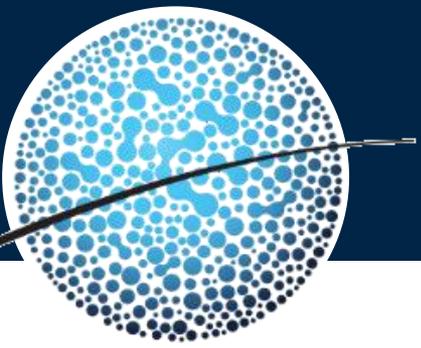


Climate Change Impact on Agricultural Drought by Considering Uncertainty of CMIP5

2014.1.15.

Jaepil Cho

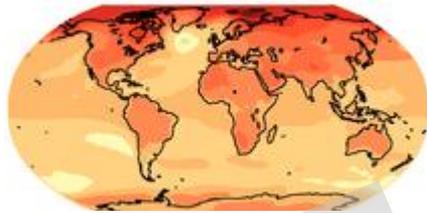
APEC Climate Center



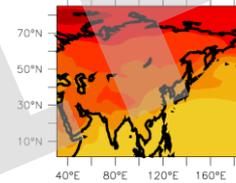
Presentation overview

- 1. Things to be considered in using climate information**
- 2. Downcaling and Bias Correction of Daily CMIP5 Data**
- 3. Analysis of Precipitation Index**
- 4. Assessment of Climate Change Impact on Agricultural Reservoirs**
- 5. Suggestions**

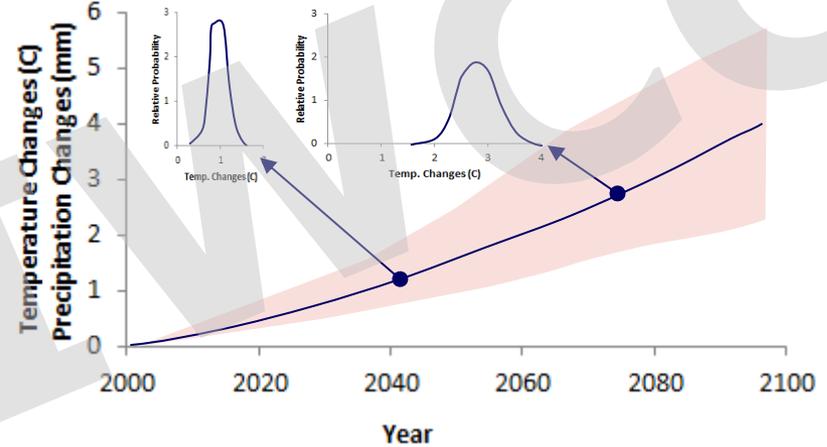
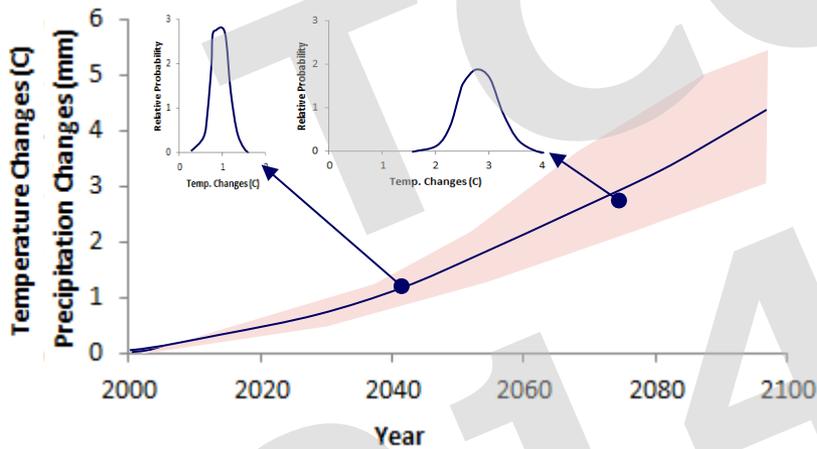
Uncertainty in Climate Change Research



Global Climate Change



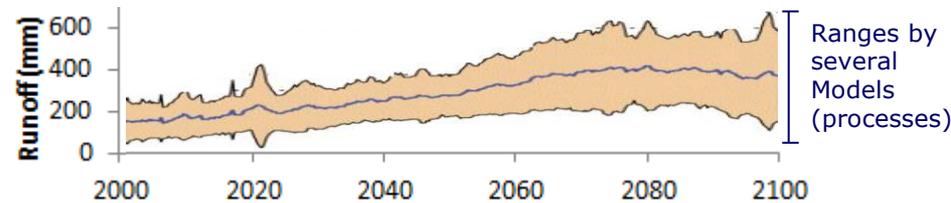
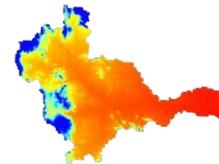
Regional Downscaling



Year



Watershed-scale Hydrology Modeling



Ranges by several Models (processes)

Uncertainty = f (GCMs, Time, Domain, Variables, **Processes**)

How well does a climate change scenario reproduce the spatial and temporal pattern during the historical period?

Scenarios	Inflow (mm)	% change
Historical	988.5	
RCP8.5	1078.1	9.1

Even though we have same future projection....

- What should be premised in order to have significant meanings within the future climate change projections?
- Does scenario-based data reproduce the characteristics of rainfall (extreme, spatial/temporal patterns) during the historical period, compared to the observations ?

**Reproducibility
for historical period**

How reliable are projections of future climate change scenarios?

