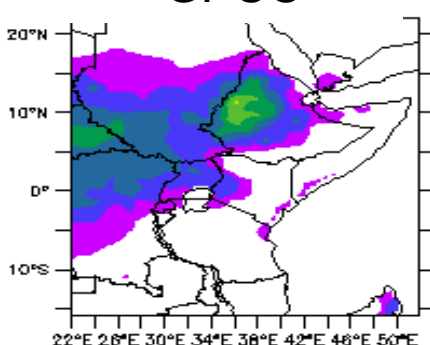
UC—WRF



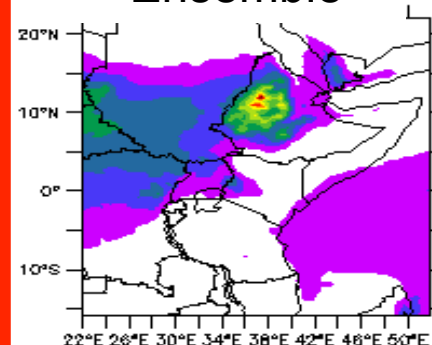
UGAM—CRCM5



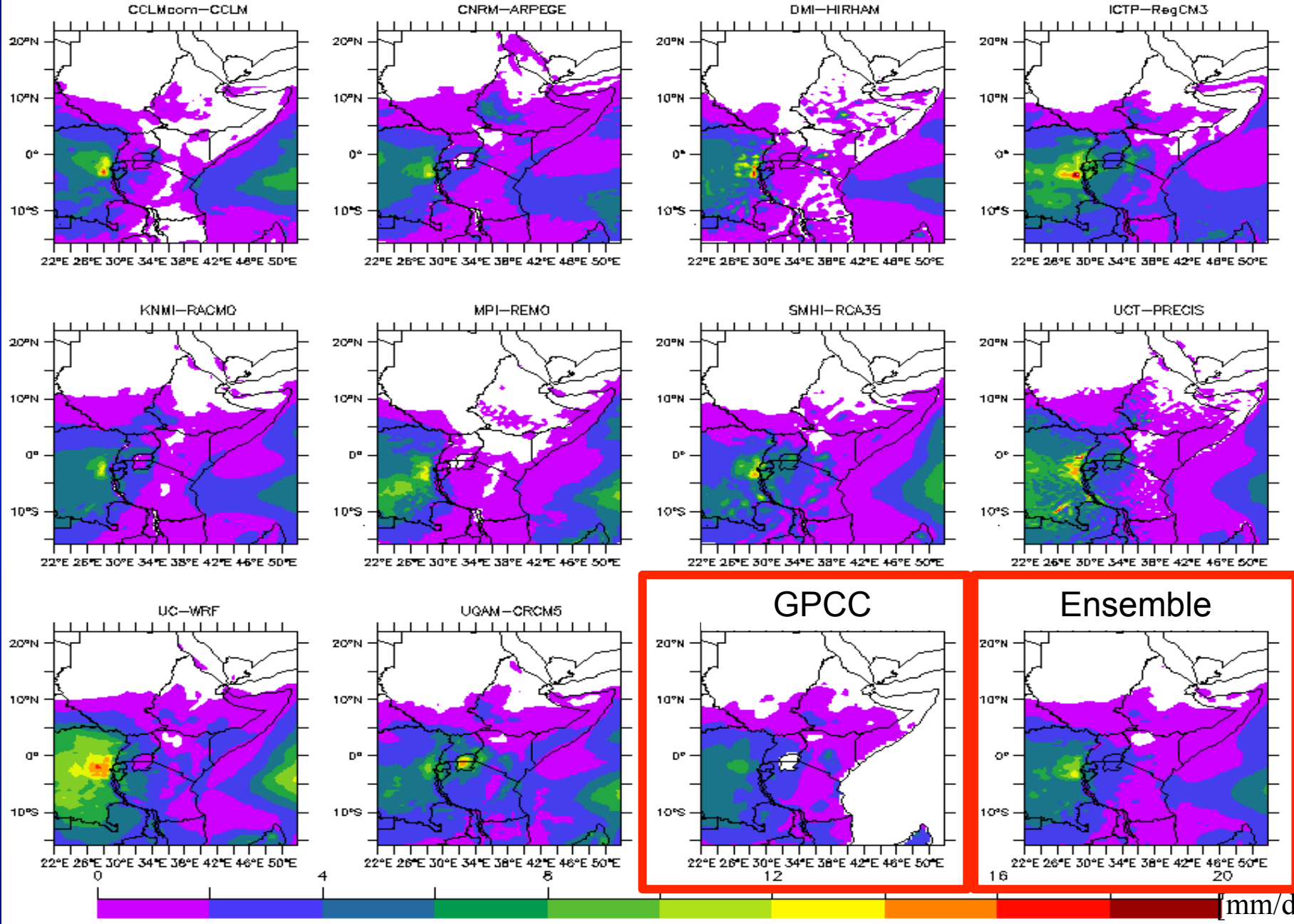
GPCC



Ensemble



CORDEX Africa : East Africa Rainfall (1990-2008 ; OND)



CORDEX

Africa:

10-Yr

Return

Values of

Daily

Precipitation

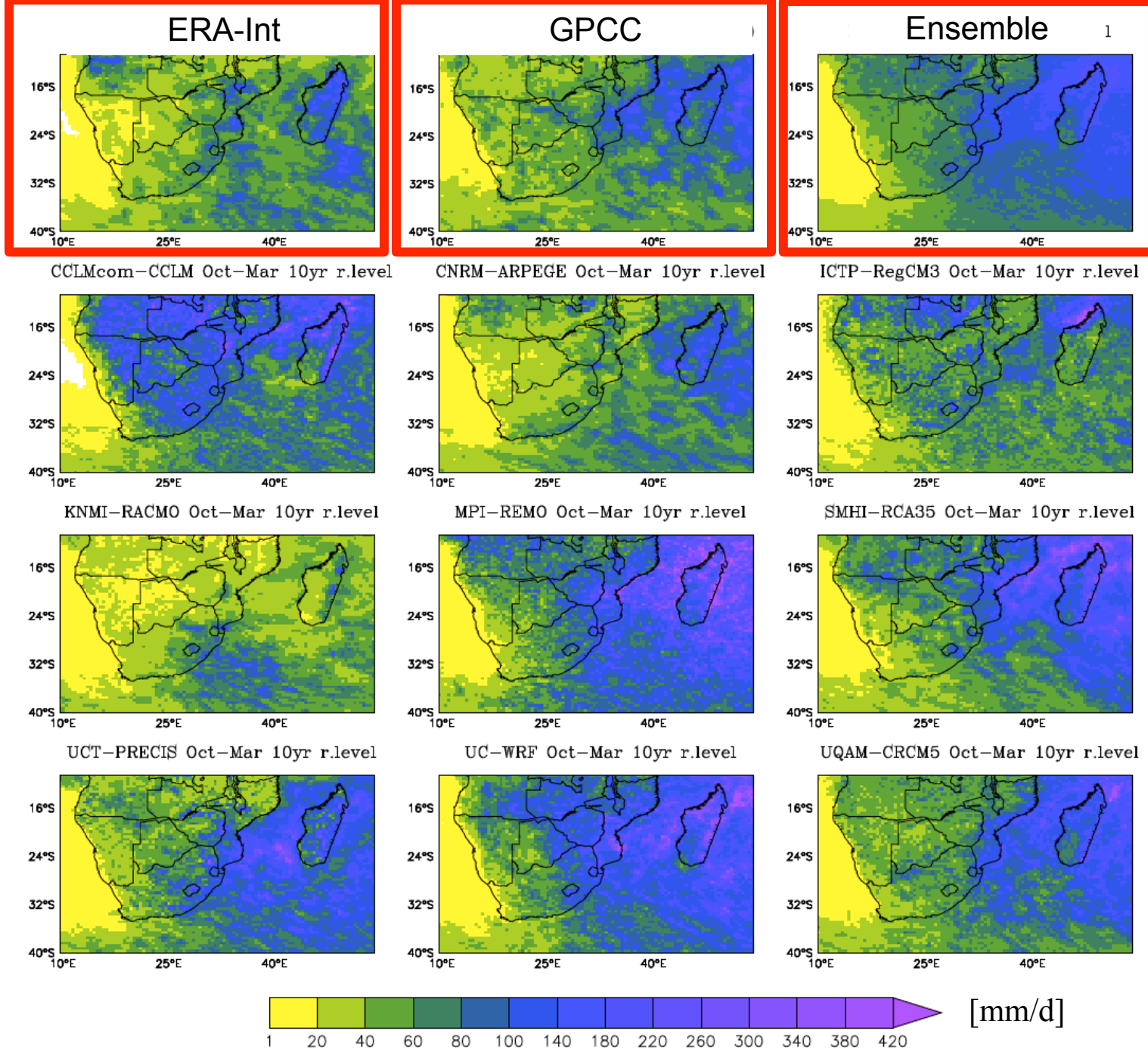
-

Southern

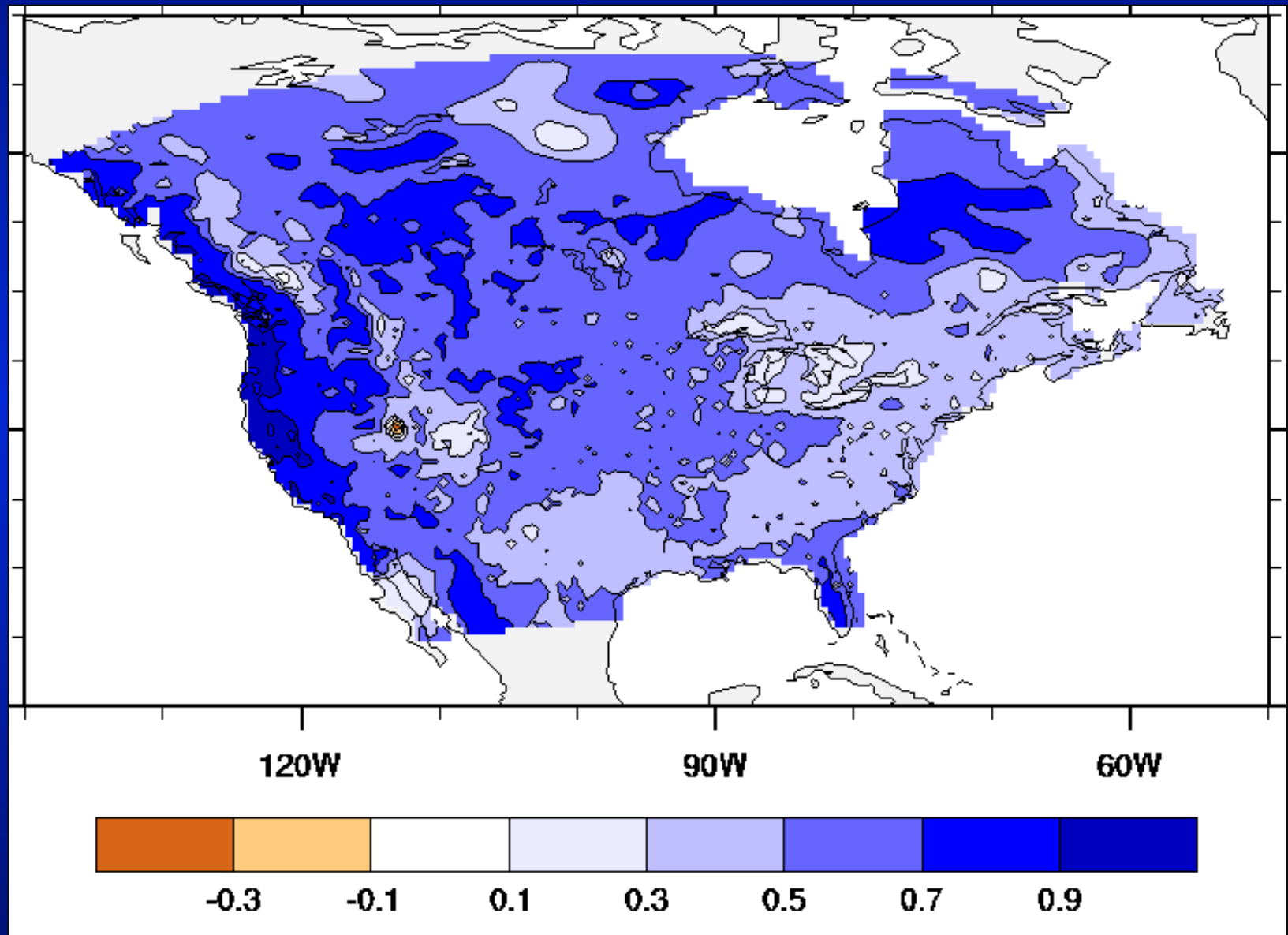
Africa

(1990-2006 ;

