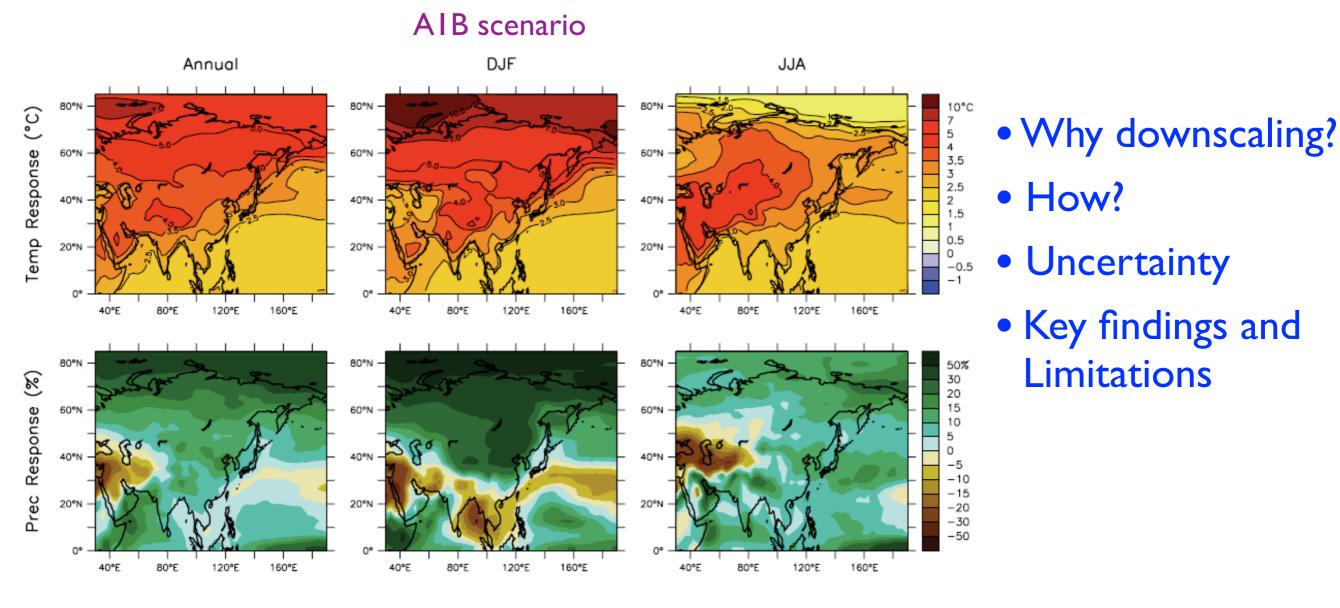
#### Regionalization of the IPCC AR4 Climate Projections over East Asia and Taiwan with Statistical Downscaling

Cheng-Ta Chen and Ya-Ru Chang, National Taiwan Normal University, Department of Earth Sciences NCDR Taiwan Climate Change Projection and Information Platform Project Team



# Why downscaling?

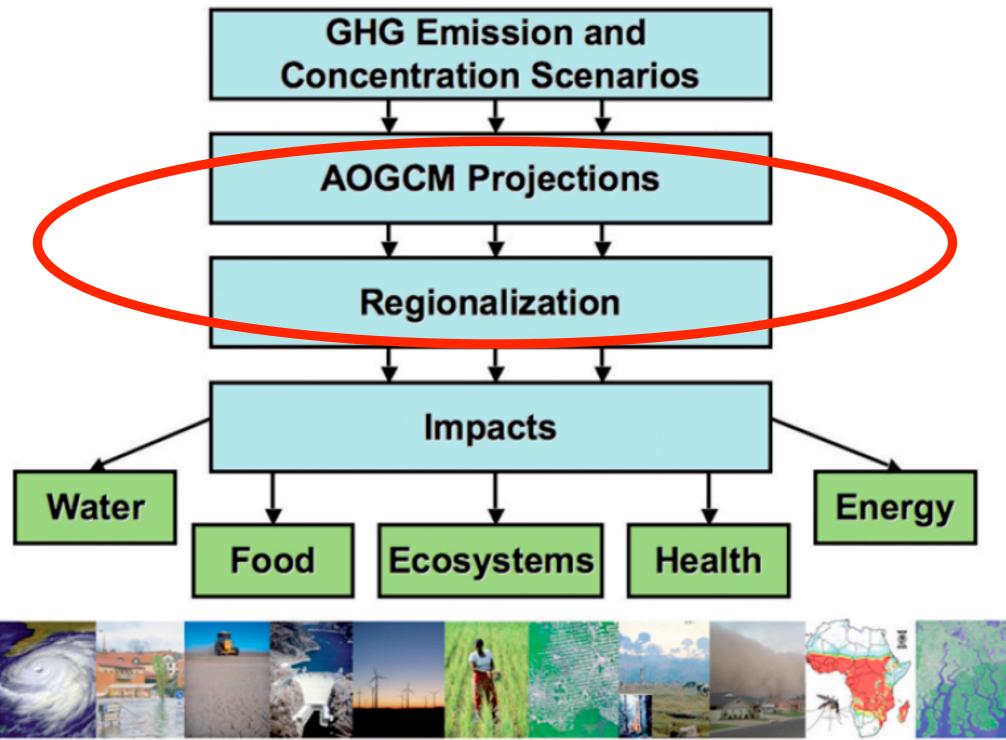
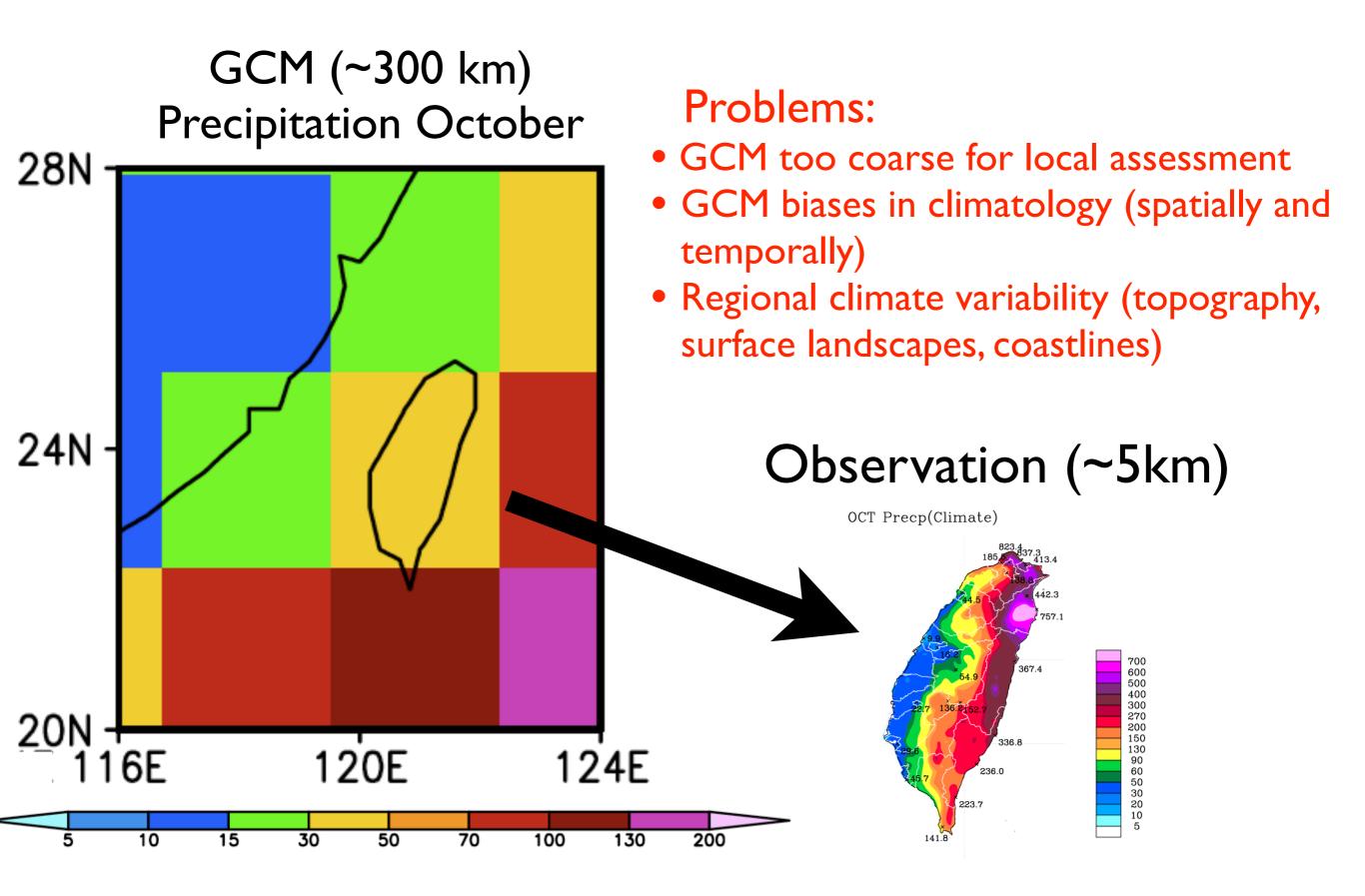