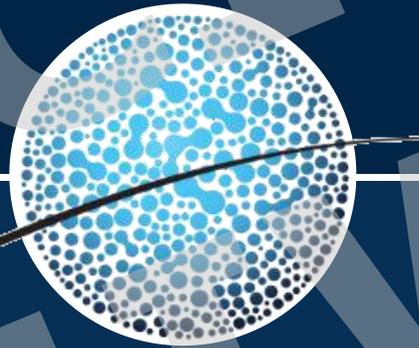
Scenarios	Inflow	% Change
Historical	988.5	
RCP8.5: GFDL-ESM2G	1198.0	21.2
RCP8.5: inmcm4	953.7	-3.5

- **How decisions can be utilized when opposite signals are projected in the same watershed?**
- **Even though MME-based projection shows same projection, what kind of additional information should be provided for decision-making?**

Uncertainty of future projections

Scenarios	Inflow (mm)	% change
Historical	988.5	
RCP8.5	1078.1	9.1

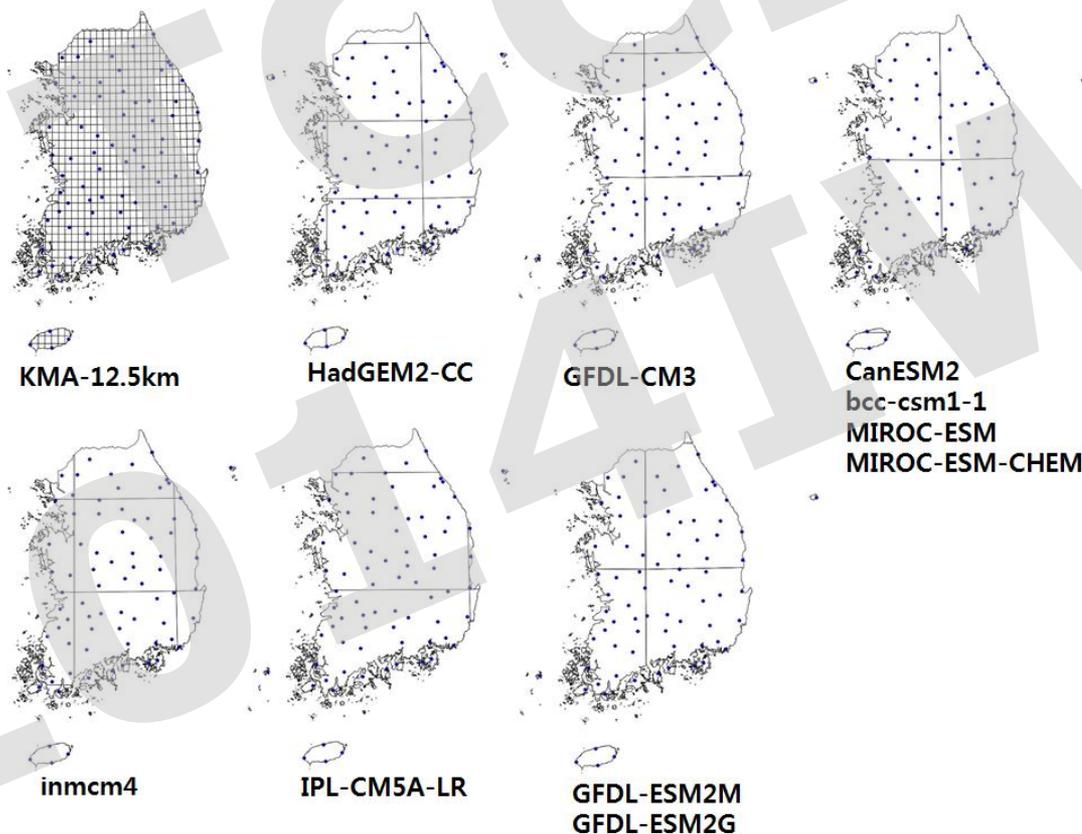
Even though we have same future projection ..



Downscaling and Bias Correction of Daily CMIP5 Data

Used CMIP5 Data

- Daily 6 weather variables (Precipitation, Min/Max Temperature, Wind speed, Relative humidity, Solar radiation) → 1 RCM & 33 GCMs