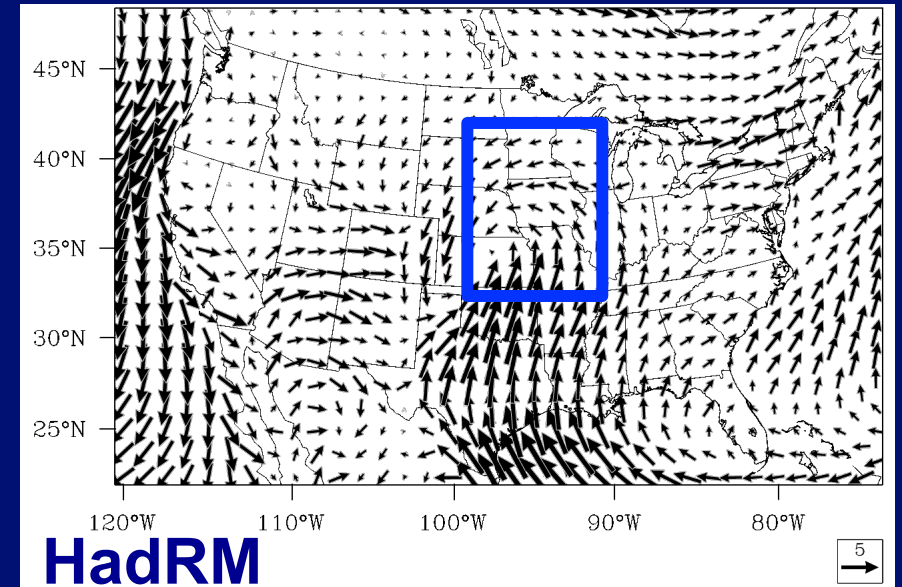
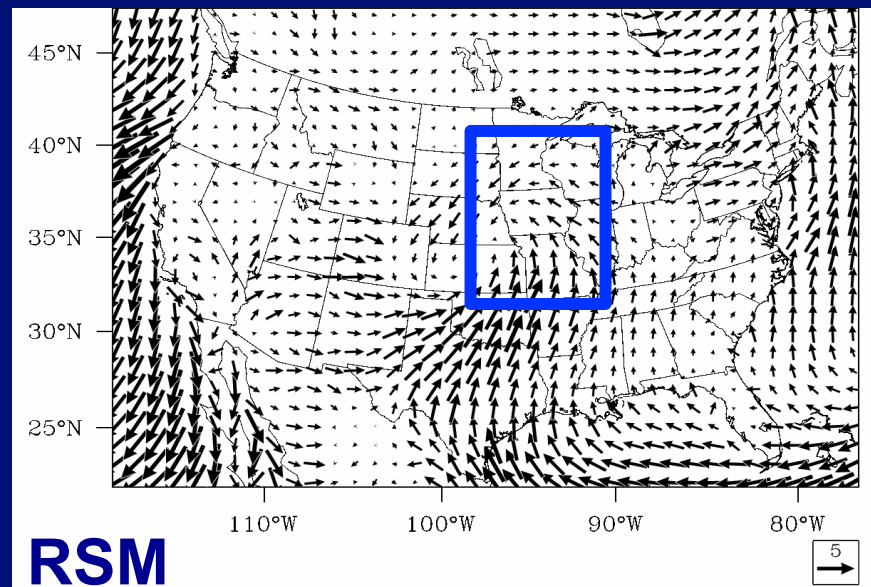
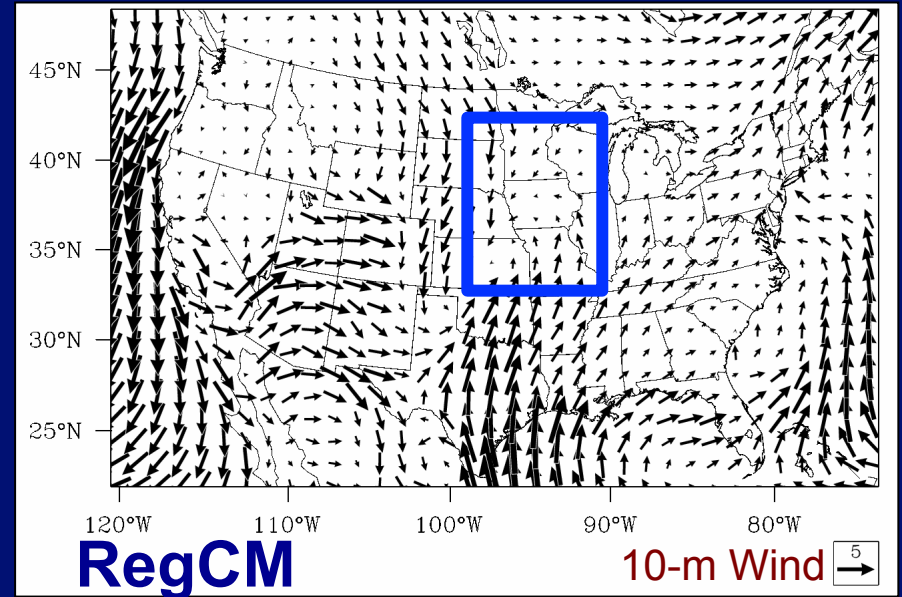
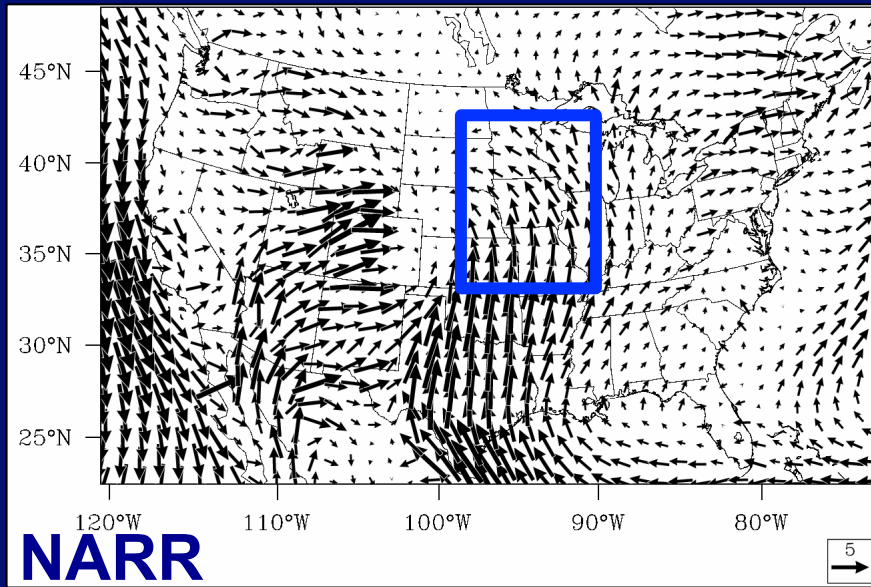
Oct-Mar)



NARCCAP: Correlation of Monthly Precipitation Time Series



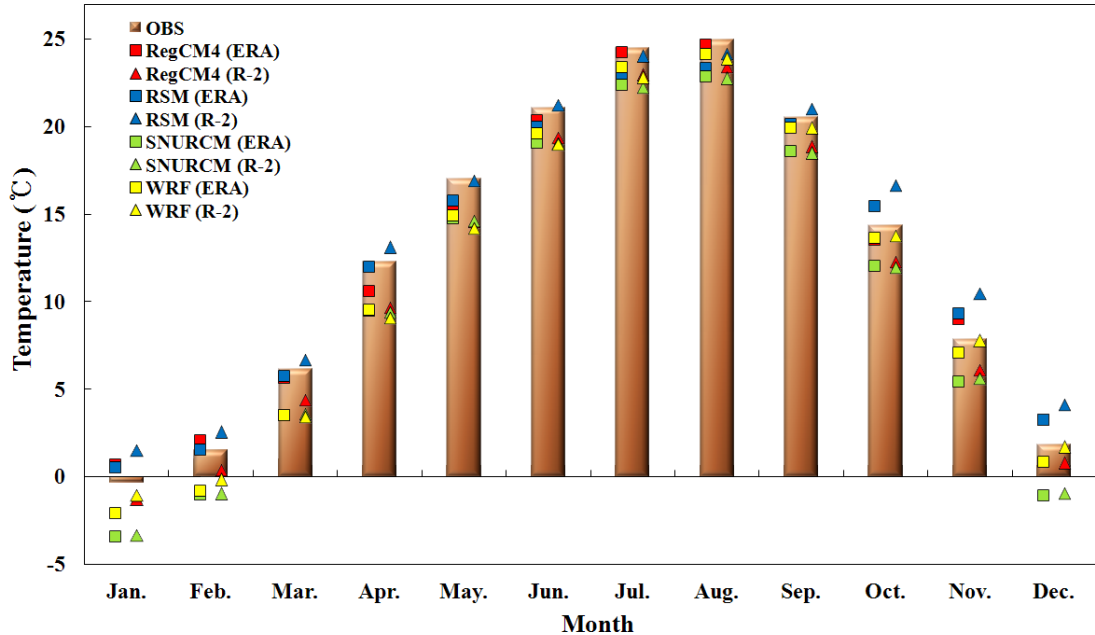
NARCCAP: Composite Structure Extreme Prec Events - JJA