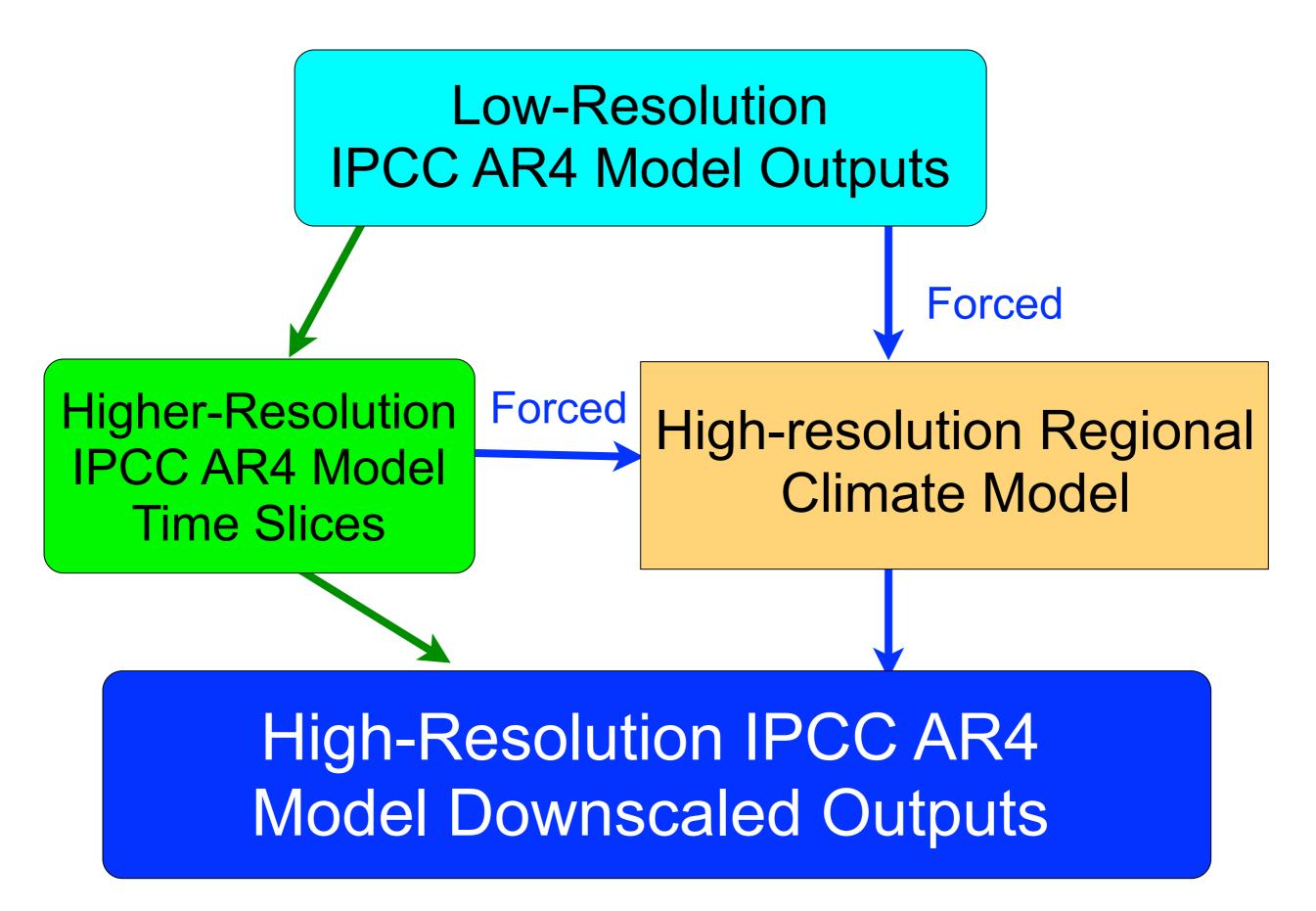
Figure 3 — Schematic depiction of the steps involved in the production of climate change information usable for impact assessment work via regionalization methods

Source: Giorgi (2008)

# Why downscaling?



## **Dynamical Downscaling**



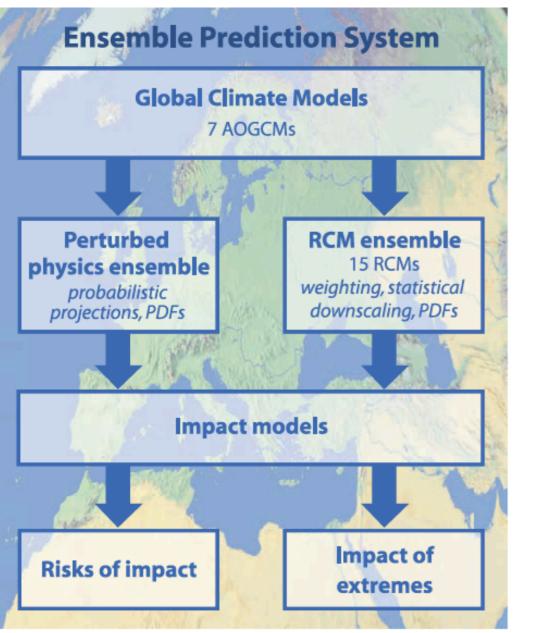
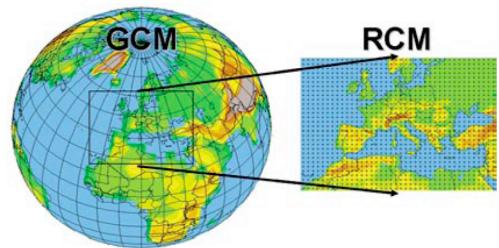
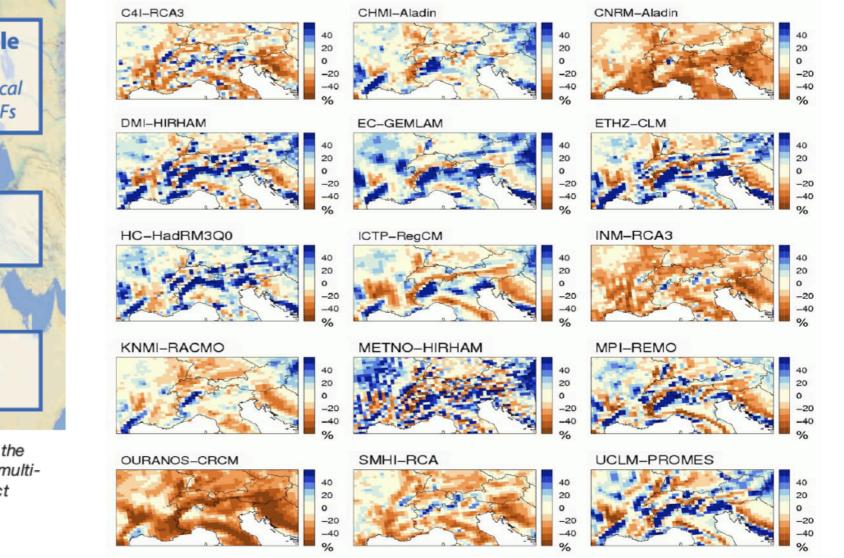


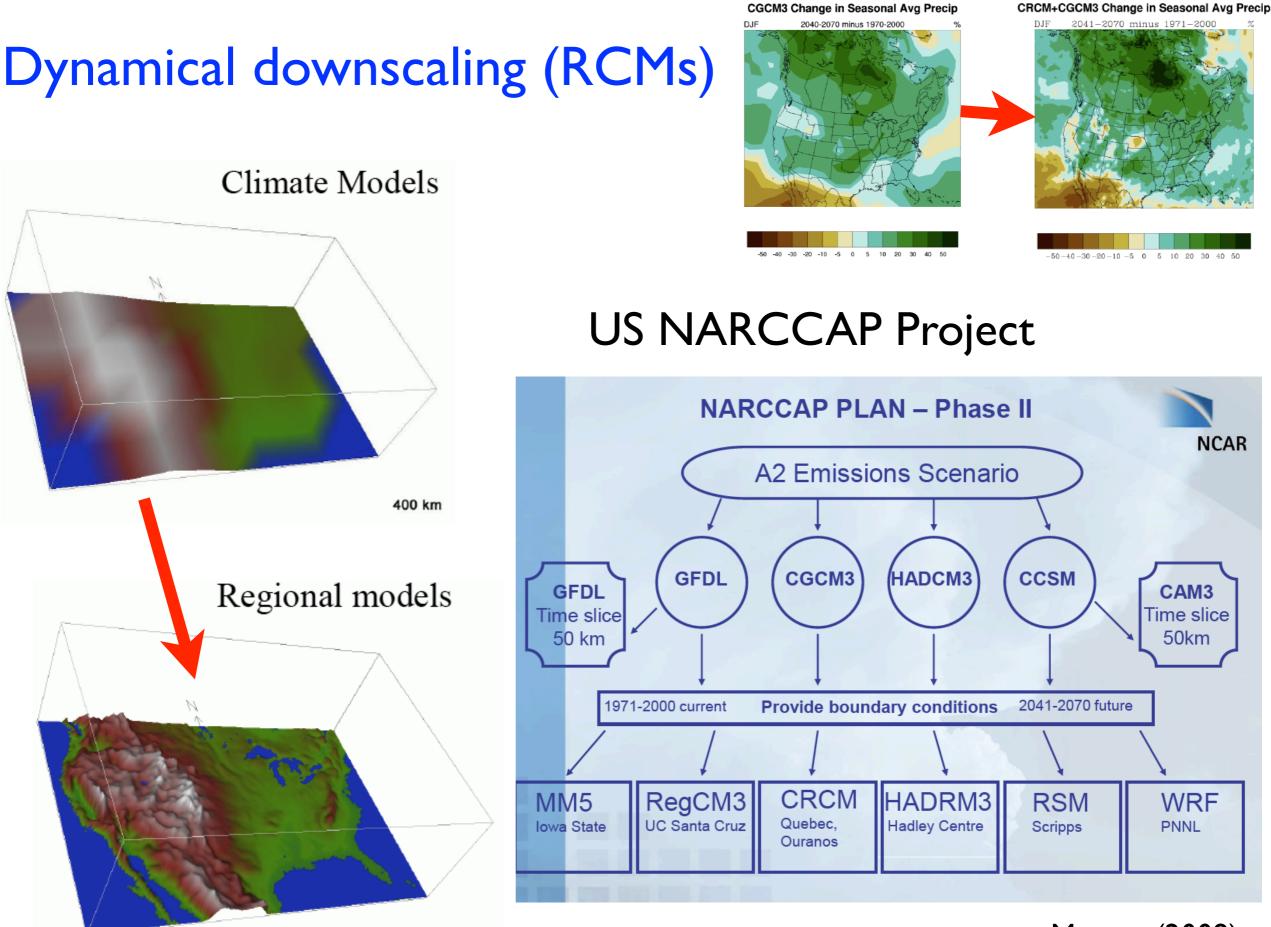
Figure 2.1: Linkages between the modelling components of the ensemble prediction system (EPS), as developed for use at multidecadal to centennial timescales, and the methods of impact assessment using outputs from the system.