RCP8.5 Scenario

- KMA RCM
- 10 GCMs

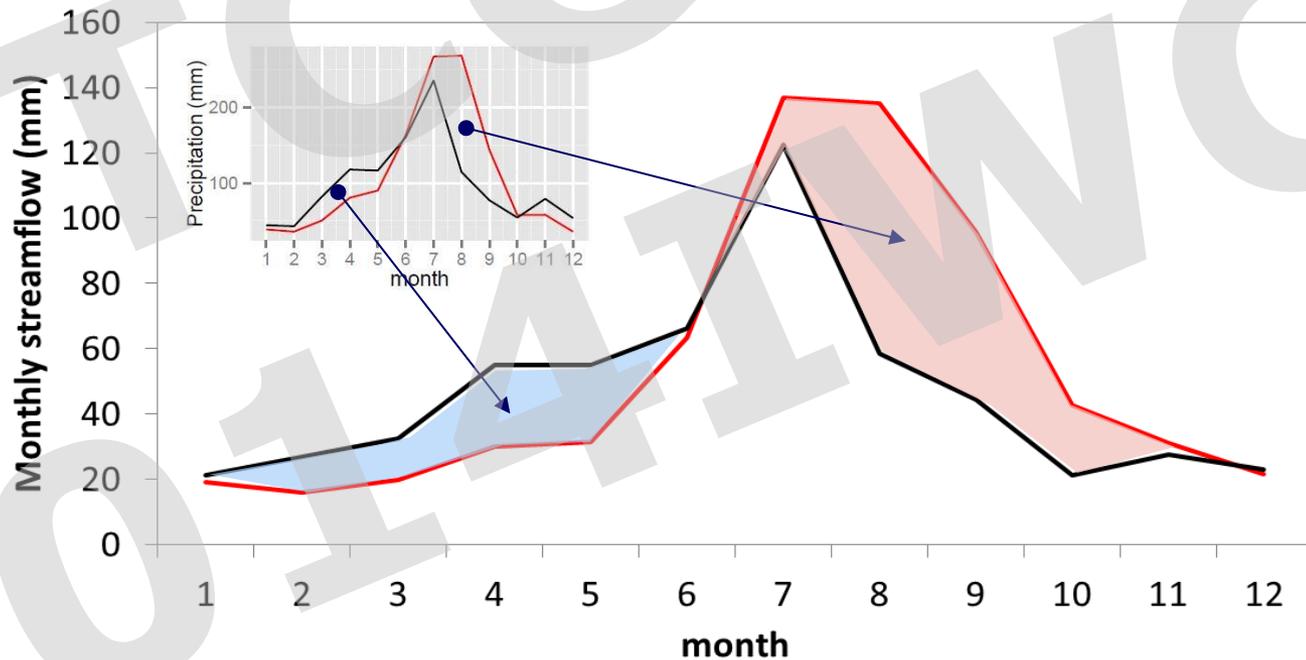
Climate Data

- Precipitation
- Min. temperature
- Max. temperature
- Wind speed
- Relative humidity
- Solar radiation

➡ **Downscaling is required**

Why do we need a bias correction?

- Streamflow simulated using **observed weather input**
- Streamflow simulated using **KMA-12.5km weather input**

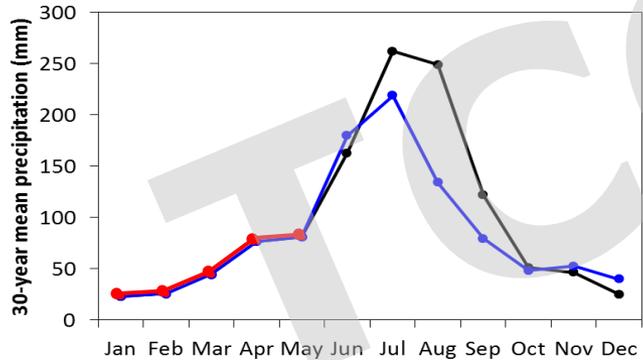


Monthly streamflow pattern follows the precipitation pattern

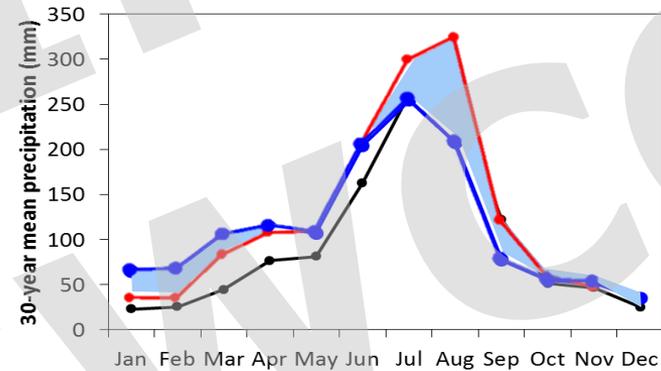
➡ Bias Correction (BC) is necessary

Bias Correction of Daily CMIP5 Data (Non-parametric Quantile Mapping methods)

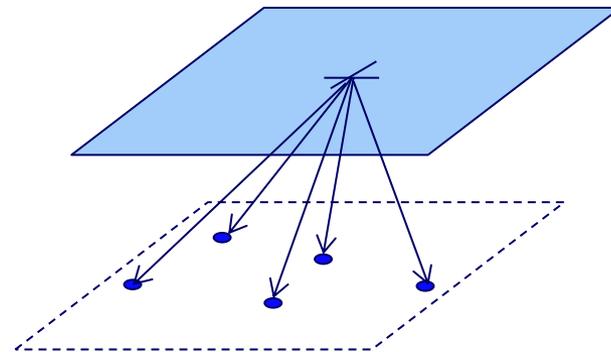
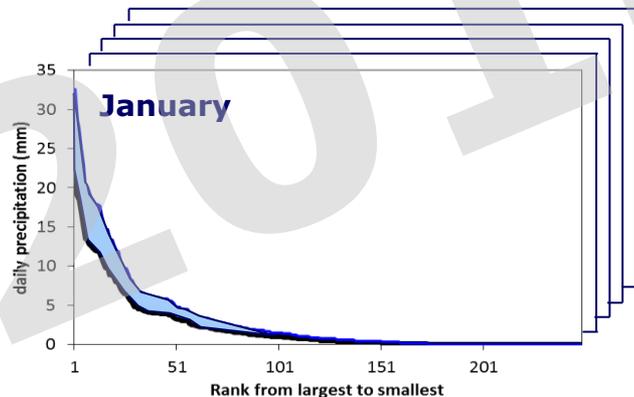
Historical (1976~2005)



Future (2011~2040)

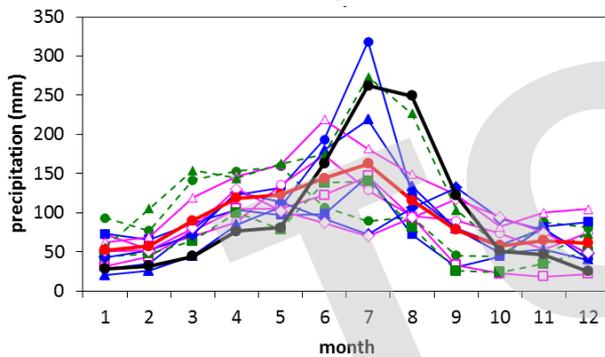


- Observed or Historical
- Before quantile mapping
- **Quantile mapping information**
- After quantile mapping