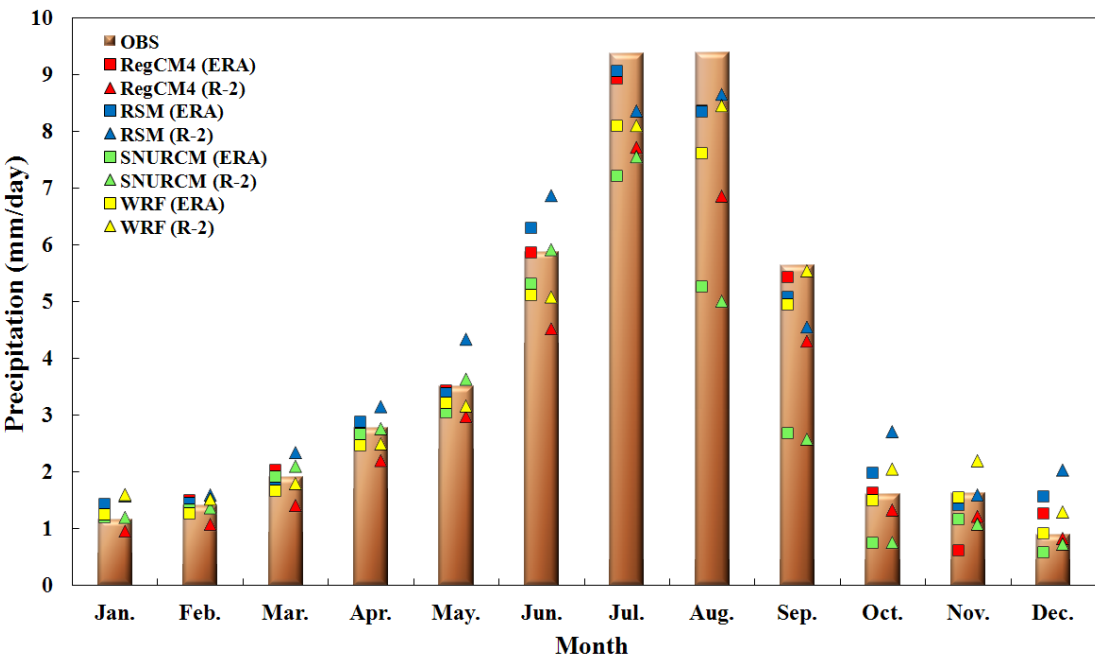
Temperature

CORDEX: East Asia

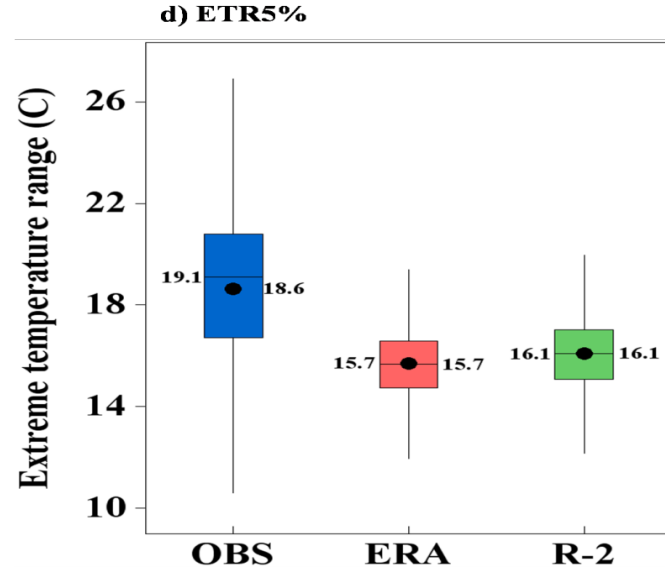
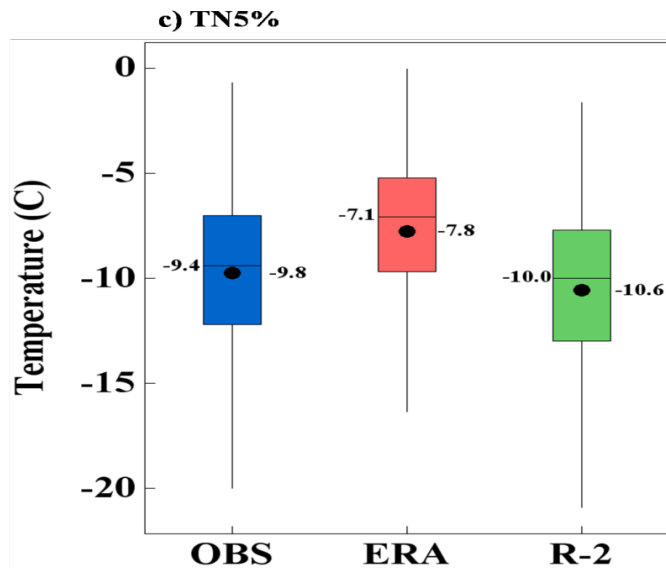
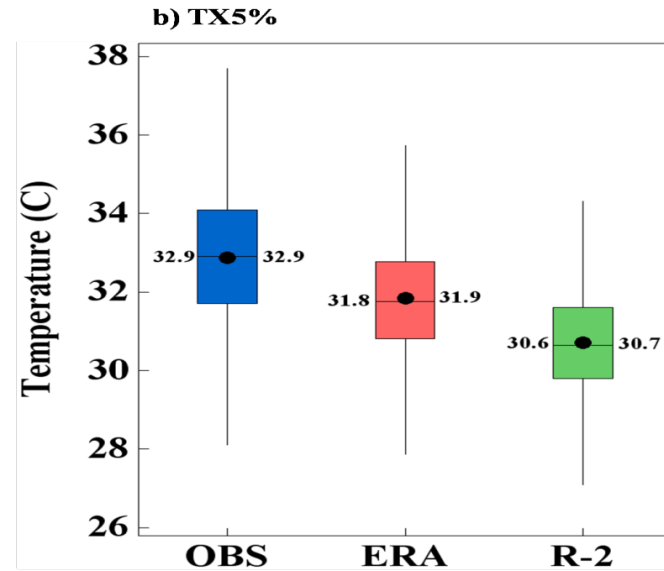
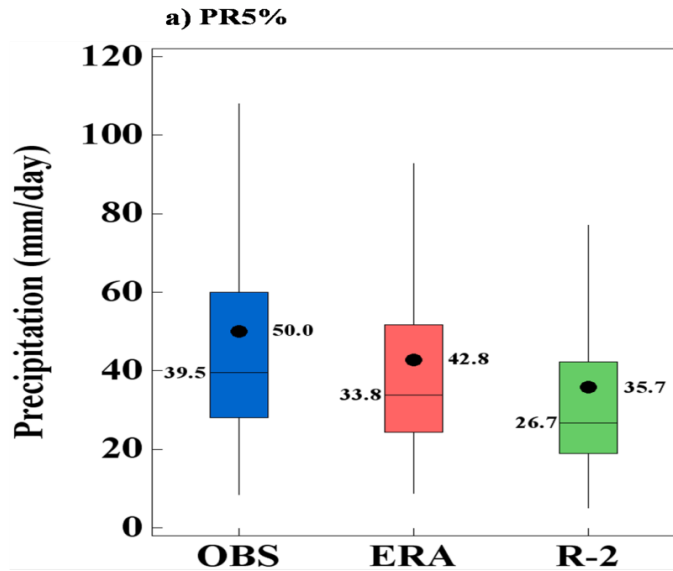
- WRF, SNURCM(MM5), RegCM4, RSM
- NCEP/DOE & ERA-interim Forcing
- 20 years (1989-2008)



Precipitation



Extreme Climate Indices (RegCM4 : Korea sub-region)



NARCCAP: HadRM3P T Biases



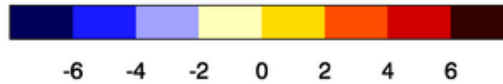
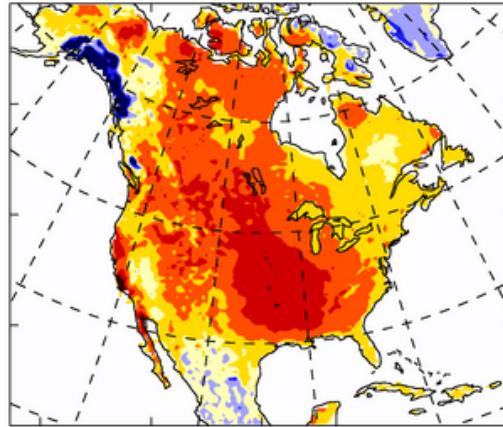
HadRM3P JJA/
DJF temperature
biases:

Top – ERA-In
boundaries

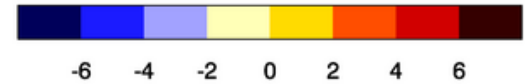
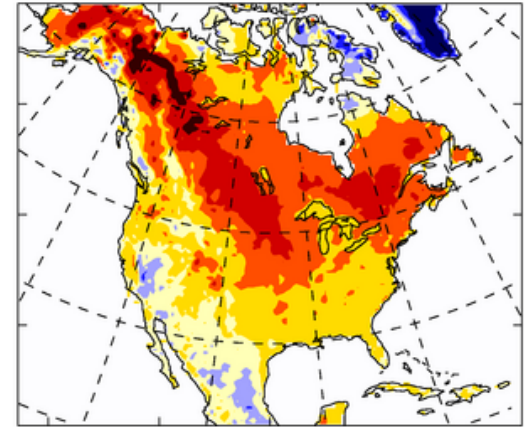
Bottom – NCEP
boundaries

Bias pattern similar
– magnitude much
greater with NCEP
boundaries

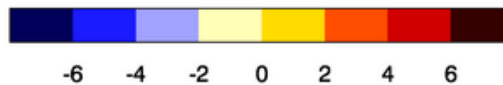
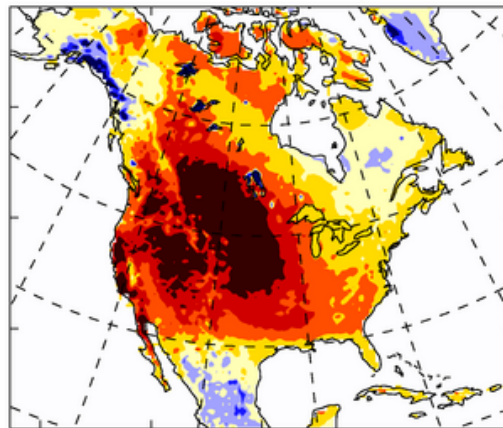
The multi-annual seasonal mean of 1.5m Temperature for the period 1990-1998 : model-observations
JJA HadRM3P-M2 Era Interim driven



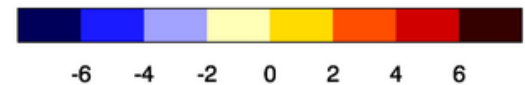
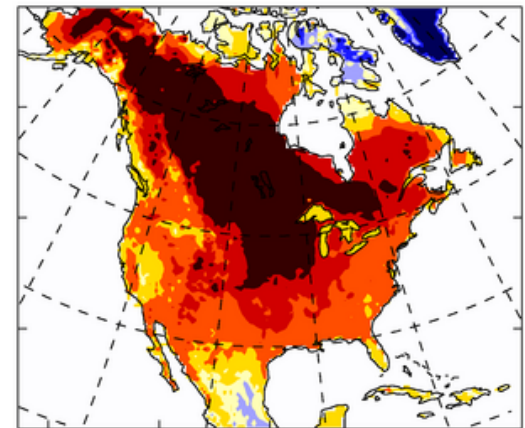
The multi-annual seasonal mean of 1.5m Temperature for the period 1990-1998 : model-observations
DJF HadRM3P-M2 Era interim driven



JJA HadRM3P-M1 ncep driven



DJF HadRM3P-M1 ncep driven

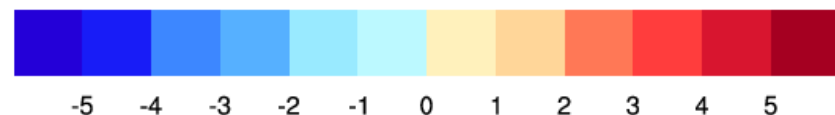
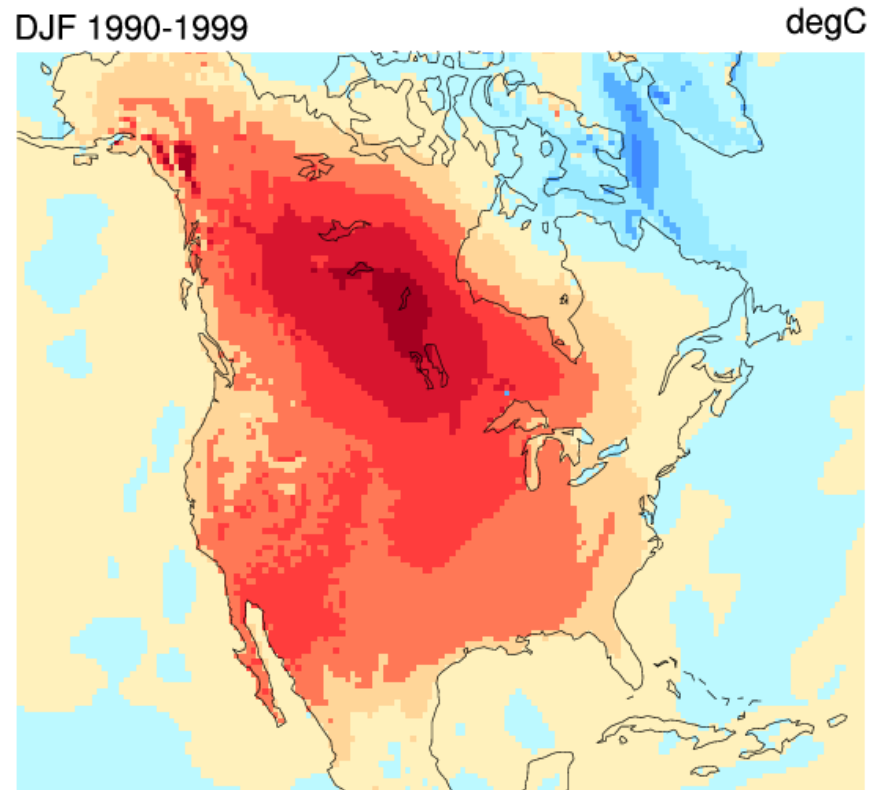


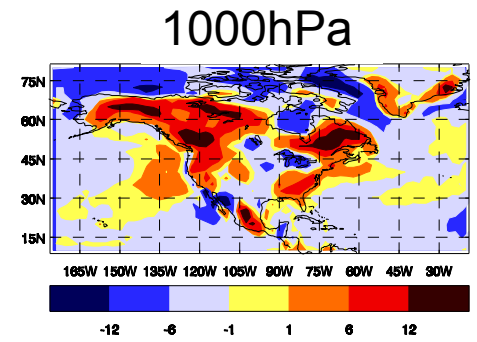
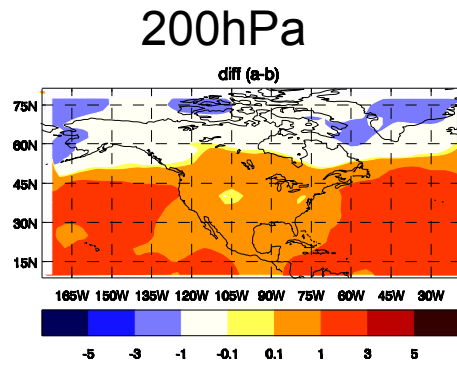
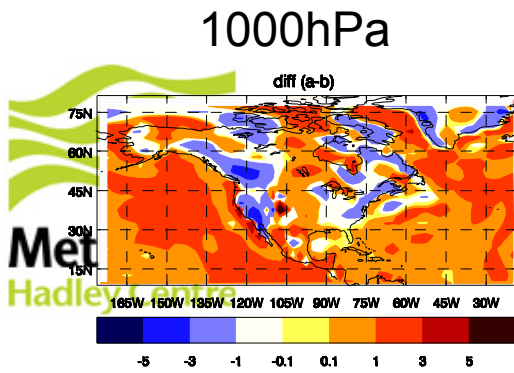
Difference in DJF surface temperature simulation when using NCEP rather than ERA-Interim boundaries