# Dynamical downscaling (RCMs)



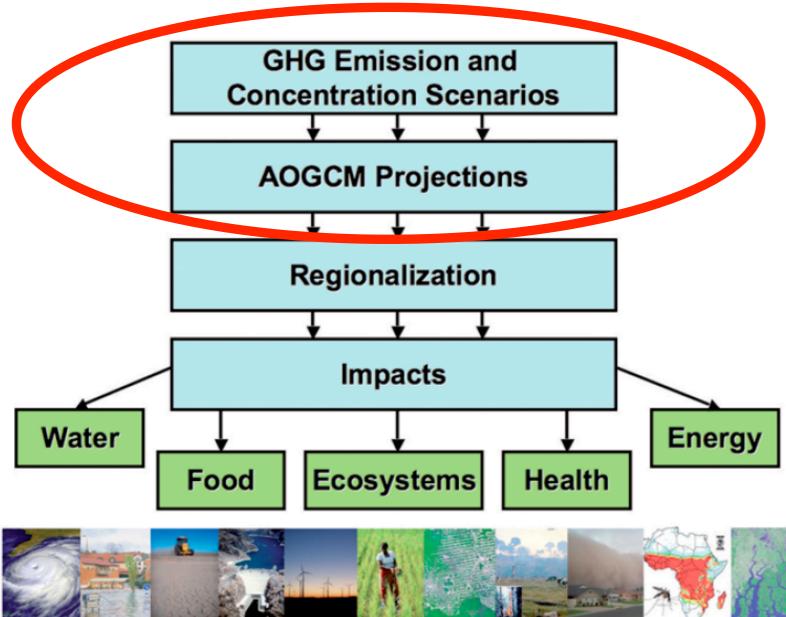
#### EU ENSEMBLE Project RT3





25 km

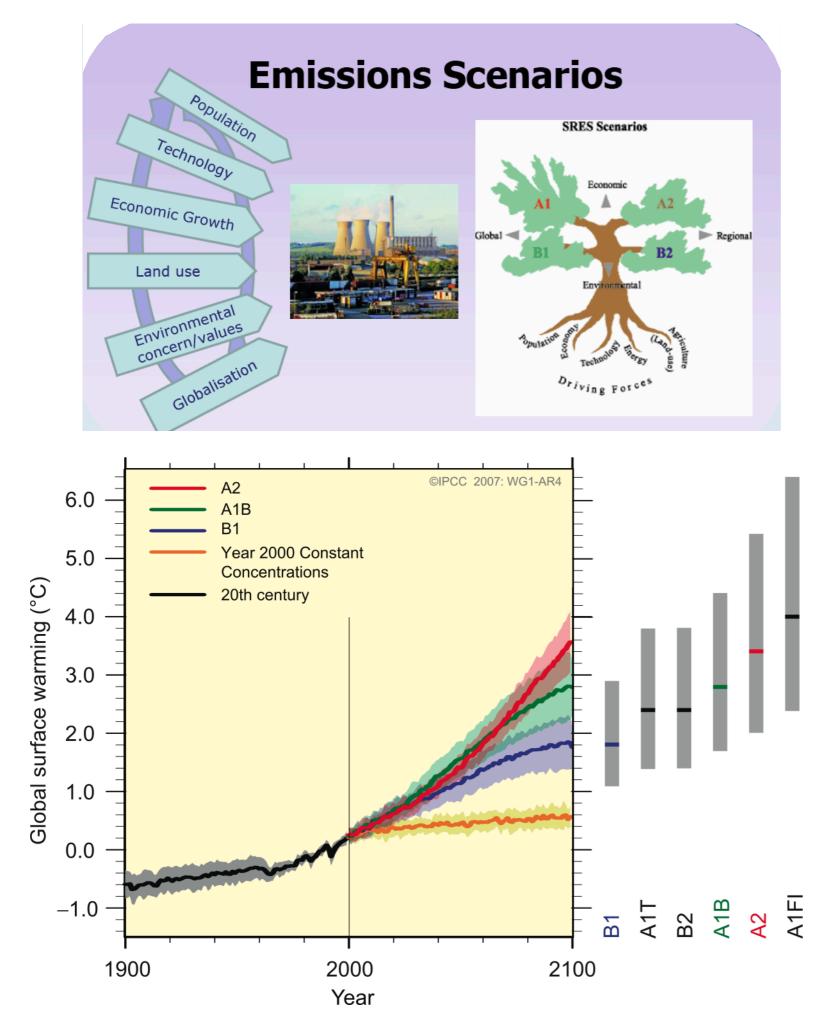
Mearns (2009)



- Uncertainties in future greenhouse gas and aerosol emissions
- Uncertainties in global and regional climate sensitivity, due to differences in the way physical processes and feedbacks are simulated in different models

Figure 3 — Schematic depiction of the steps involved in the production of climate change information usable for impact assessment work via regionalization methods

Source: Giorgi (2008)



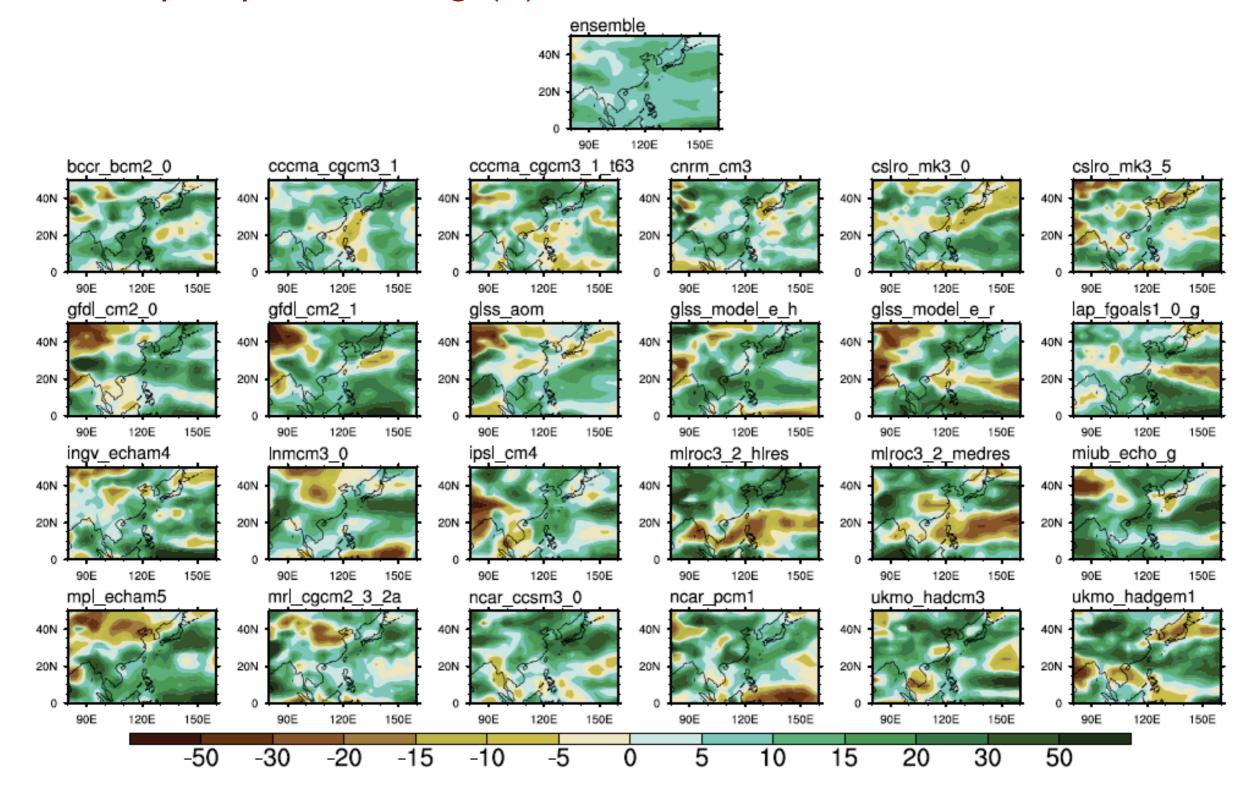
## Keep the uncertainty

- Uncertainties in future greenhouse gas and aerosol emissions
- Uncertainties in global and regional climate sensitivity, due to differences in the way physical processes and feedbacks are simulated in different models