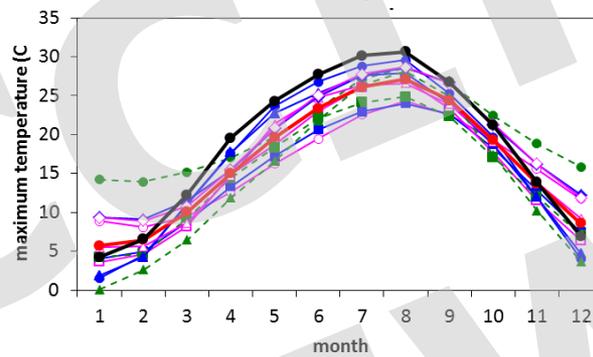


Uncertainties of weather variables before bias corection (Historical, Jeonju)

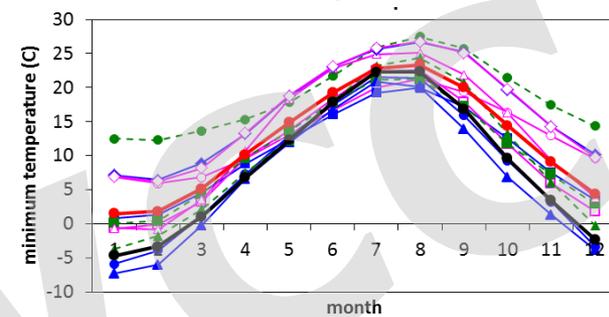
Precipitation



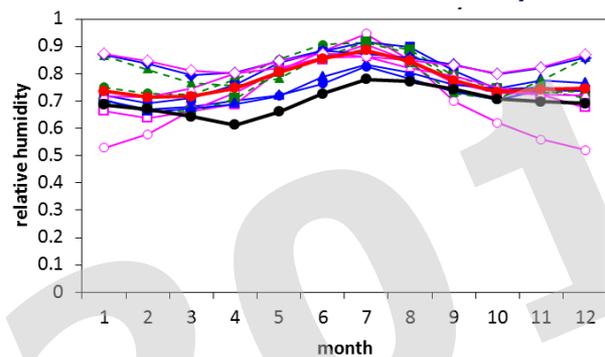
Max. temperature



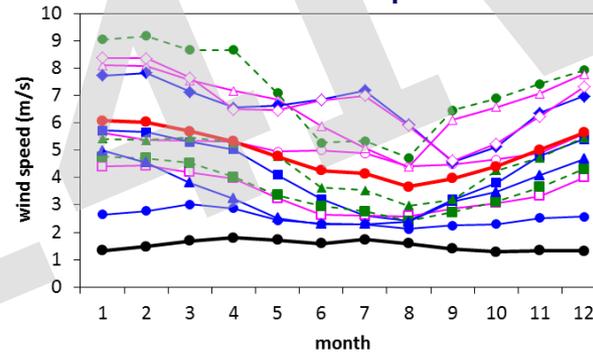
Min. temperature



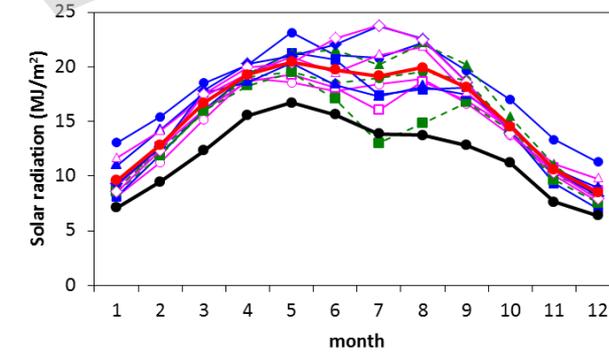
Relative humidity



Wind speed



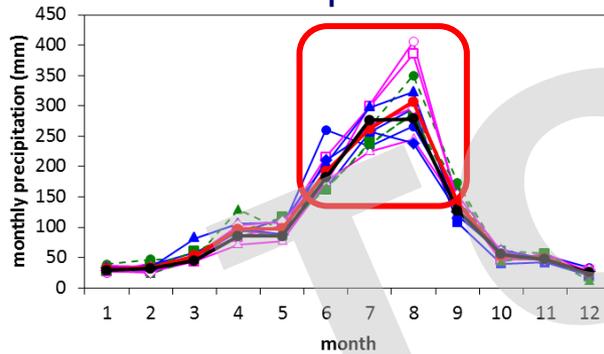
Solar radiation



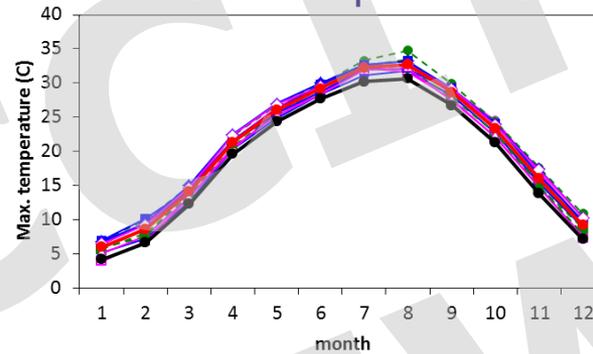
- KMA-12.5km
- GFDL-CM3
- ▲— HadGEM2-CC
- ◆— MIROC-ESM
- bcc-csm1-1
- GFDL-ESM2G
- △— Inmcm4
- ◇— MIROC-ESM-CHEM
- CanESM2
- GFDL-ESM2M
- ▲- IPSL-CM5A-LR
- MME
- Observed