HadRM3P DJF temperature bias reduced by >5K when using ERA-Interim boundary conditions

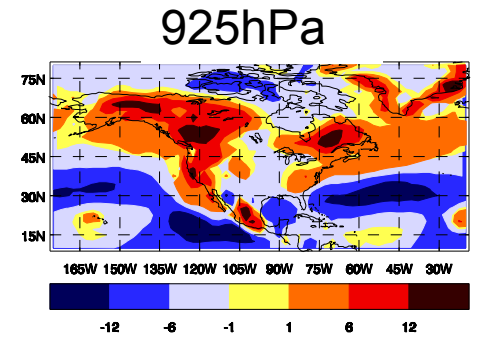
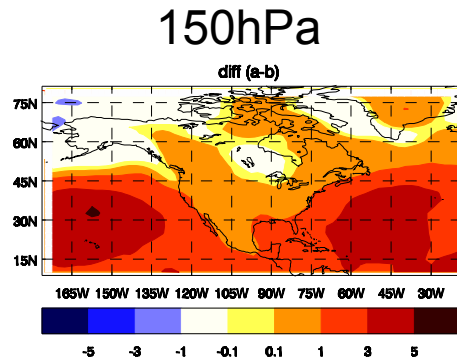
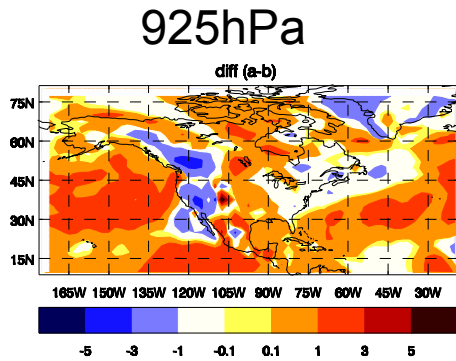
Similar signal seen in summer

HRM3 NCEP-ERA

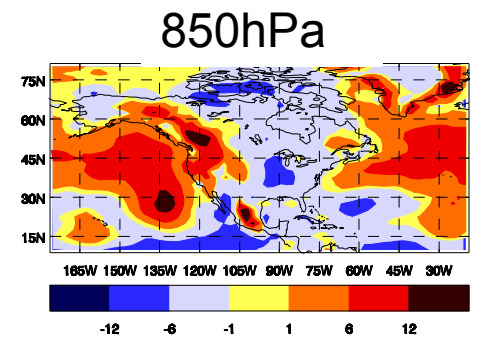
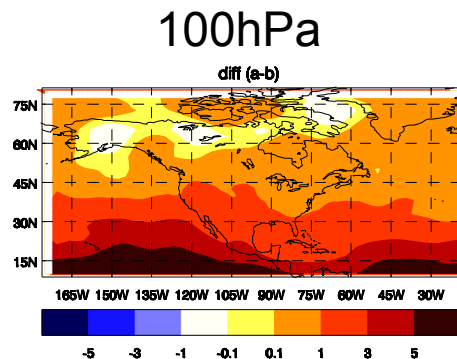
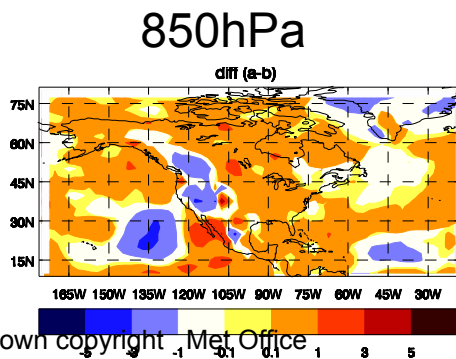




NCEP-ERA-Interim temperature (left and middle) and Rh



NCEP warmer and moister on western inflow boundary



Some reasons for differences in the downscaled reanalyses

- At western boundary, NCEP reanalysis both warmer and moister in the lower troposphere and in the upper troposphere/stratosphere
 - increases temperature and downward longwave radiation (from higher atmospheric humidities and, in winter, by increased cloud cover)
- In winter, warmer temperatures lead to
 - reduced snow cover and enhanced solar radiation absorption at the surface
- In spring/summer lead to
 - drier soils, lower evaporative cooling and lower cloud amounts enhancing surface solar radiation

Summary

- Reanalyses are central part of evaluation of dynamic downscaling:
 - provide “real world” BCs
 - provide basis for model evaluation
- Using reanalyses allows extensive testing of models against observations
 - ensembles especially important for this
- Reanalyses are not identical esp. away from well-observed regions, which can impact downscaling.