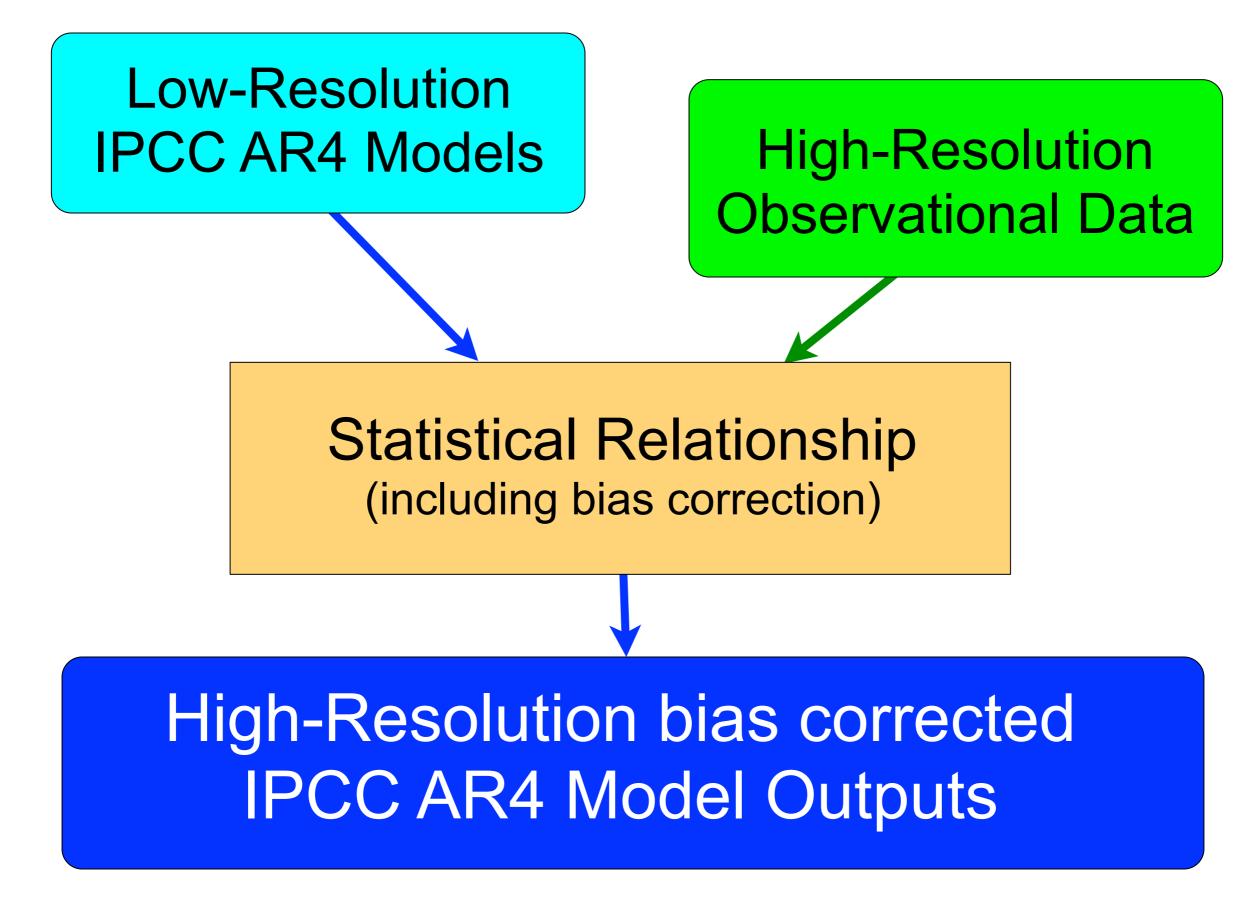
Including all scenarios and all models for future climate projection

#### Uncertainty from Global Climate Models

Summer precipitation change(%) with all IPCC AR4 models under AIB scenario



## (Dynamical-) Statistical Downscaling



# (Dynamical-) Statistical Downscaling

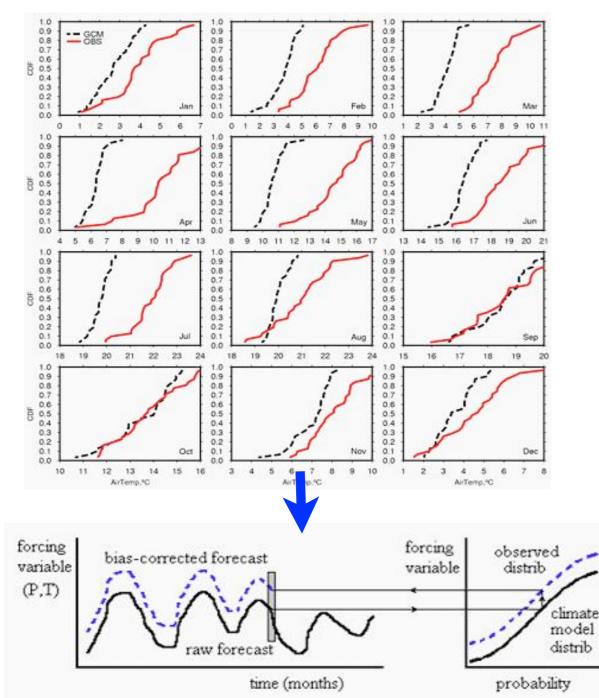
Simple Statistical Downscaling: Bias Correction Spatial Downscaling (BCSD) Wood et al. 2004, and Maurer 2007

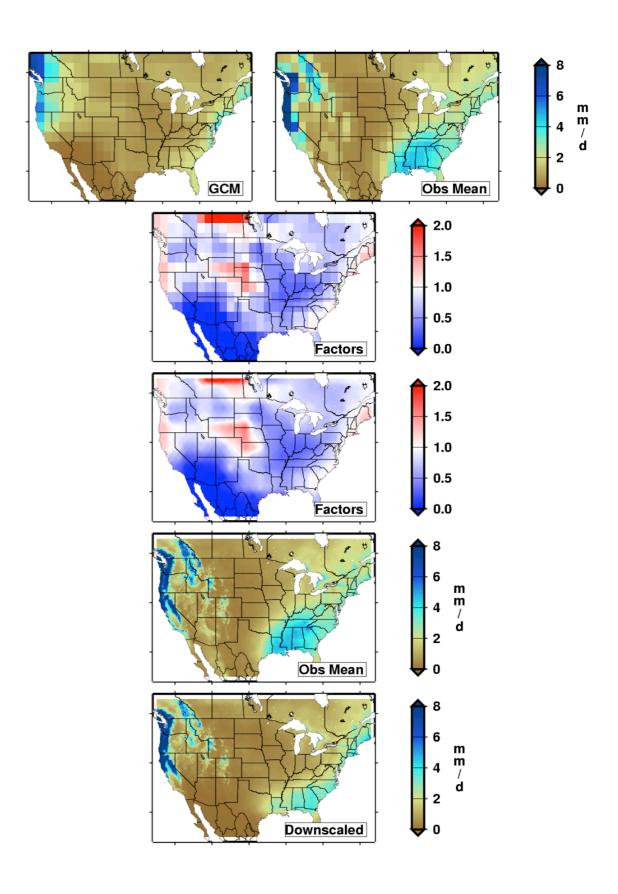
- Aggregate gridded OBS to GCM resolution
- Remove trend (if the trend is significant)
- Generate CDF of observed and GCM data
  - Q-Q mapping approach
  - limitation on extrapolation
- Add trend back in
- Resample/interpolate to finer resolution
- Apply spatial factor to account for subgrid topography

# (Dynamical-) Statistical Downscaling

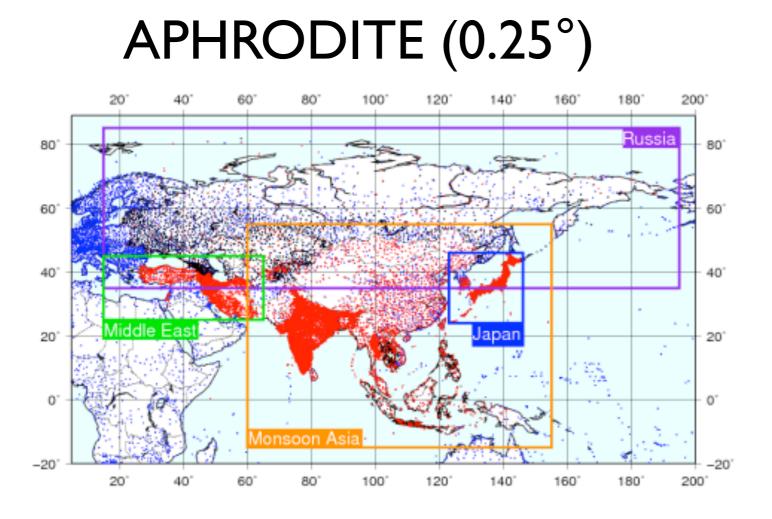
Statistical downscaling and bias correction by cumulative distribution function and interpolation