Uncertainties of weather variables after bias corection (RCP8.5, Jeonju)

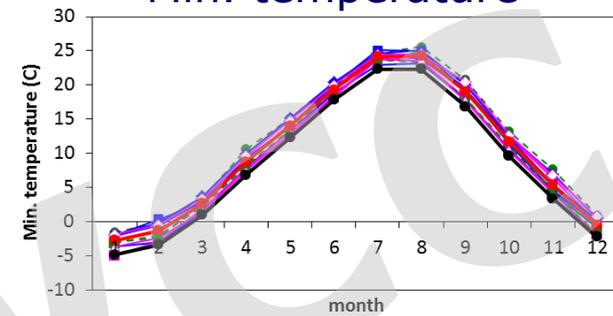
Precipitation



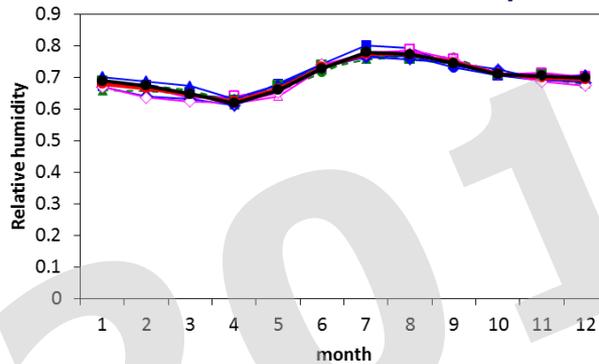
Max. temperature



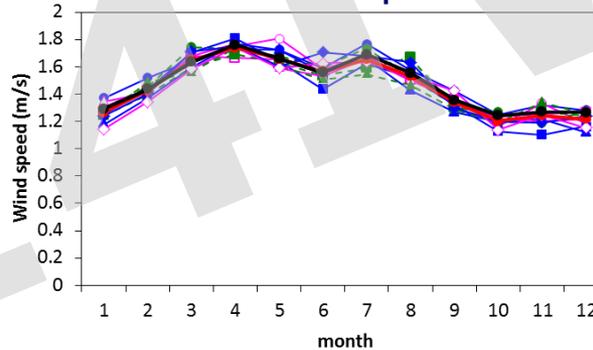
Min. temperature



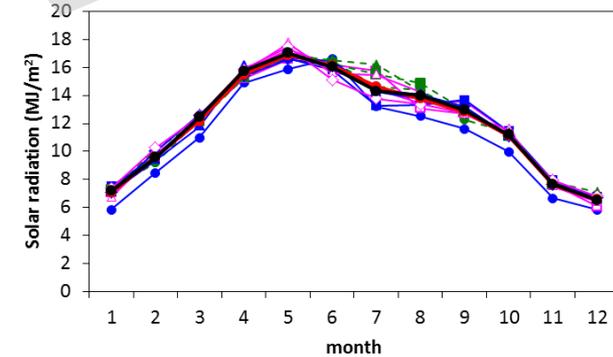
Relative humidity



Wind speed



Solar radiation



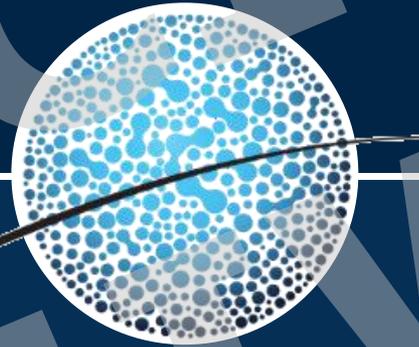
—●— KMA-12.5km
—○— bcc-csm1-1
—◆— CanESM2