EXTRA SLIDES

NARCCAP



MM5

Iowa State/
PNNL

RegCM3

UC Santa Cruz
ICTP

CRCM

Quebec,
Ouranos

HADRM3

Hadley Centre

RSM

Scipps

WRF

NCAR/
PNNL

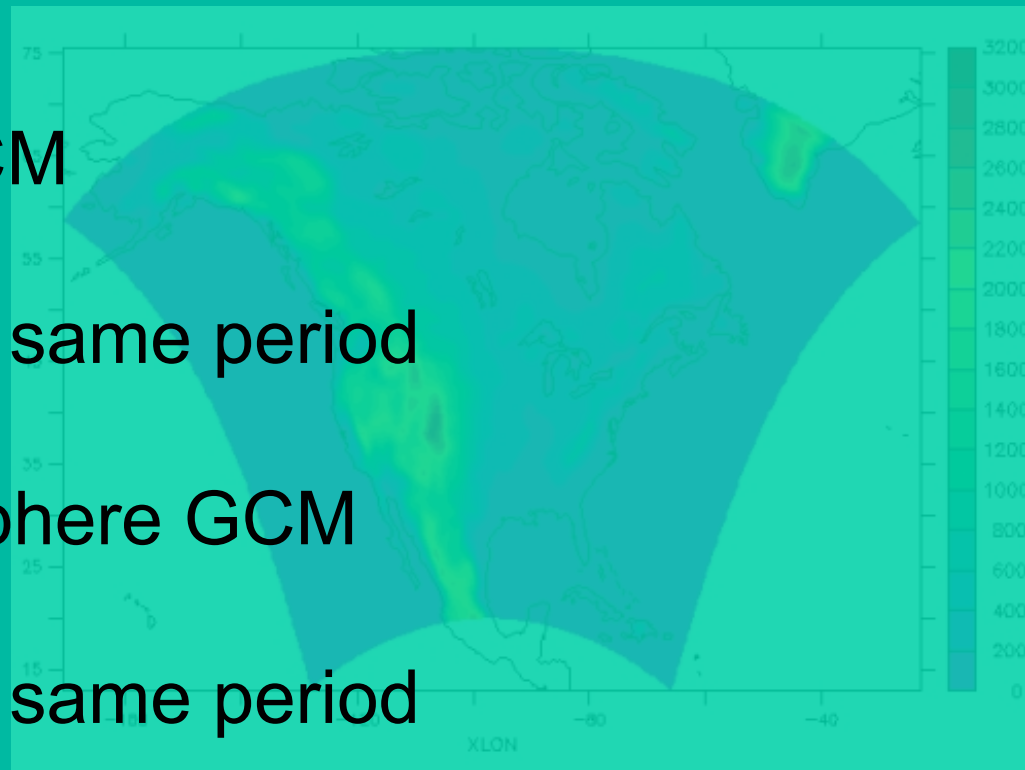
PLUS:
GFDL Atmosphere GCM

- 0.5° resolution
- specified SST/ice for same period

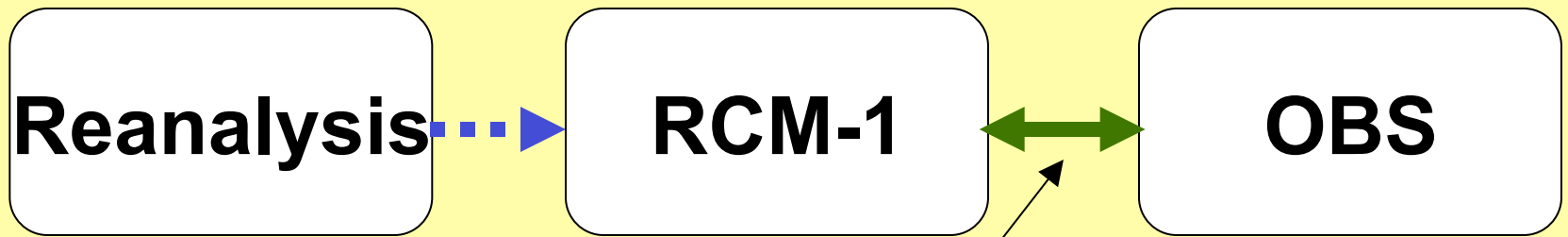
Analysis Period

NCAR CAM3 - Atmosphere GCM

- ~0.5° resolution
- specified SST/ice for same period

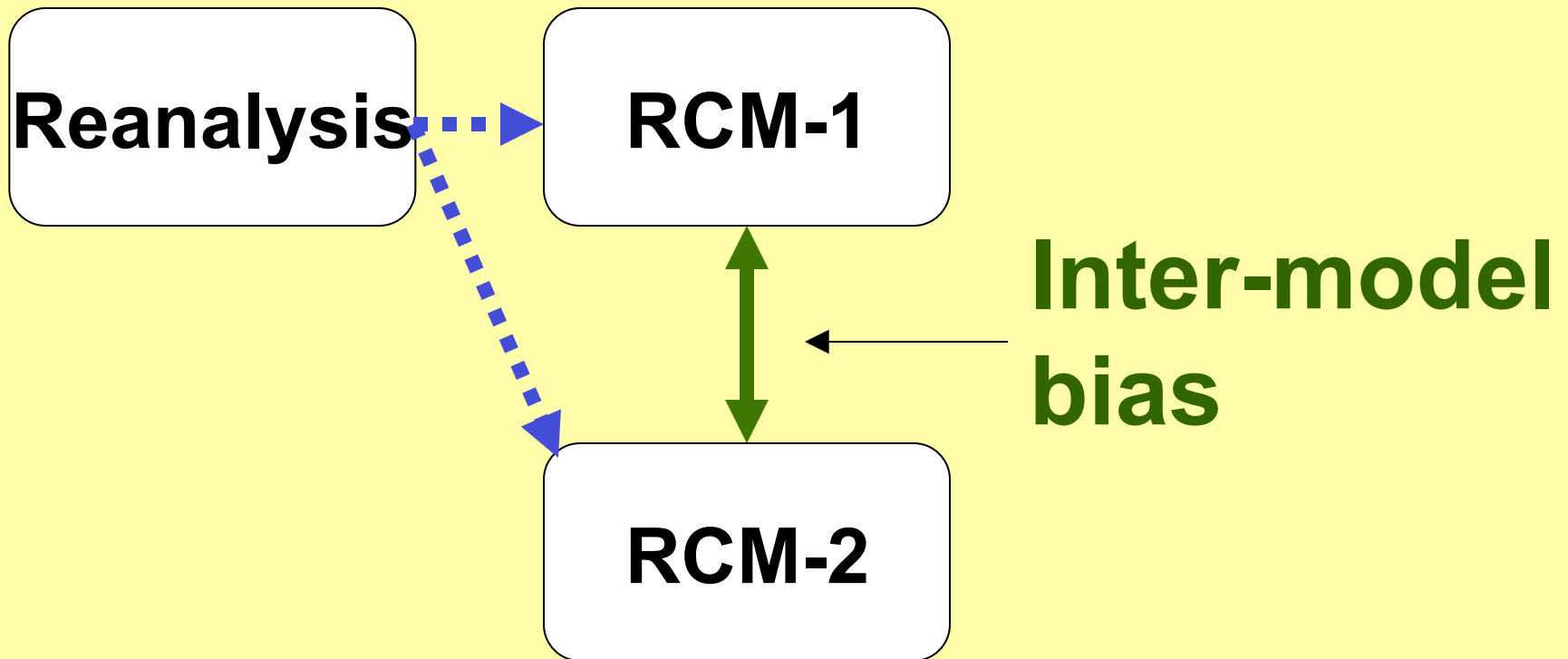


Definition of Biases

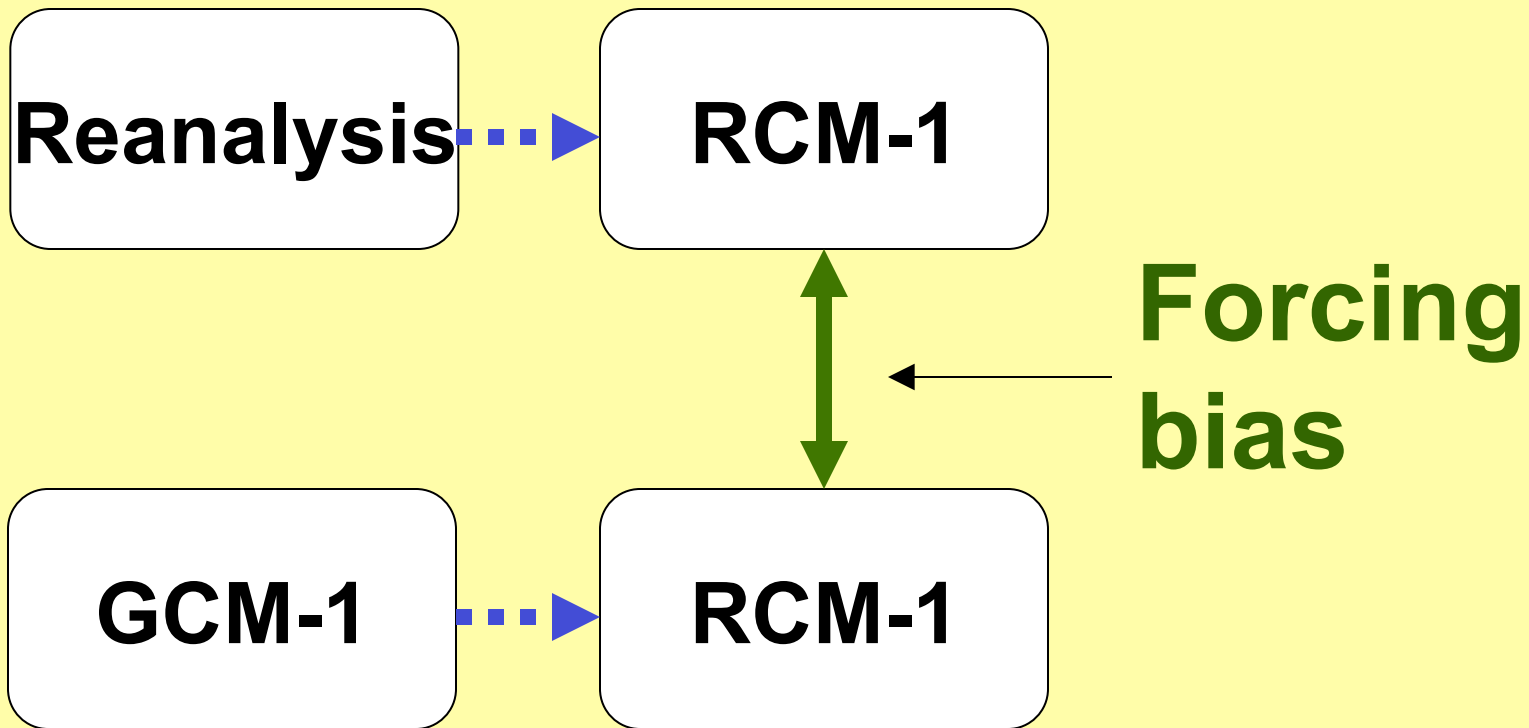


RCM (performance) bias

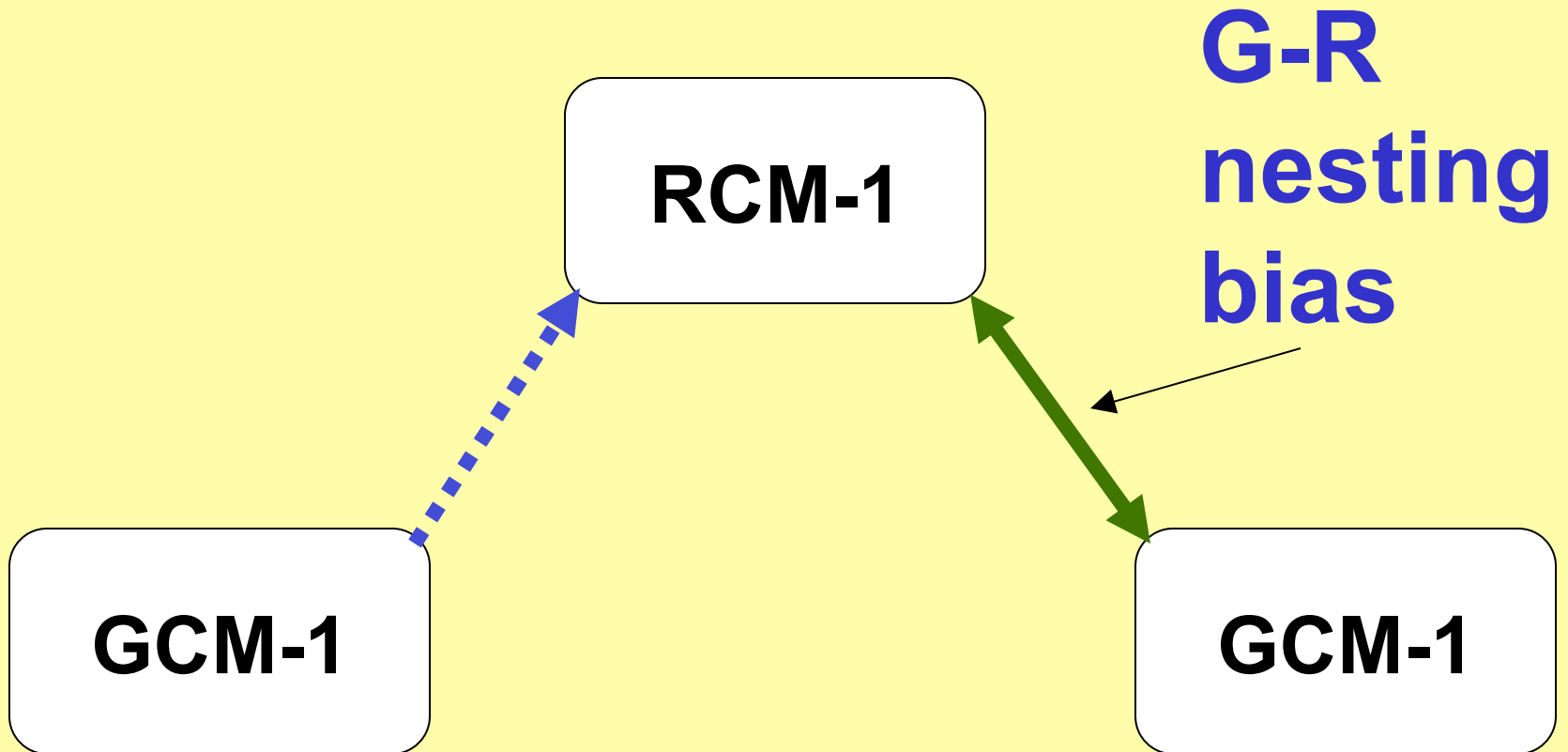
Definition of Biases



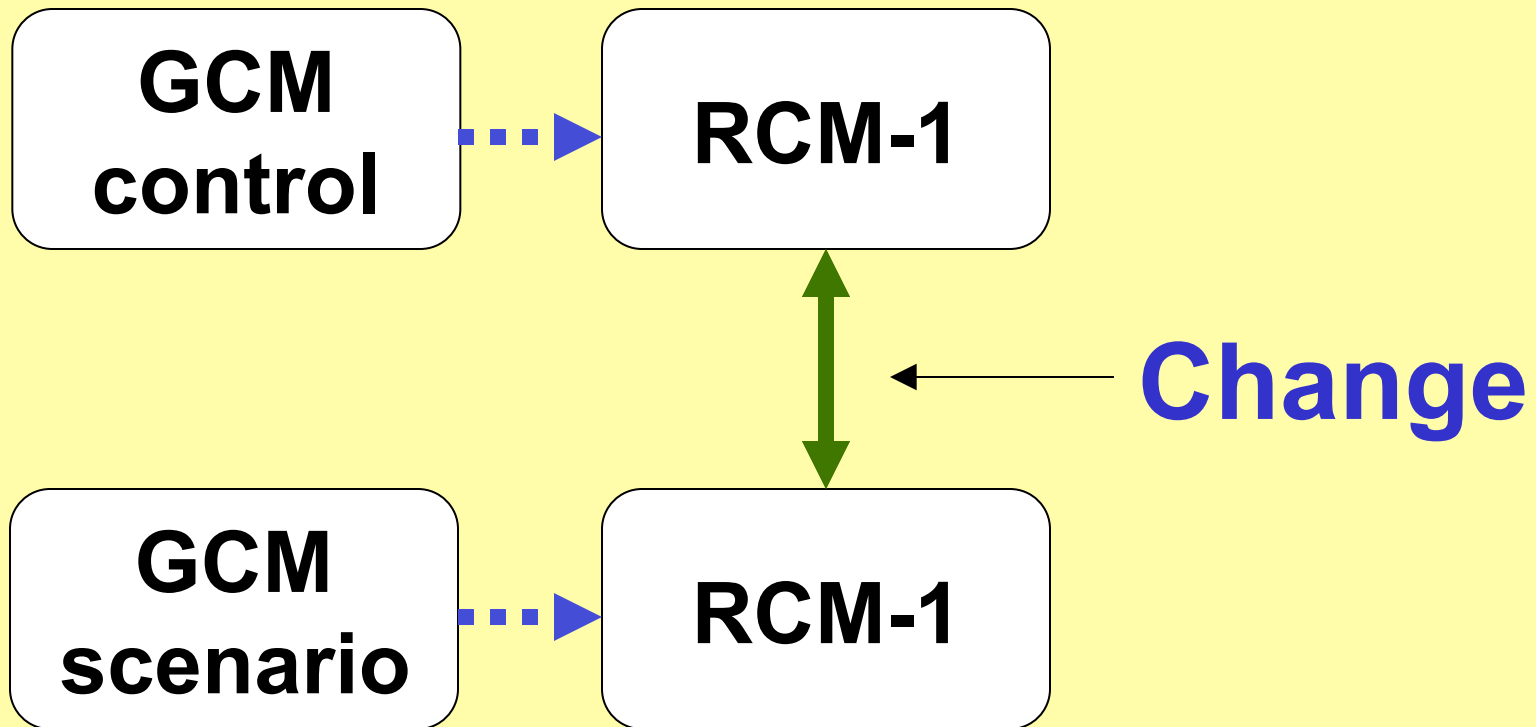
Definition of Biases

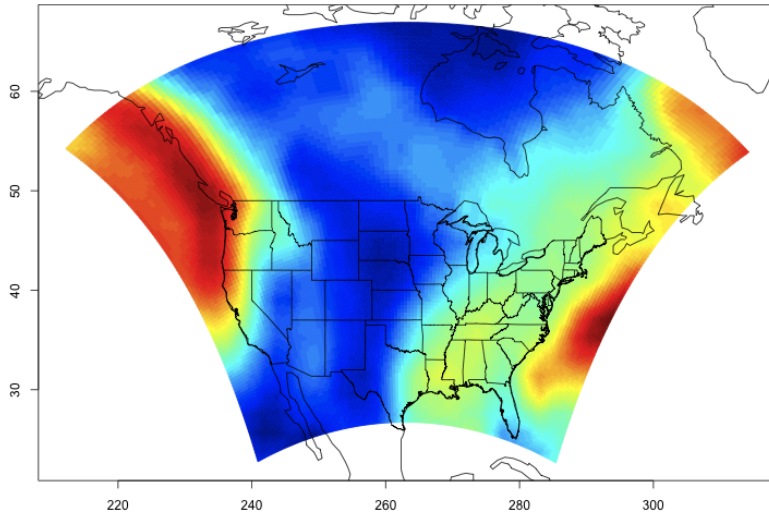


Definition of Biases

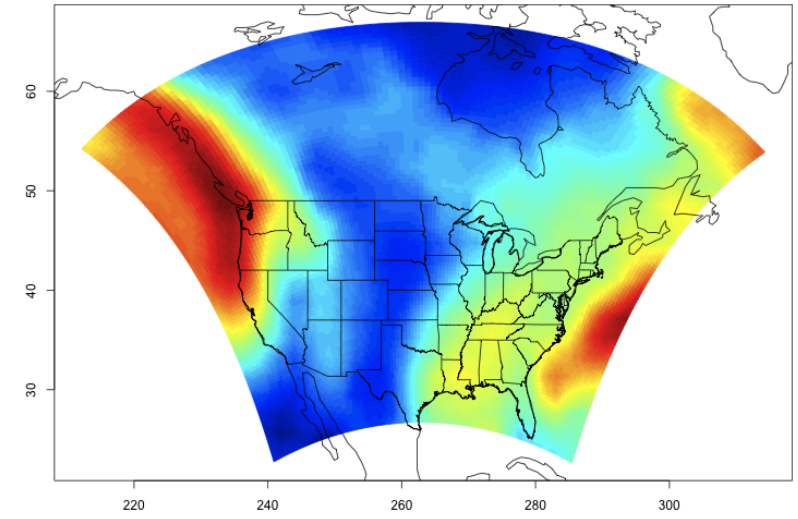


Climate Change

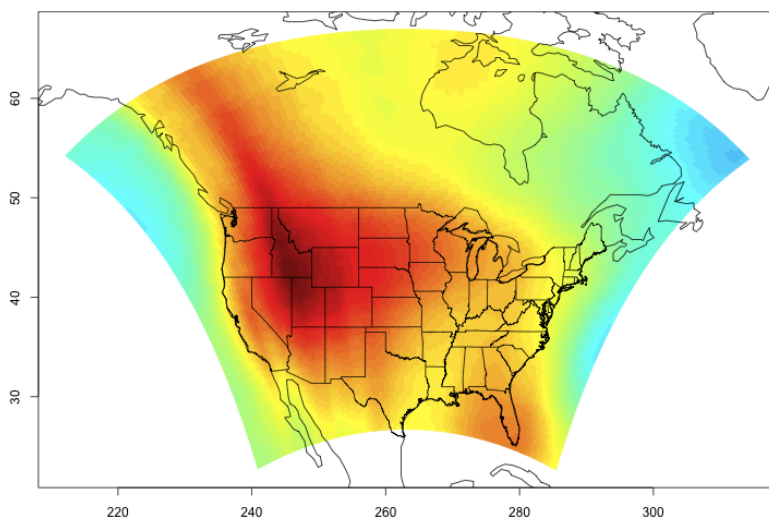




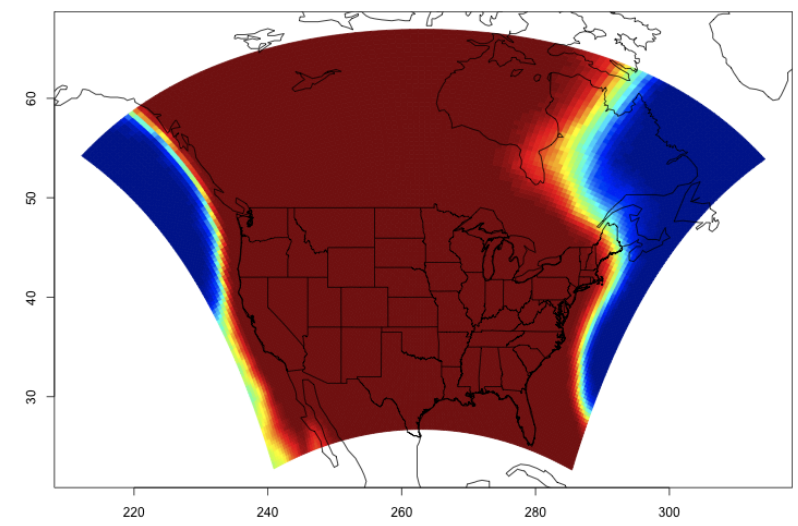
$\overline{\text{NCEP}}$



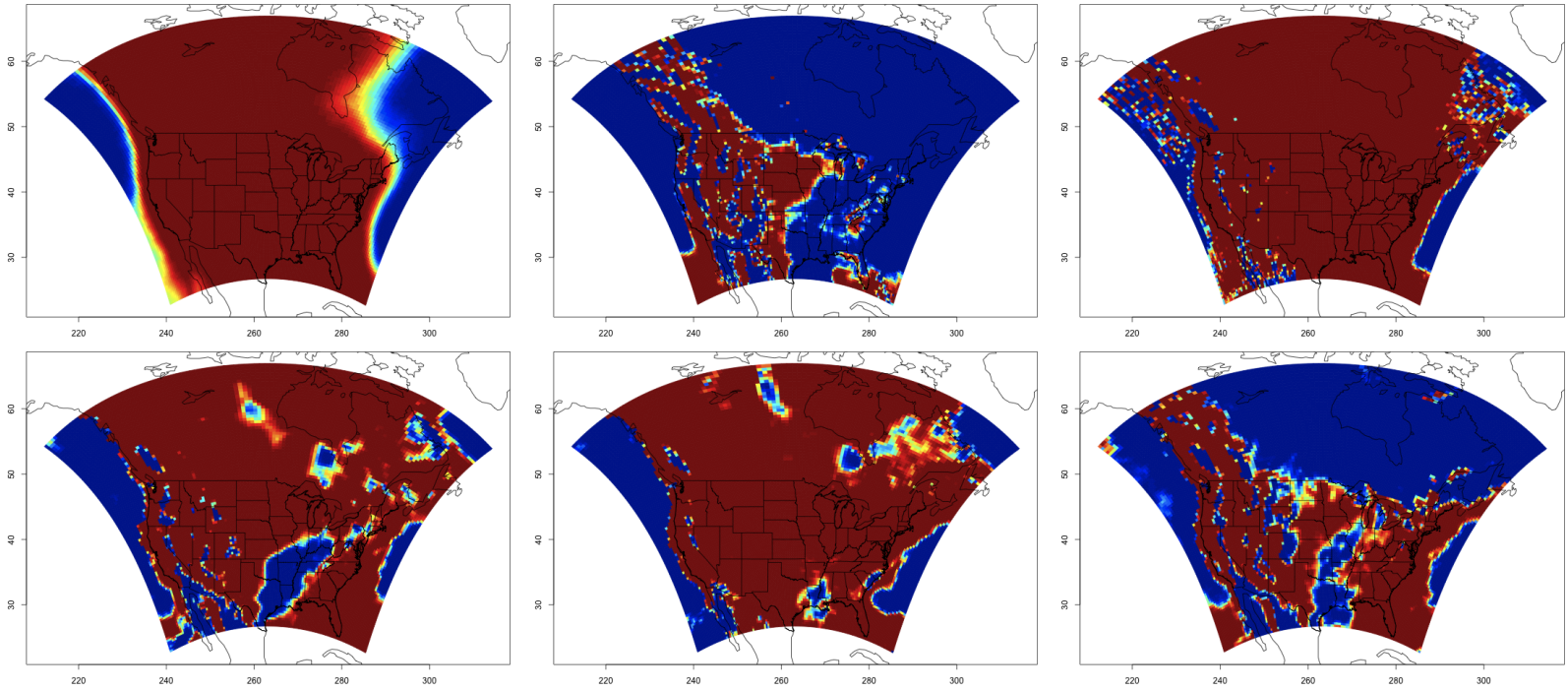
$\bar{\mu}$



$\bar{\mu} - \overline{\text{NCEP}}$



$P[\mu > \overline{\text{NCEP}}]$



ALL CRCM ECPC
 MM5 RCM3 WRFP

Performance of CORDEX RCMs in Simulating East African Rainfall