Wood et al. 2004, and Maurer 2007





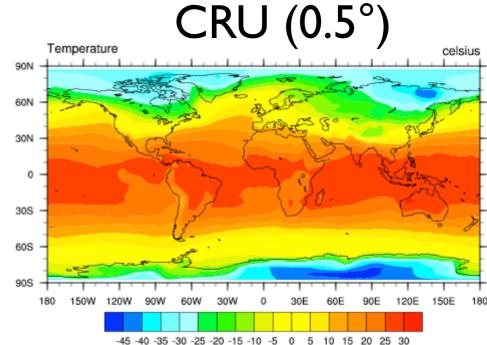
#### Require long-term high-resolution observations



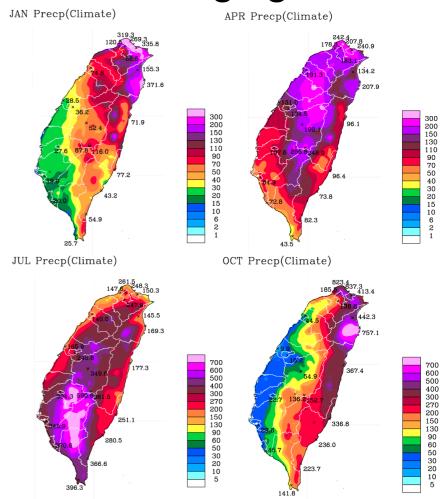
Current version: V1003R1 (Download) »Re

ad) »Readme »Errata

Name	Domain	Resolution	Period		
Monsoon Asia (MA)	60°E-150°E, 15°S-55°N		1951-2007		
Middle East (ME)	15°E-65°E, 25°N-45°N	0.5° and 0.25°, daily			
Russia (RU)	15°E-165°W, 34°N-84°N	,			

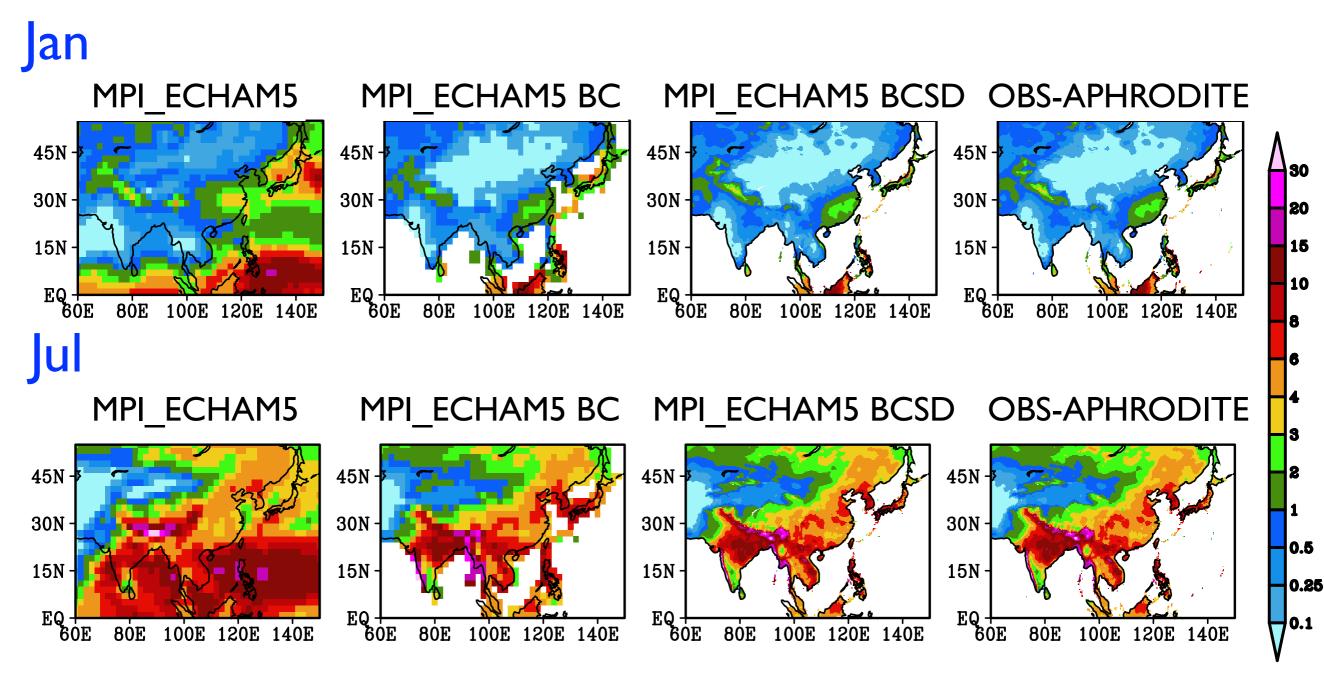


#### Automatic rain gauges in Taiwan



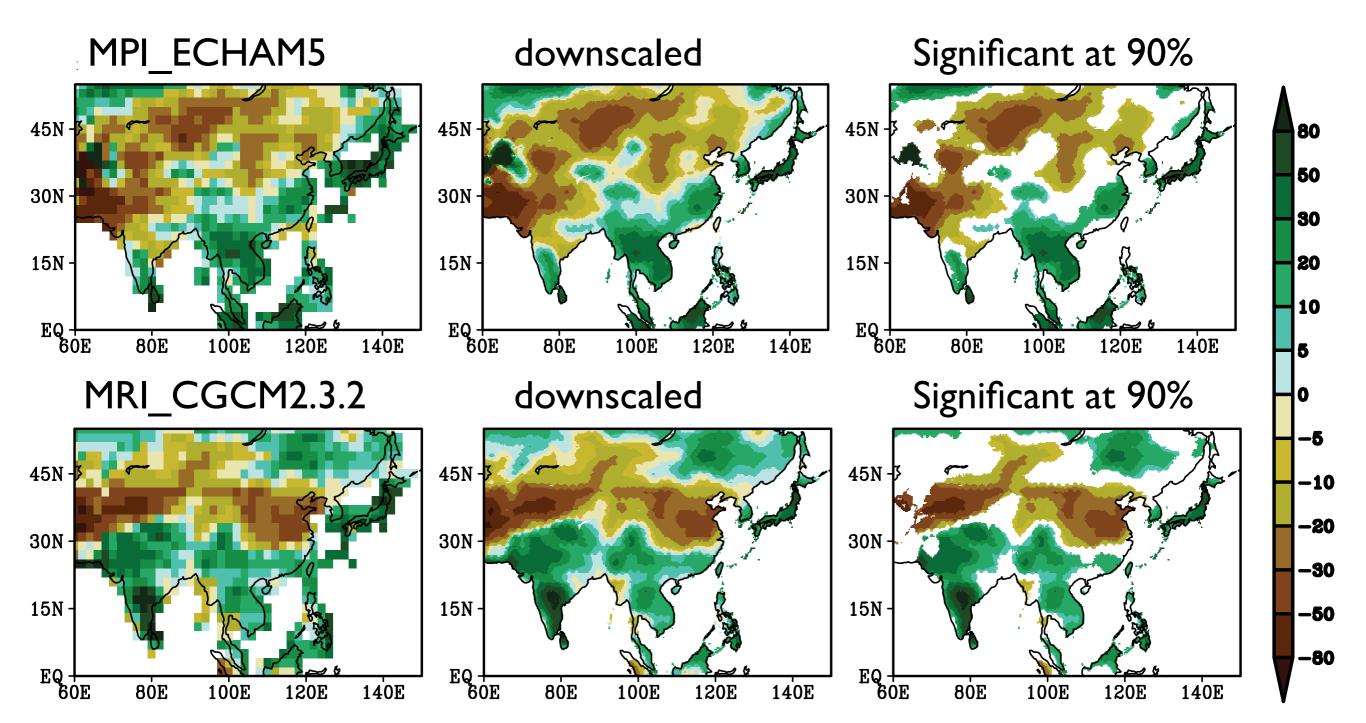
# Validation

• Bias corrected and downscaled of current climate using APHRODITE rainfall analysis