—■— GFDL-CM3
—□— GFDL-ESM2G
—■— GFDL-ESM2M

—▲— HadGEM2-CC
—△— Inmcm4
—▲— IPSL-CM5A-LR

—◆— MIROC-ESM
—◇— MIROC-ESM-CHEM
—●— MME

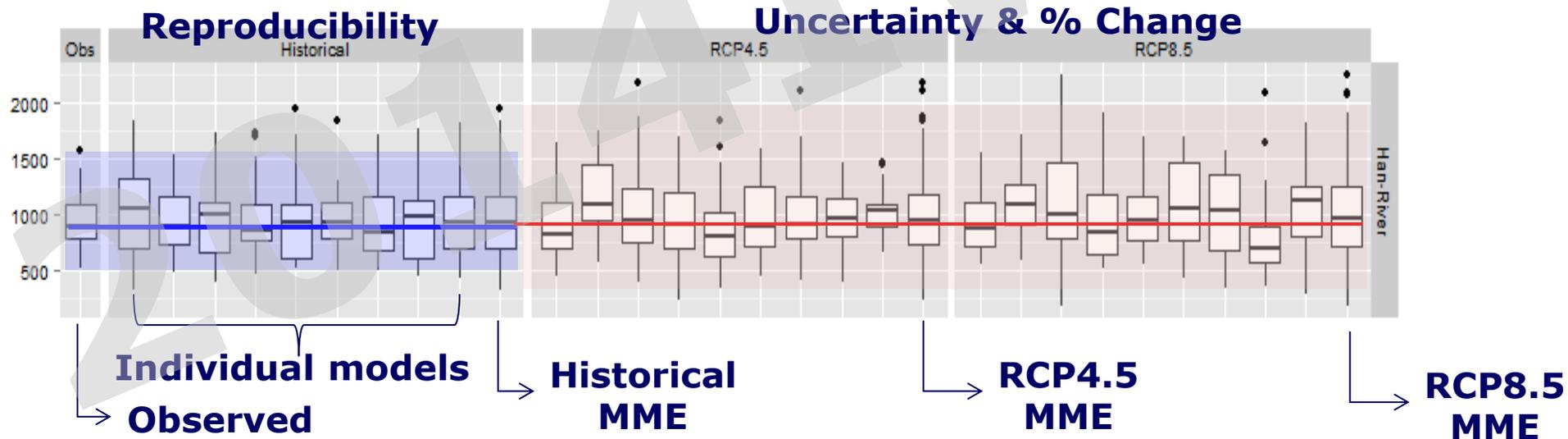
—●— Historical



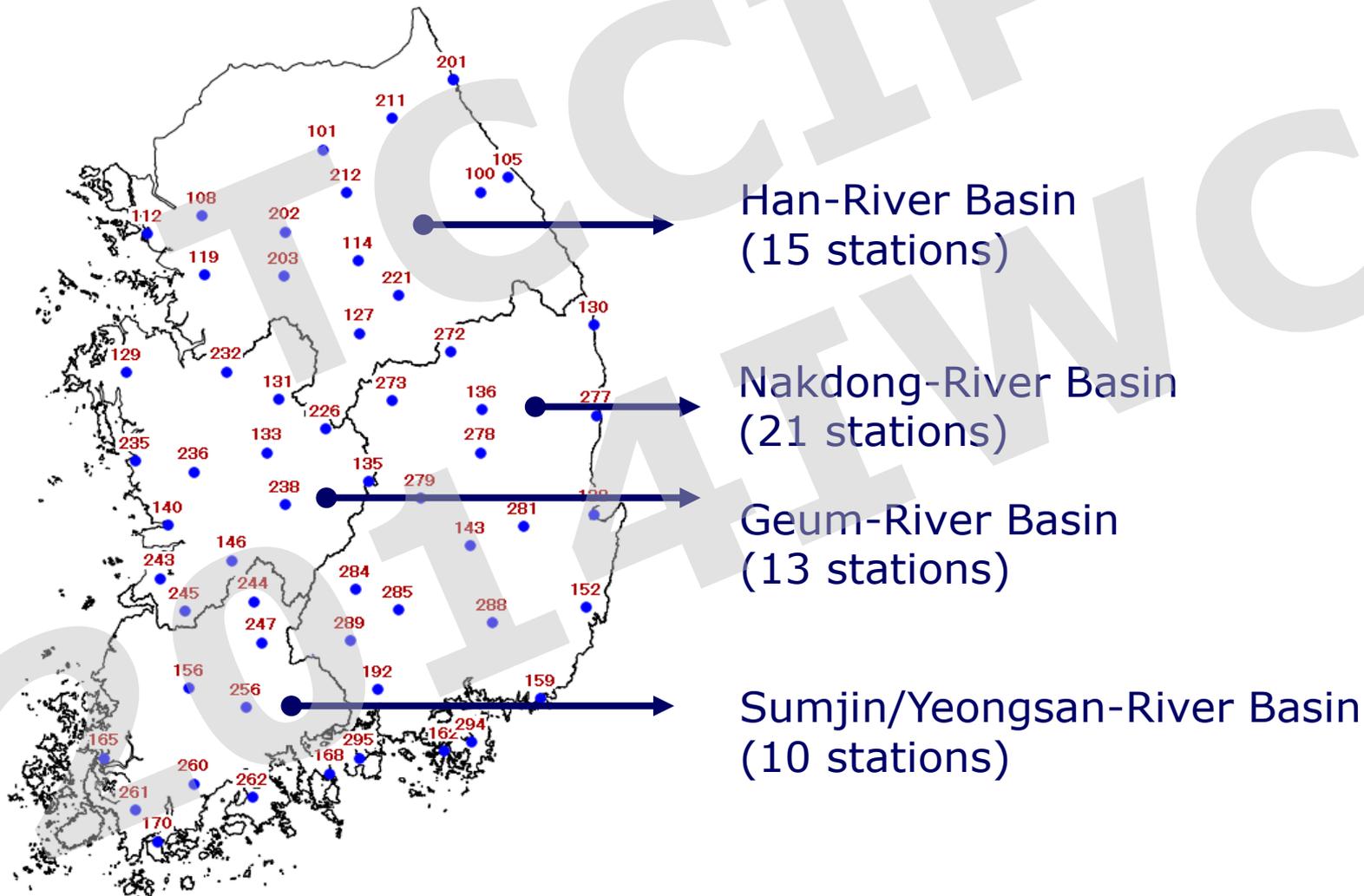
Analysis of Precipitation Index

Precipitation index