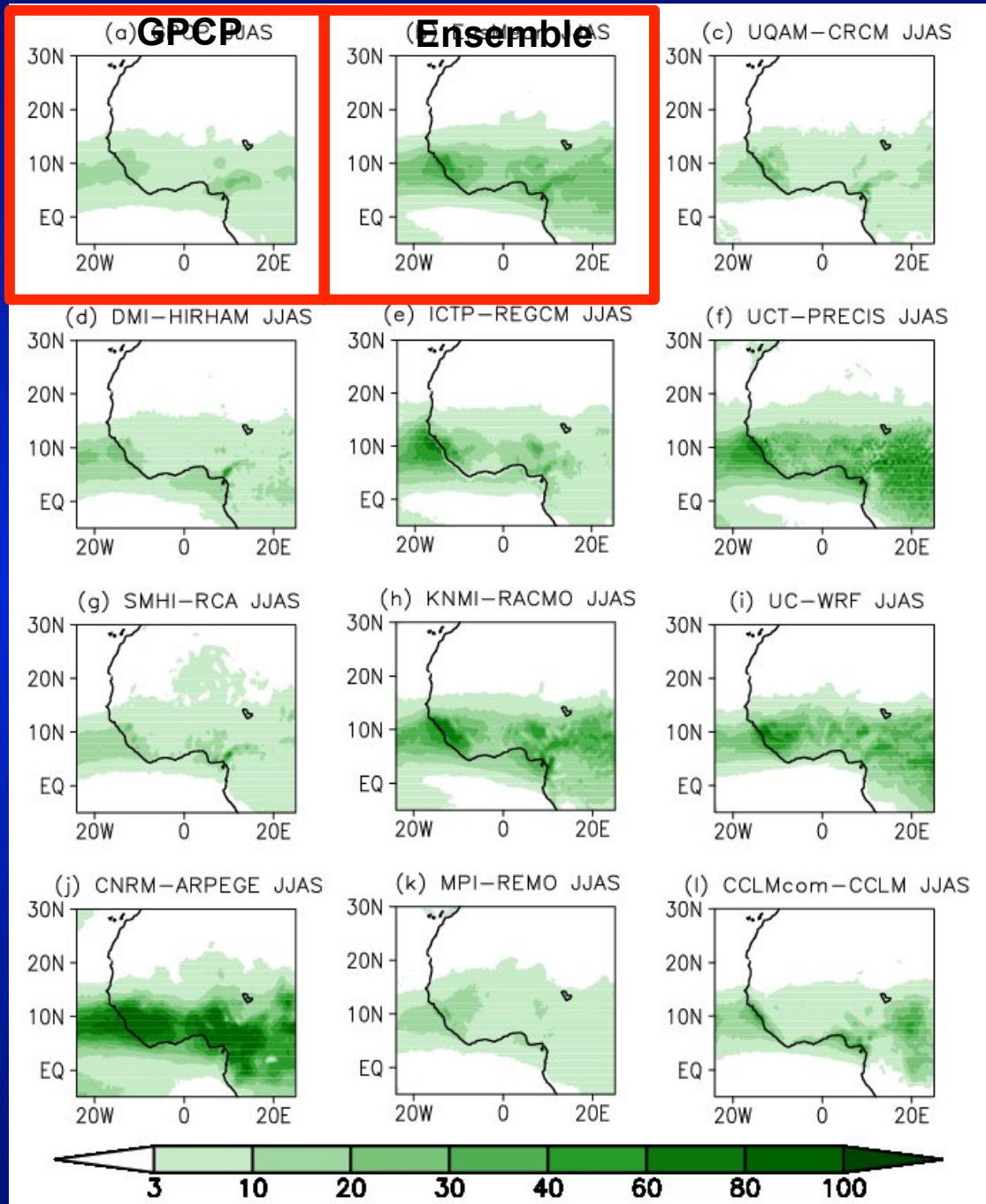
The first two figures below show the spatial distribution of rainfall for seasons JAS (fig1 a) and OND (fig1 b) averaged from 1990 to 2008 for the RCMs simulations and GPCC observation over the East Africa. The seasonal choice was based on the rainy season in the region. The maximum rainfall is located in north of equator during JAS(fig1a), and south of the equator during OND (fig1b) which is in association with the movement of ITCZ.

The results shows that most of the RCMs simulated the realistic rainfall belt well along the ITCZ except overestimation the magnitude. The ensemble is found to be more close to observation in representing spatial rainfall distribution over the region.

Fig2 shows the comparison between CORDEX-RCMs simulation and observed(GPCC) spatial averaged seasonal mean rainfall anomaly during OND in CORDEX Africa sub-region 5. Most of the RCMs realistically simulate the inter-annual rainfall variability in agreement with the observation, except overestimation in some of the RCMs. All the RCMs capture the 1997 high rainfall event which is associated with strong El Nino occurrence.