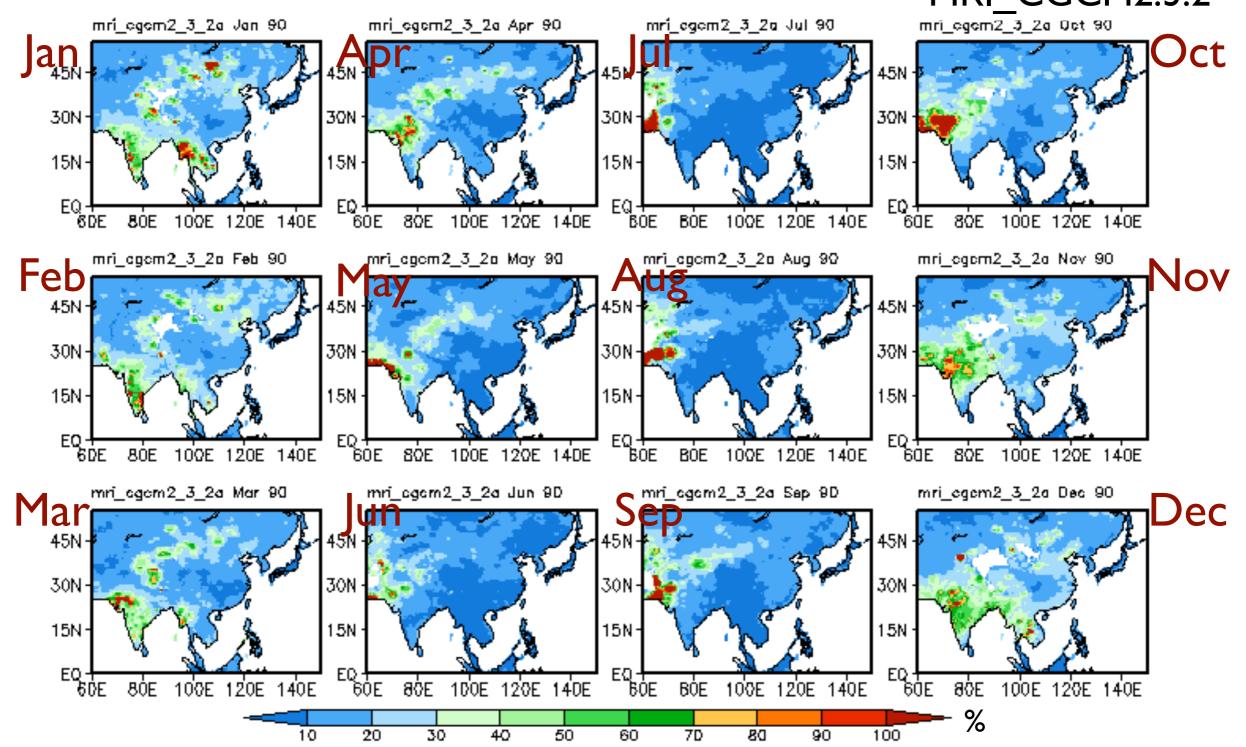


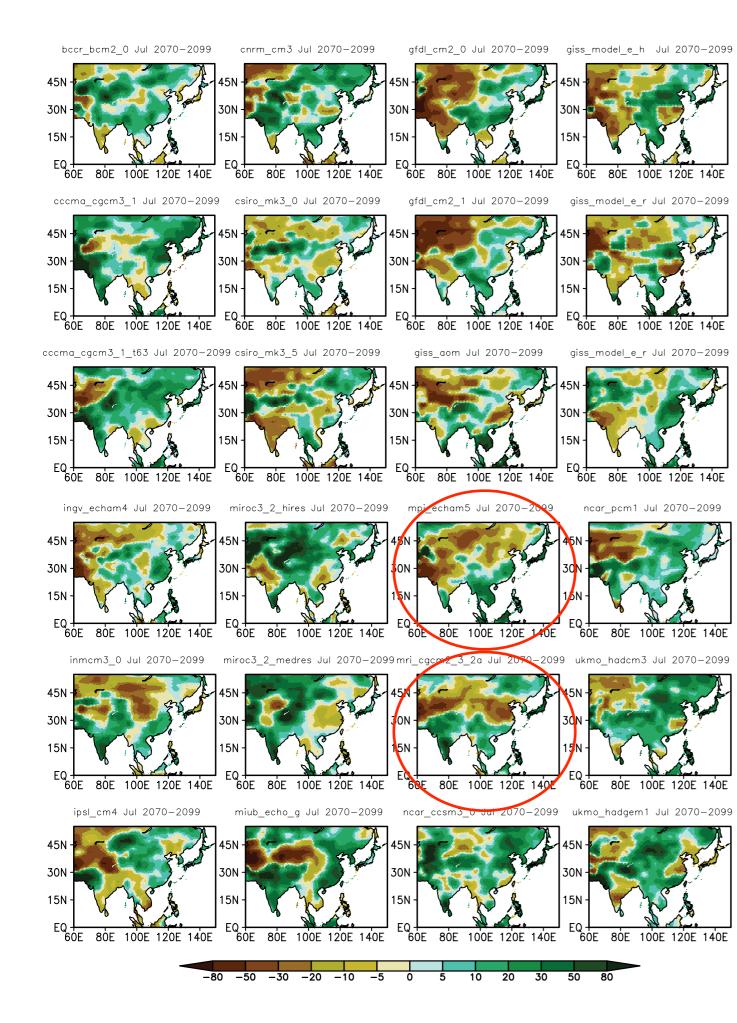
mm/day

- Projected future(2070-2099) change(%) in July precipitation
- Downscaling projected change to 0.25° resolution
- Significant test on the downscaled change by bootstrapping 30 out of 40 years data from present climate



- 90th percentile of downscaled error estimate from bootstrapping 30 out of 40 years data from present climate
- Typically less than 10% error with regional monthly rainfall more than 2 mm/day MRI\_CGCM2.3.2





- Downscaled projected future(2080-2099) precipitation change(%) in July of 24 IPCC AR4 AOGCMs to 0.25° resolution under AIB scenario
- How to deal with model uncertainty?