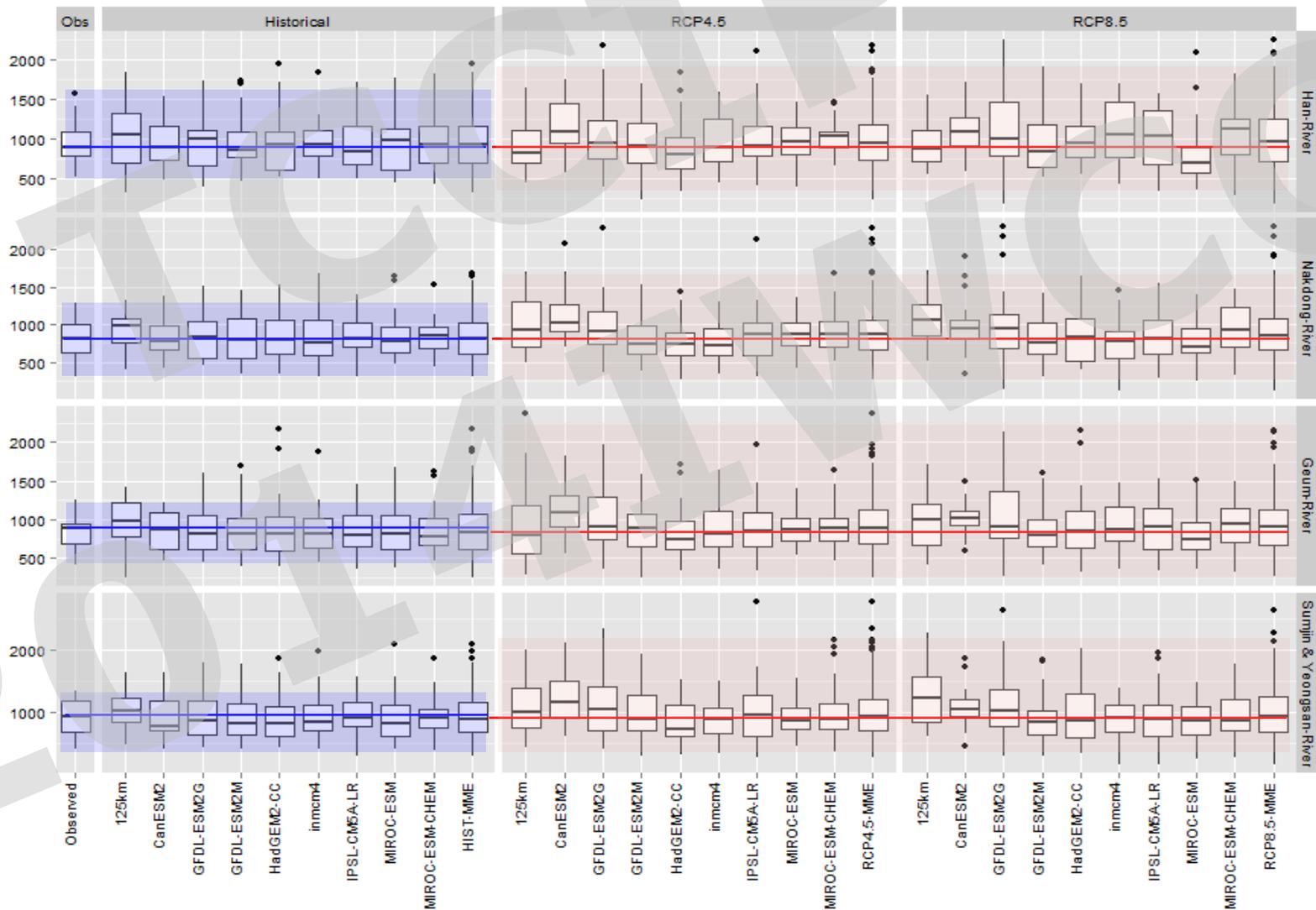
- Wet season (Jun~Sep) precipitation
 - Max. daily precipitation
 - Number of days daily precipitation $\geq 80\text{mm}$
 - Dry season (Oct~May) precipitation
 - Maximum number of consecutive dry days
- } Flood
 } Drought