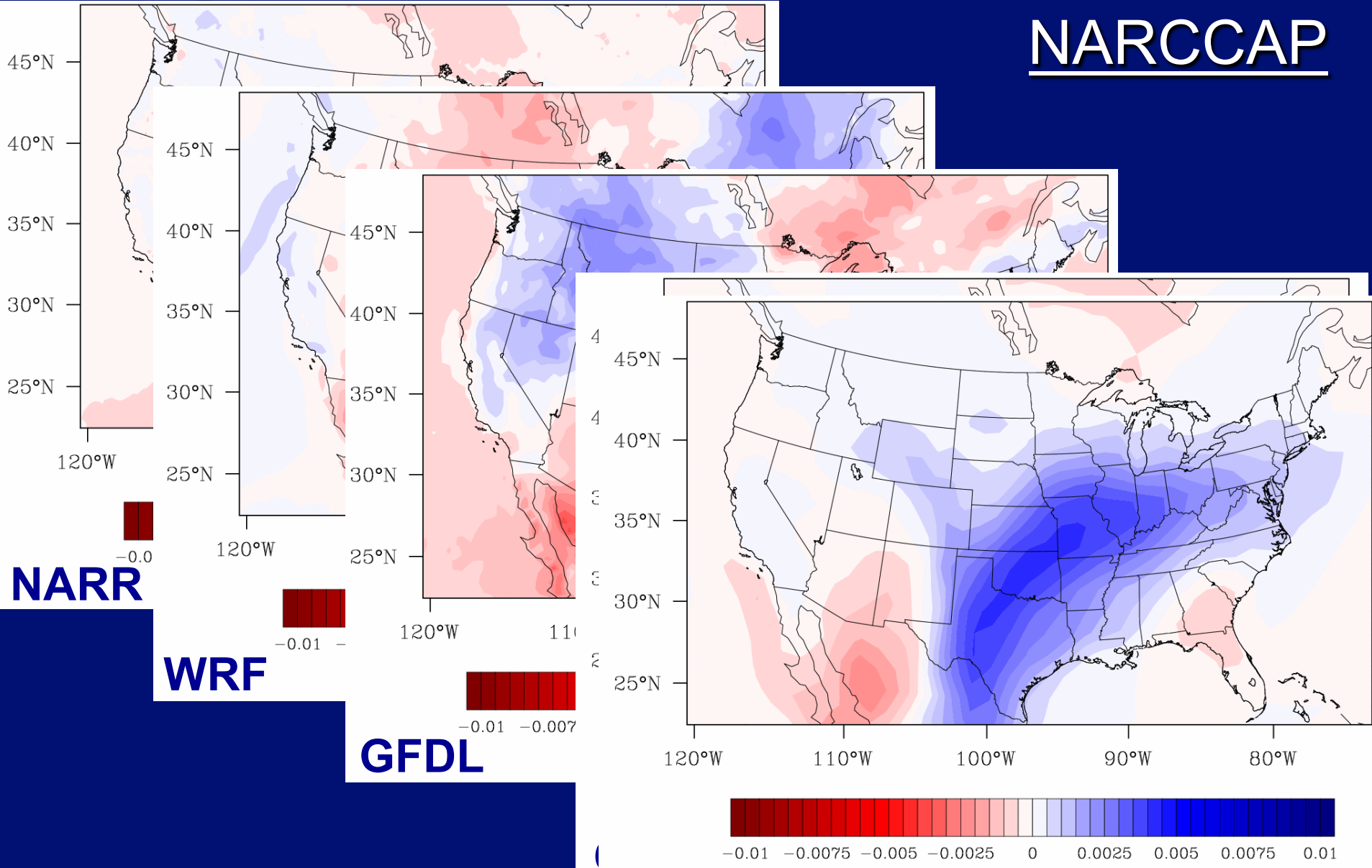
In general, the analysis indicating that the CORDEX RCMS are useful in providing many regional details that GCMs fail to capture due to the limited resolution. The first notable problem in most of the CORDEX RCMs is the overestimation of rainfall.

CORDEX Africa: Frequency of West Africa Wet Spells (1989-2008 ; JJAS)

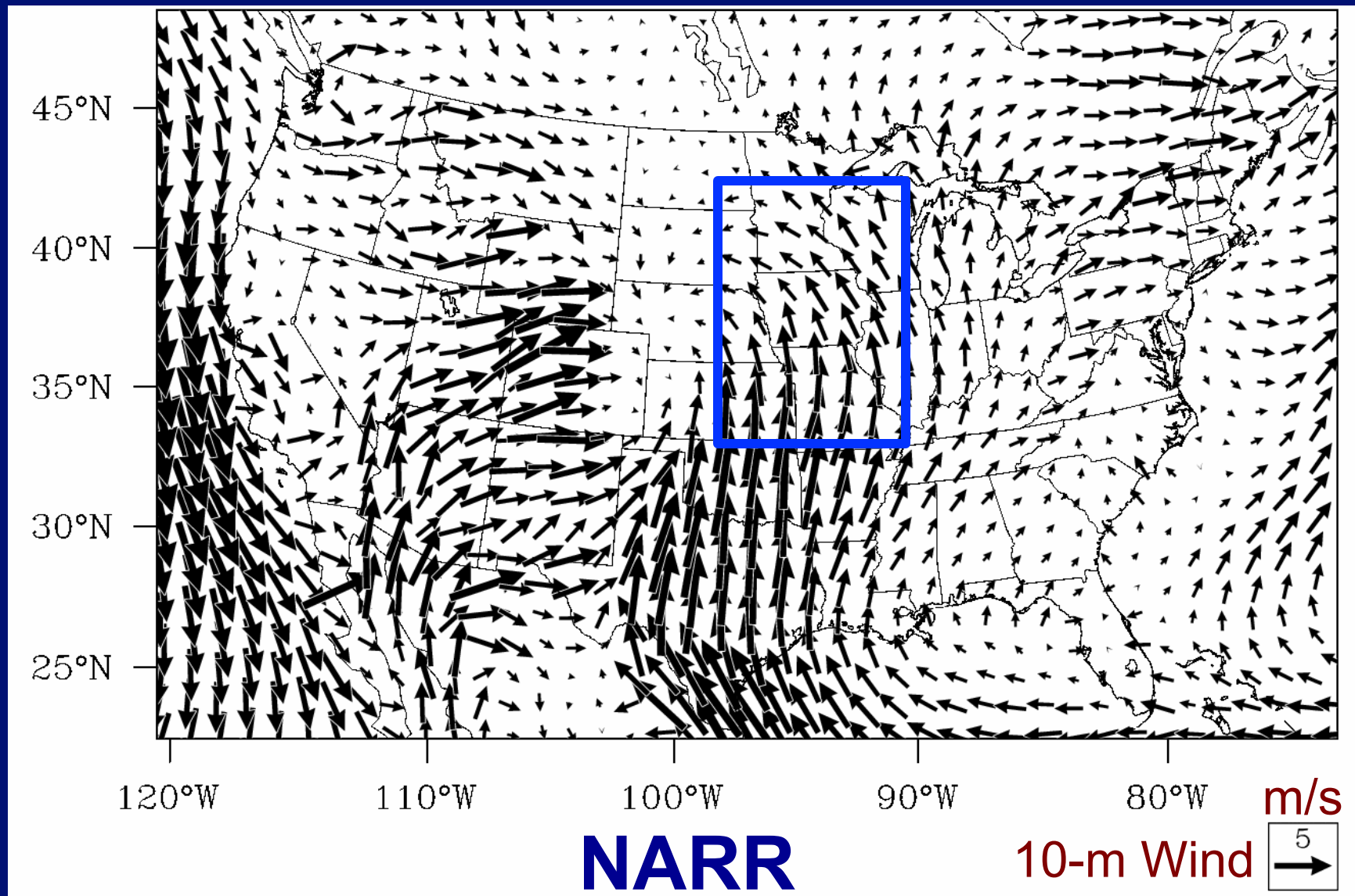


Composite Structure of Extreme Events - JJA

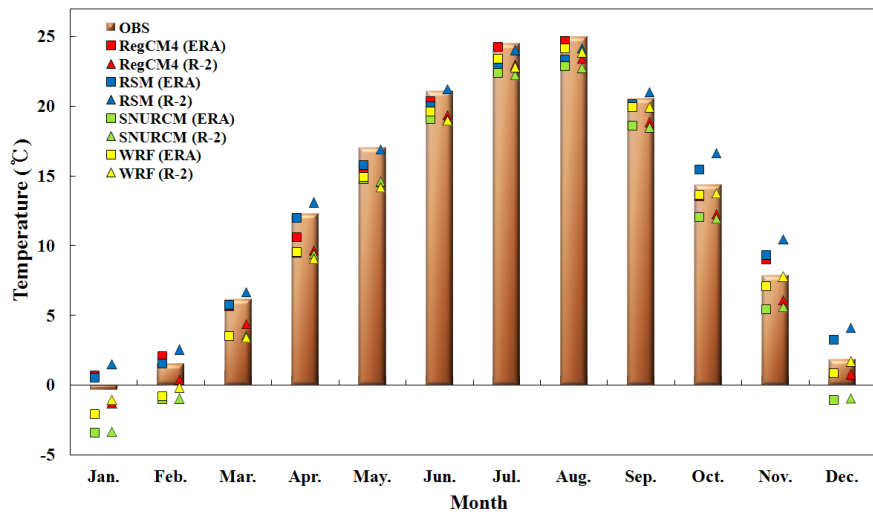
NARCCAP



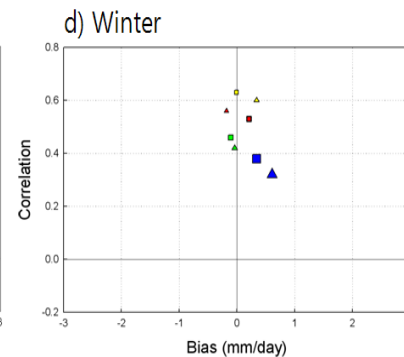
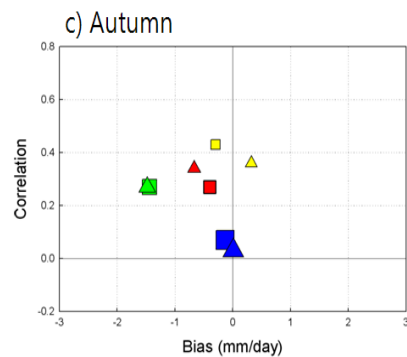
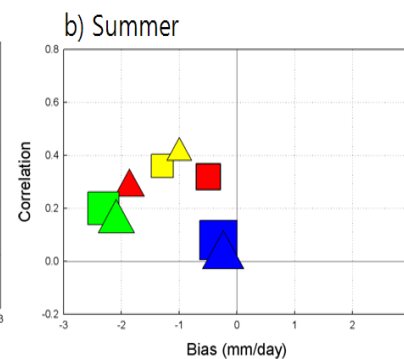
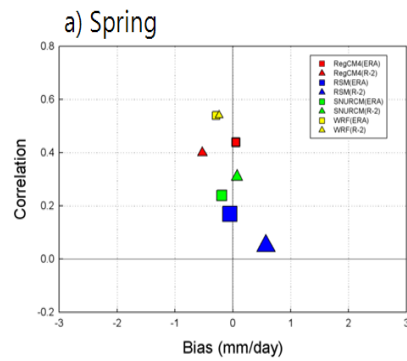
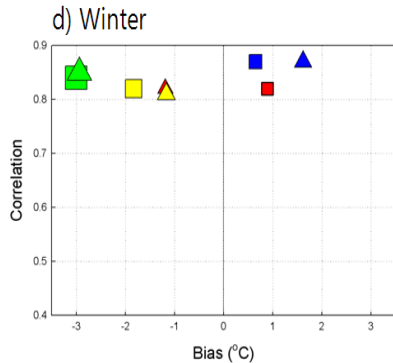
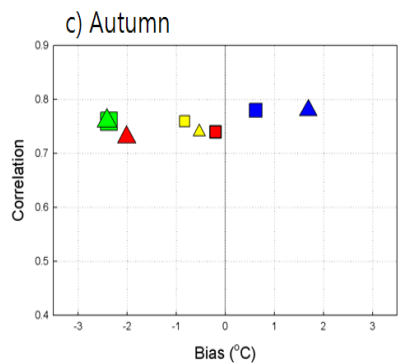
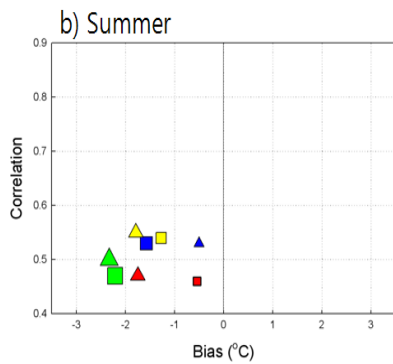
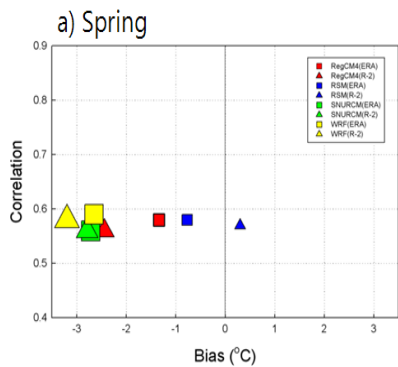
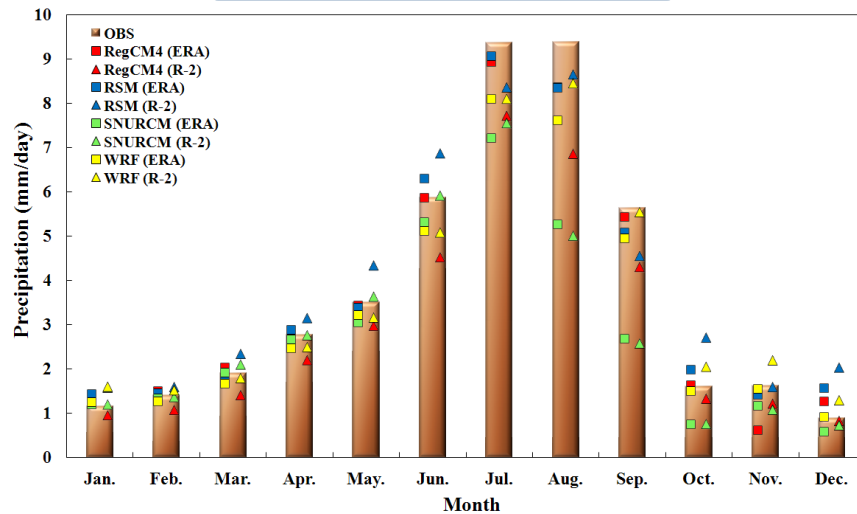
Composite Structure of Extreme Events - JJA