# Approaches to describe the uncertainty from local model responses

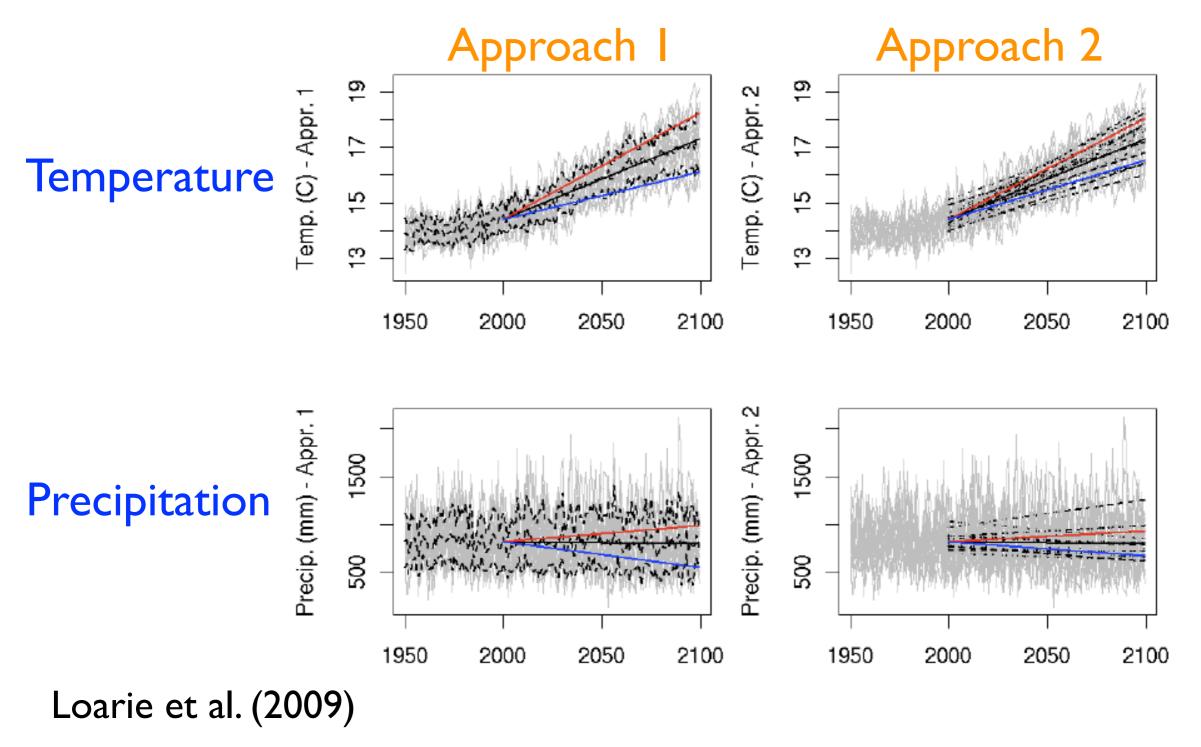
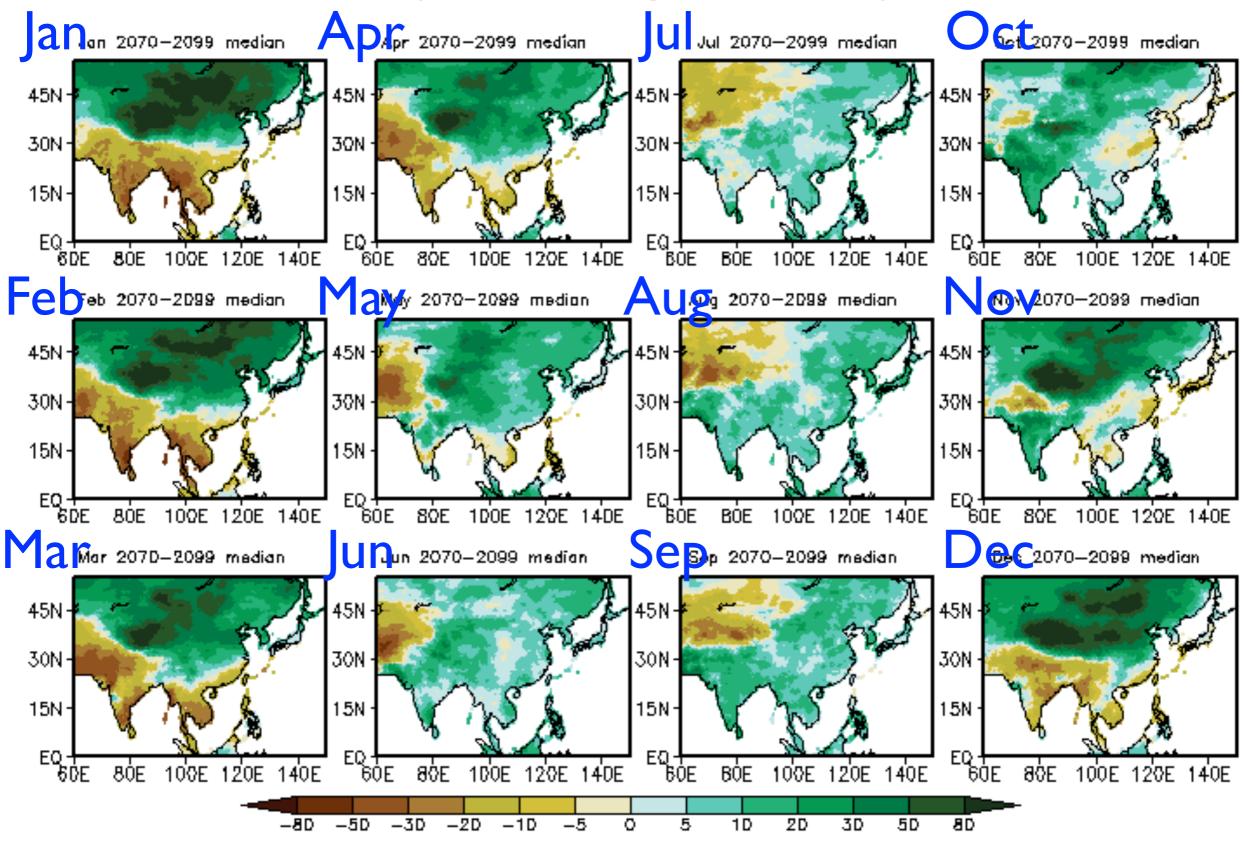
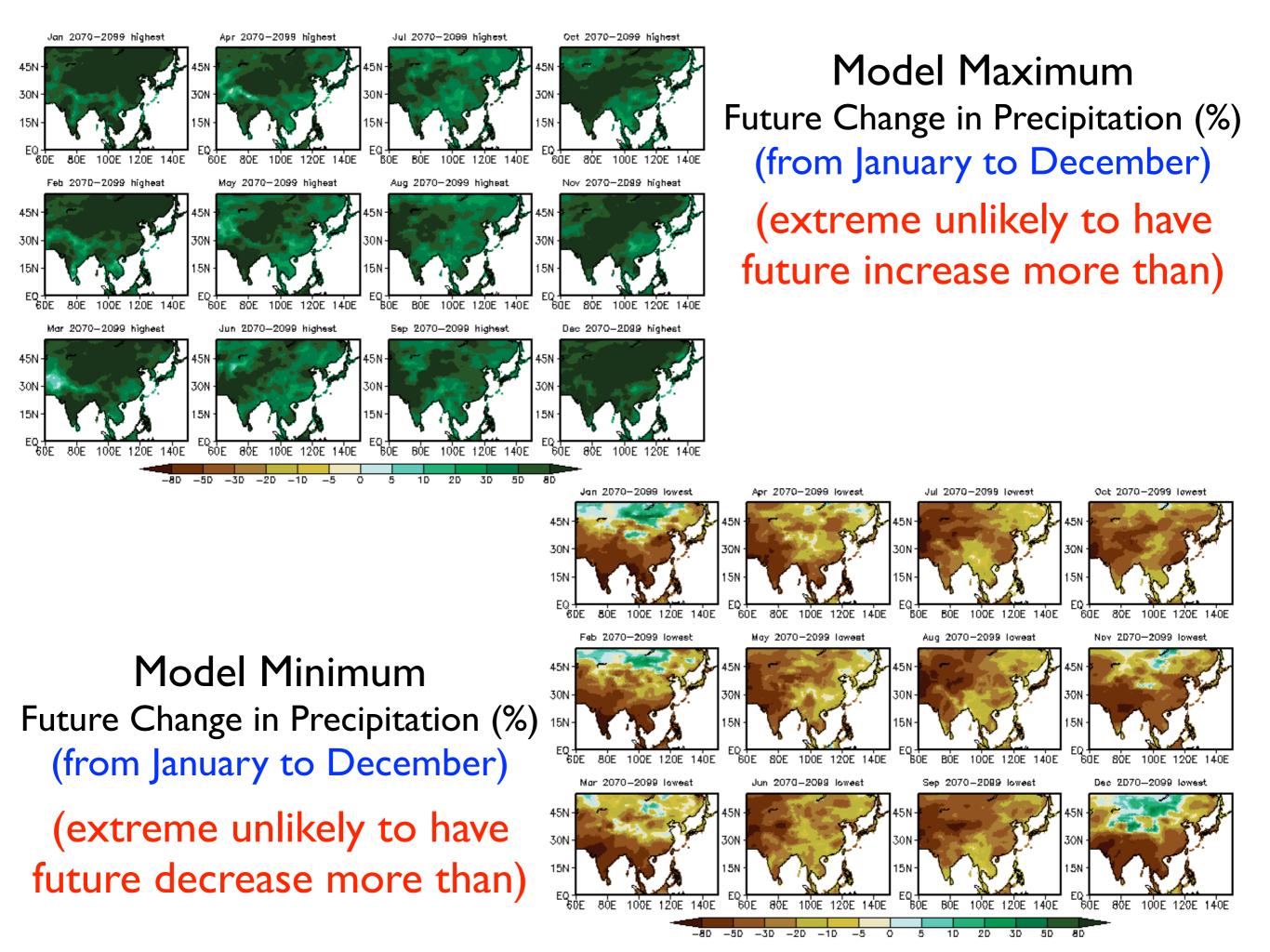


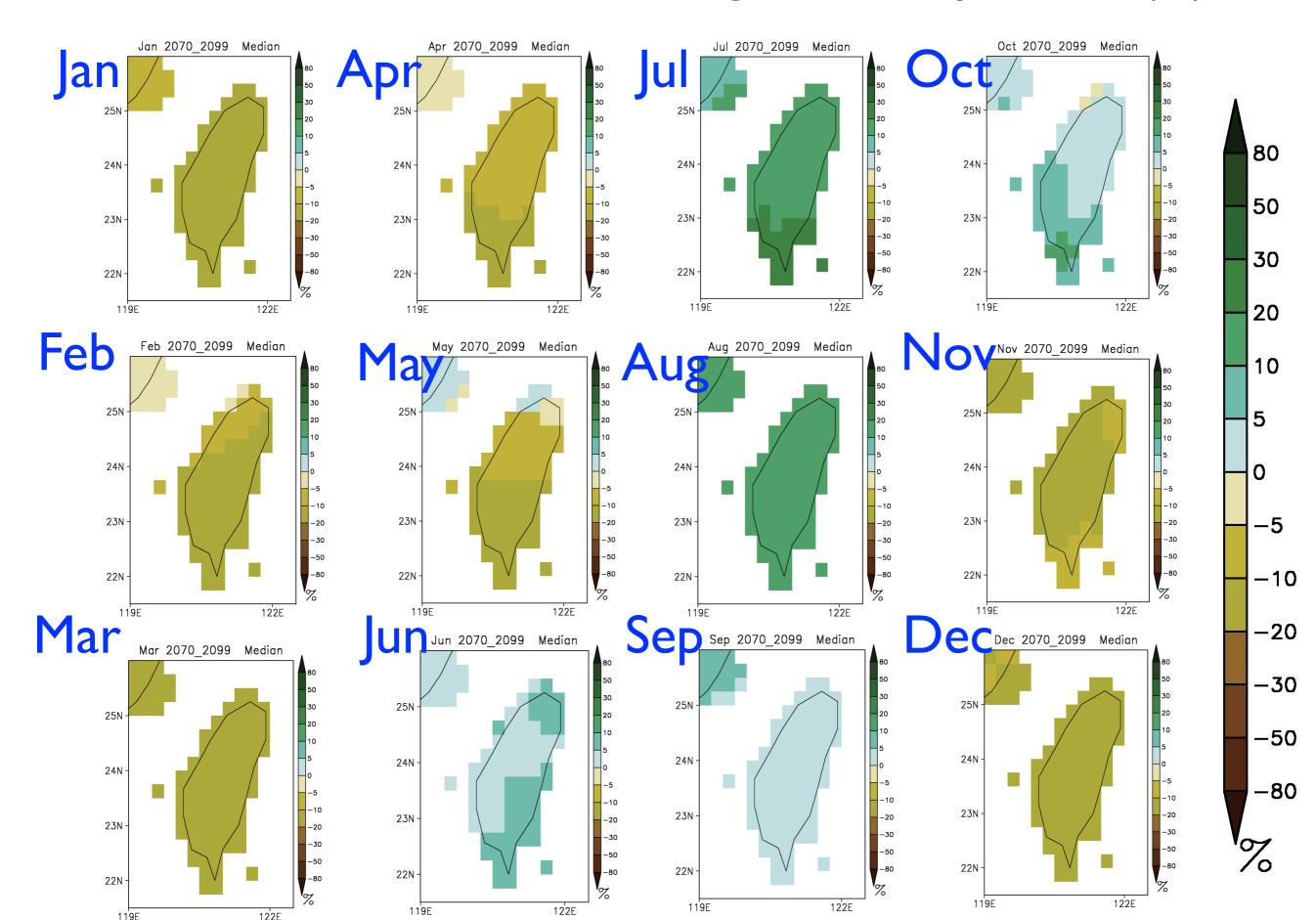
Figure S6 Comparison of temporal gradients using two approaches for a single pixel in California.