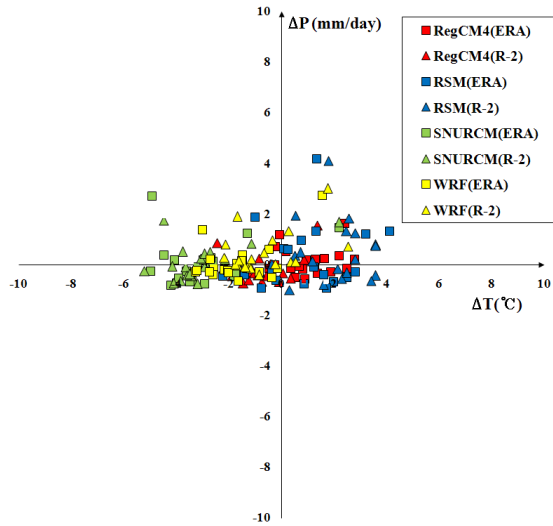
Temperature



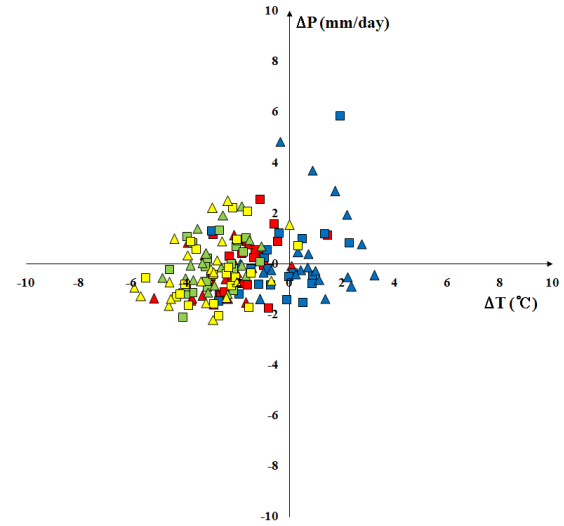
Precipitation



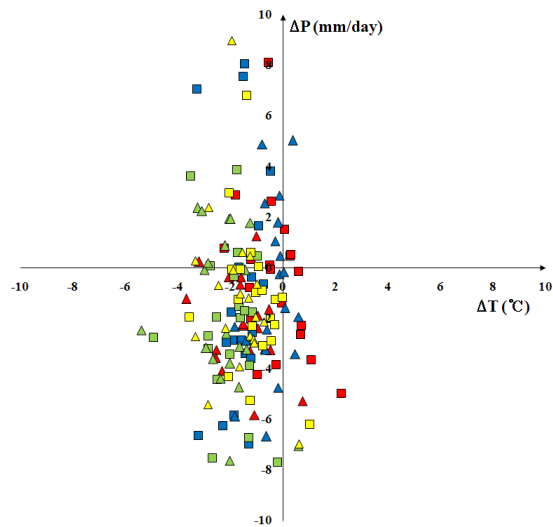
a) Jan.



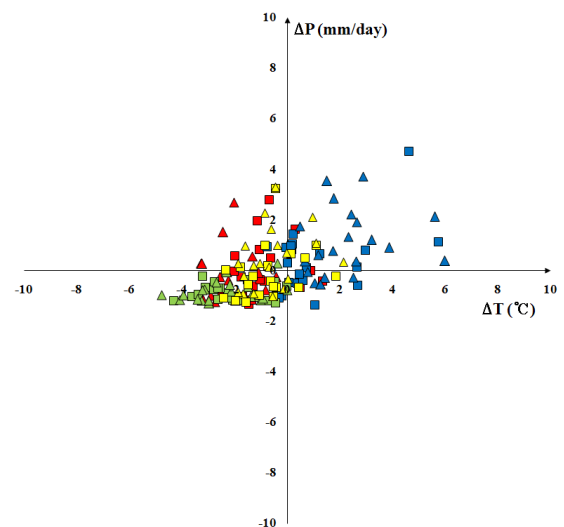
b) April

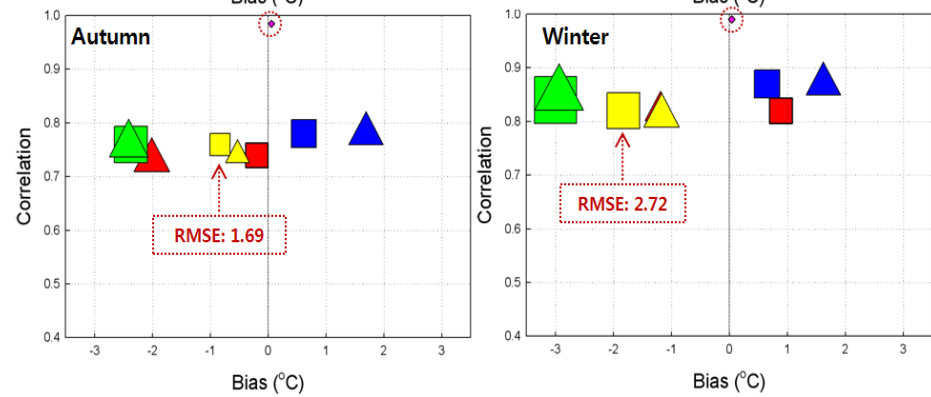
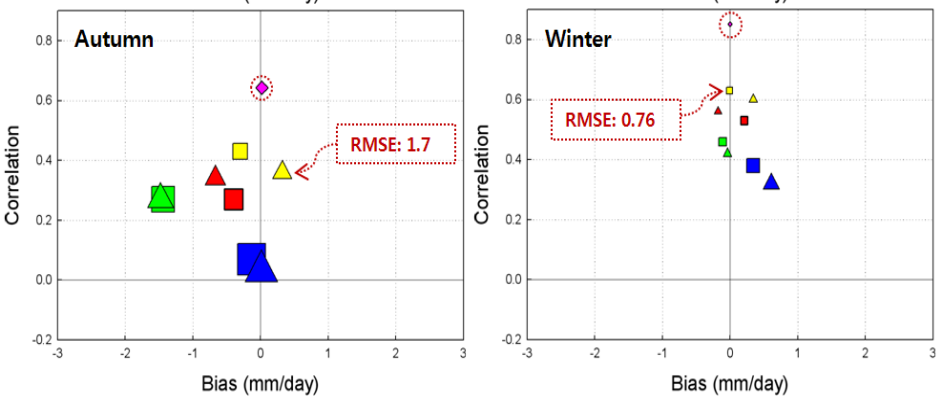
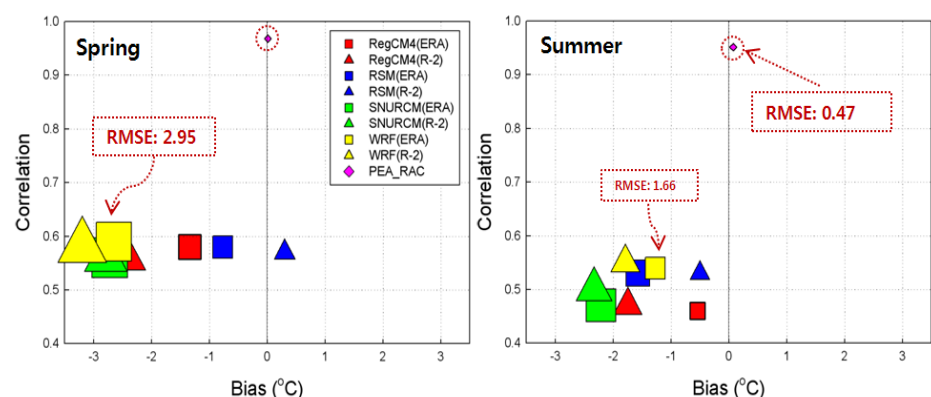
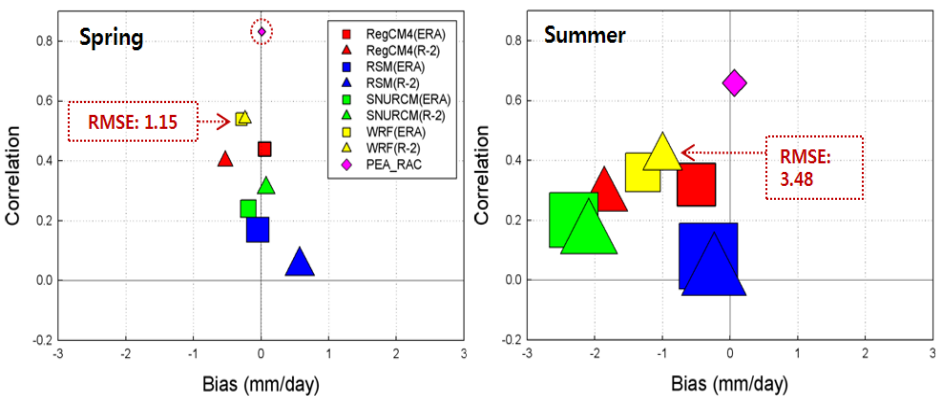


c) July



d) Oct.







Summary of reanalysis downscaling analysis

- HadRM3P-NCEP has by far the largest seasonal temperature biases (see BAMS article)
- HadRM3P has no systematic tendency for significant positive temperatures biases (Xu et al., 2006 over China, Marengo et al., 2009 over South America, Kamga et Buscalet, 2006, over Africa)
- Previous studies have used ECMWF boundary conditions, ERA-15, ERA-40 or ERA-Interim reanalyses
- HadRM3P-ERA-Interim data produces significantly lower (5K) temperature biases - and similar in magnitude to the other RCMs.

Some reasons for differences in the downscaled reanalyses

- At the boundary of the NARCCAP domain the NCEP data are both warmer and moister in the lower troposphere and in the upper troposphere/stratosphere – directly increasing temperature and increasing downward longwave radiation (from higher atmospheric humidities and, in winter, by increased cloud cover)
- In winter the warmer temperatures lead to reduced snow cover thus enhanced solar radiation absorption at the surface and in spring/summer lead to drier soils, lower evaporative cooling and lower cloud amounts enhancing surface solar radiation