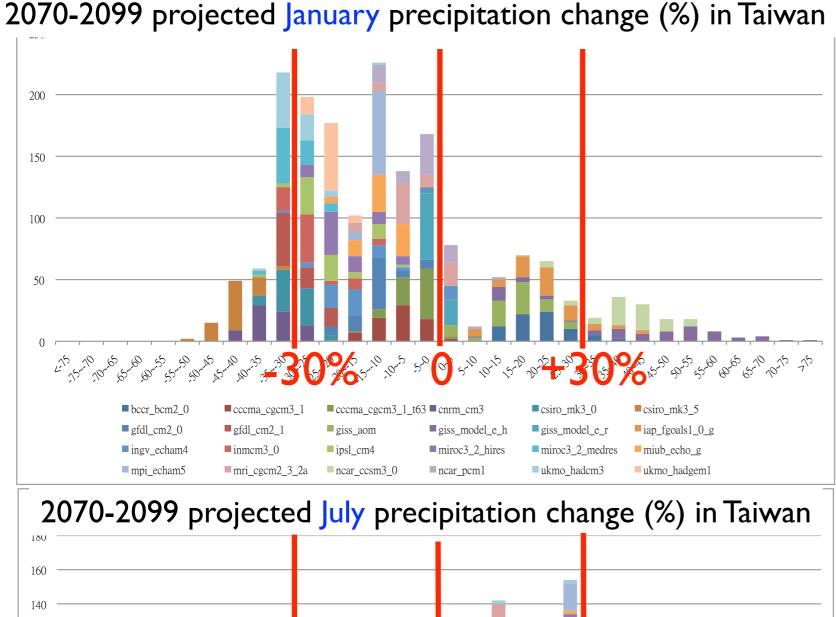
#### Model Median Future Change in Precipitation (%) (more likely than not)





#### Model Median Future Change in Precipitation (%)





25000-15-10 10-5

giss\_aom

■ ipsl cm4

ncar ccsm3 0

cccma\_cgcm3 1

mri\_cgcm2\_3\_2a

gfdl cm2 1

inmcm3 0

giss\_model\_e\_h

miroc3 2 hires

ncar pcm1

cccma\_cgcm3\_1\_t63 cnrm\_cm3

csiro\_mk3\_5

miub\_echo\_g

ukmo hadgem1

iap\_fgoals1\_0\_g

csiro\_mk3\_0

giss\_model\_e\_r

ukmo hadcm3

miroc3\_2\_medres

120

100

80

60

40

20

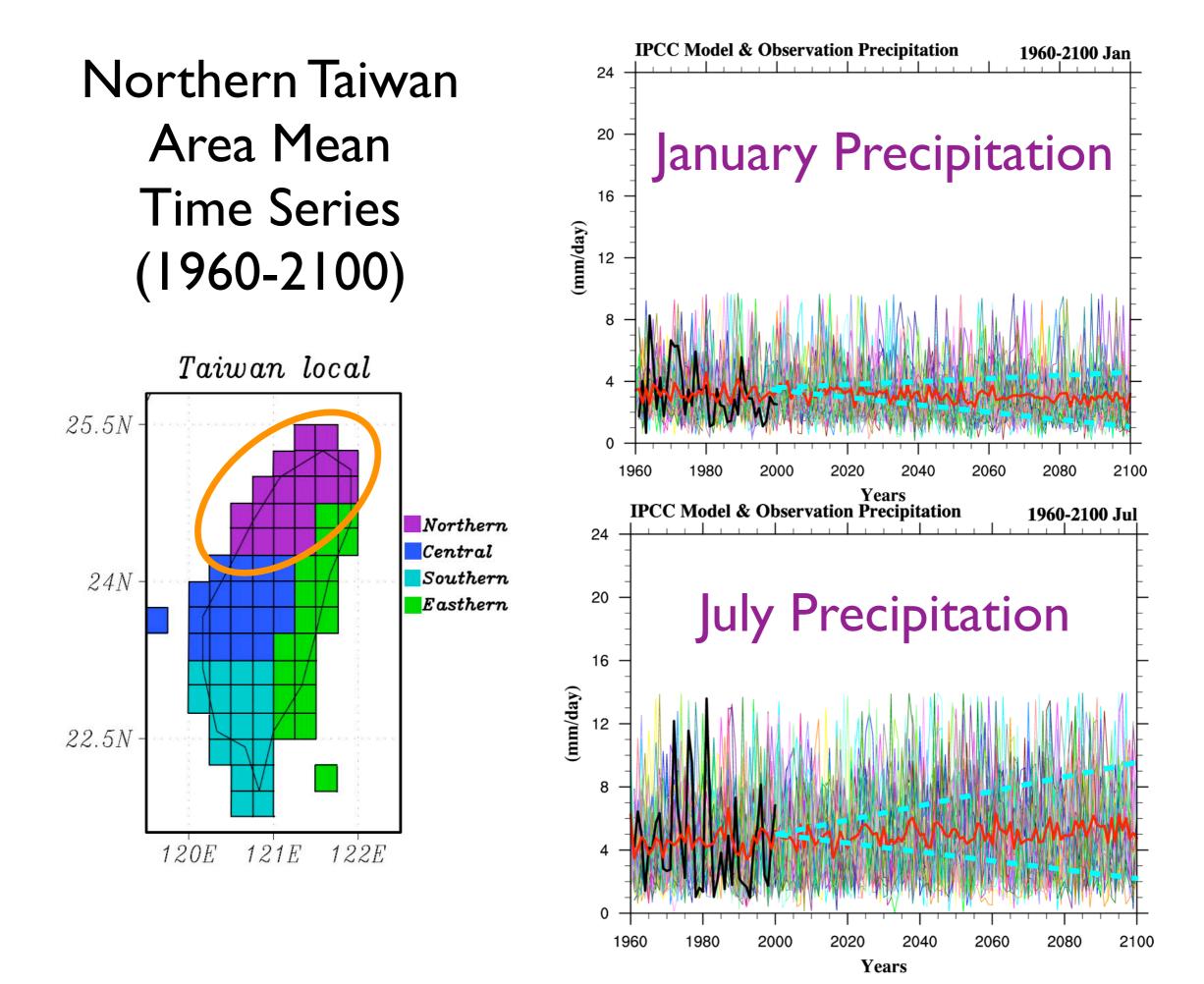
bccr\_bcm2\_0

■ gfdl\_cm2\_0

■ ingv echam4

mpi echam5

- One can also focus on area of interests and construct PDF for projected future precipitation from all models (Taiwan: 75 grids x 24 models)
  - Ensemble of opportunity (probability)



#### Regional averages of temperature and precipitation projections

		Temperature Response (deg C)						Precipitation Response (%)							
Region	season	min	10	25	50	75	90	max	min	10	25	50	75	90	max
Northern Taiwan	DJF	1.9	1.9	2.I	2.7	3.I	3.4	3.8	-44	-29	-20	-13	-3	7	33
	MAM	Ι.7	1.8	2.2	2.6	2.9	3.4	3.9	-31	-24	-14	-8	6	17	36
	JJA	1.7	1.8	2.2	2.7	3.0	3.3	4.0	-15	-12	-1	14	29	46	64
	SON	1.6	1.8	2.2	2.6	3.I	3.3	3.8	-33	-25	-10	8	21	28	34
Central Taiwan	DJF	1.8	1.9	2.0	2.6	3.0	3.3	3.7	-49	-33	-22	-15	-4	6	22
	MAM	1.6	1.8	2.I	2.6	2.9	3.3	3.8	-36	-25	-16	-10	3	17	41
	JJA	I.8.	1.8	2.2	2.7	3.0	3.2	4.0	-15	13	2	14	26	64	69
	SON	1.6	I.8.	2.2	2.6	30	3.3	3.7	-34	-23	-7	11	25	31	45
Southern Taiwan	DJF	1.7	1.7	2.0	2.5	2.9	3.2	3.4	-47	-34	-22	-13	-5	5	8
	MAM	1.5	1.8	2.0	2.5	2.8	3.0	3.6	-41	-26	-21	-14	-5	22	34
	JJA	Ι.7	1.7	2.2	2.5	2.9	3.2	4	-20	-19	7	16	26	69	76
	SON	I.5	I.7	2.I	2.6	2.9	3.I	3.6	-28	21	-8	13	25	36	55
Eastern Taiwan	DJF	1.8	1.8	2.0	2.6	3.0	3.3	3.7	-44	-31	-20	-12	-3	5	17
	MAM	I.5	1.8	2.I	2.5	2.8	3.3	3.8	-37	-25	-18	-11	Ι	20	36
	JJA	1.7	1.7	2.2	2.6	2.9	3.2	4.0	-17	-15	3	15	26	57	64
	SON	١.6	I.8.	2.I	2.6	3.0	3.2	3.7	-30	-23	-10	10	23	33	43

## Summary and Concluding Remarks

- Must consider the other major uncertainties (emission scenario, model, etc.) regarding future climate in addition to downscaling to local scale. Probabilistic projection can better represent the uncertainty.
- Large resources are needed for dealing with all the uncertainties using dynamical downscaling approach. Statistical approach is a relatively simple alternative.
- Although the uncertainties can be more easily included with statistical downscaling approach, one should aware about the assumption, limitation and caveats of this type of climate information regionalization tool:
  - long-term high-resolution observation availability

.....

- statistical relationship between model data and observation remains valid for periods outside calibration period
- only limited area with local change passed statistical significance test