4-River Basins

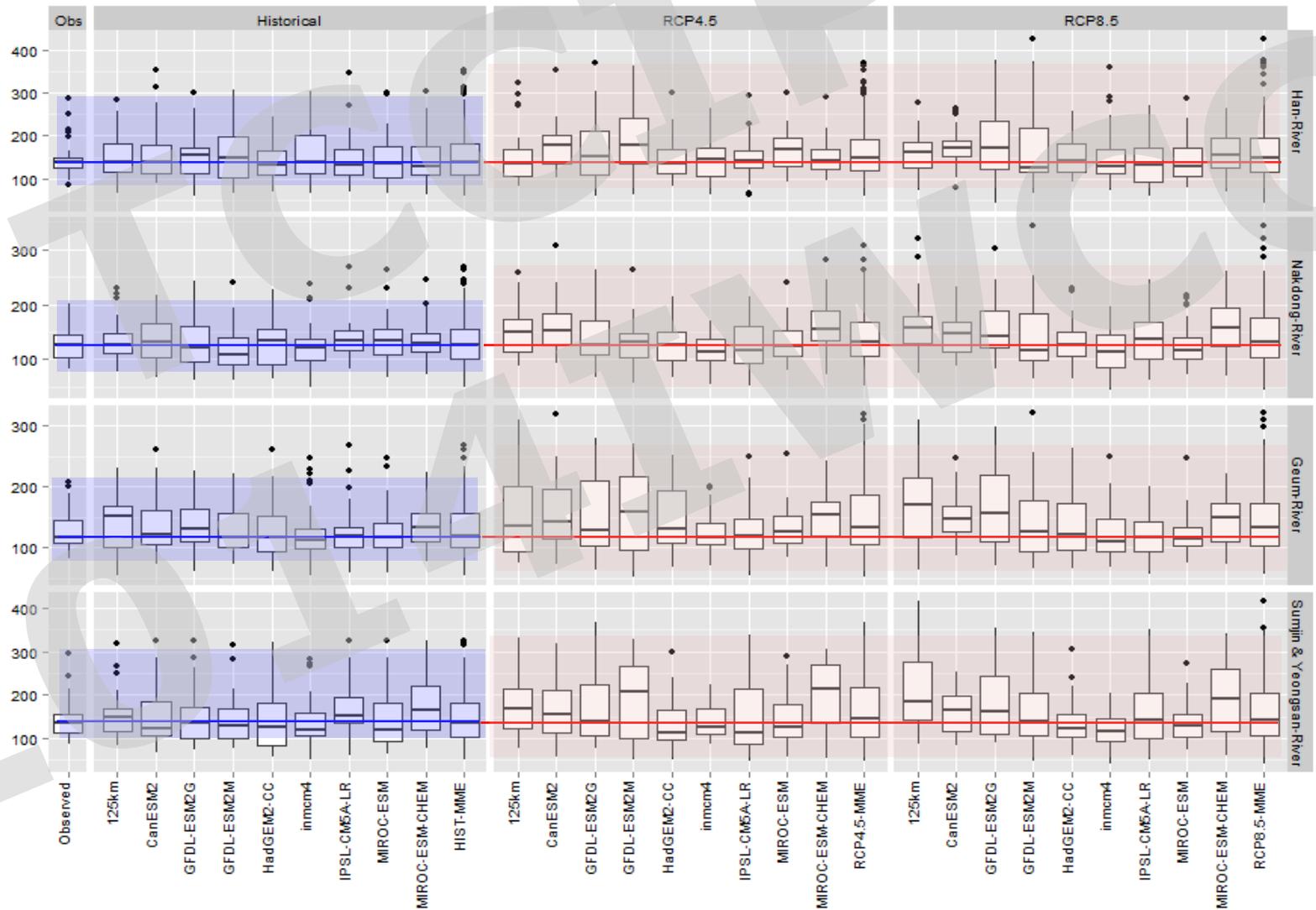


Wet season (Jun~Sep) Precipitation (mm)



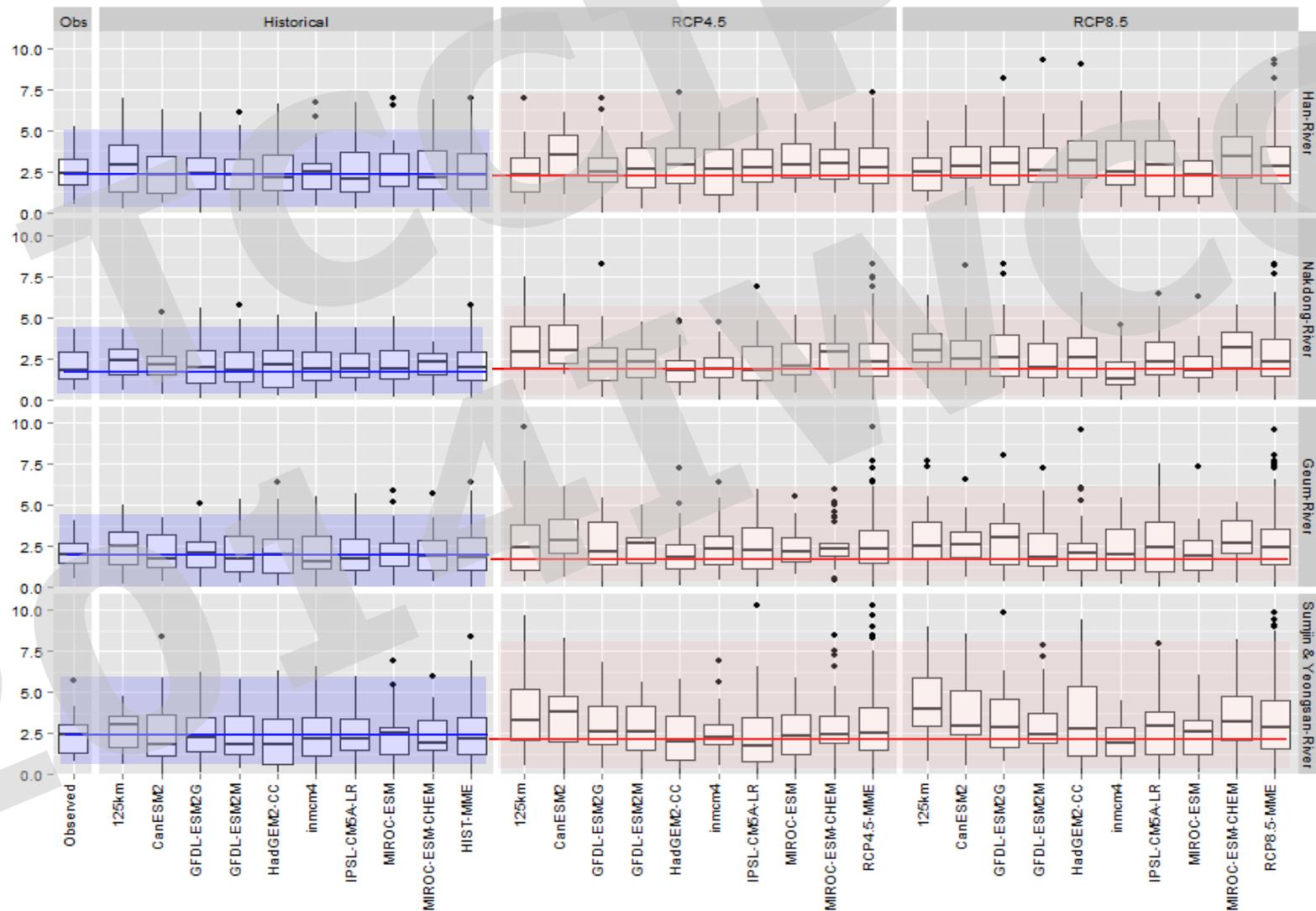
Basins

Daily maximum precipitation (mm)



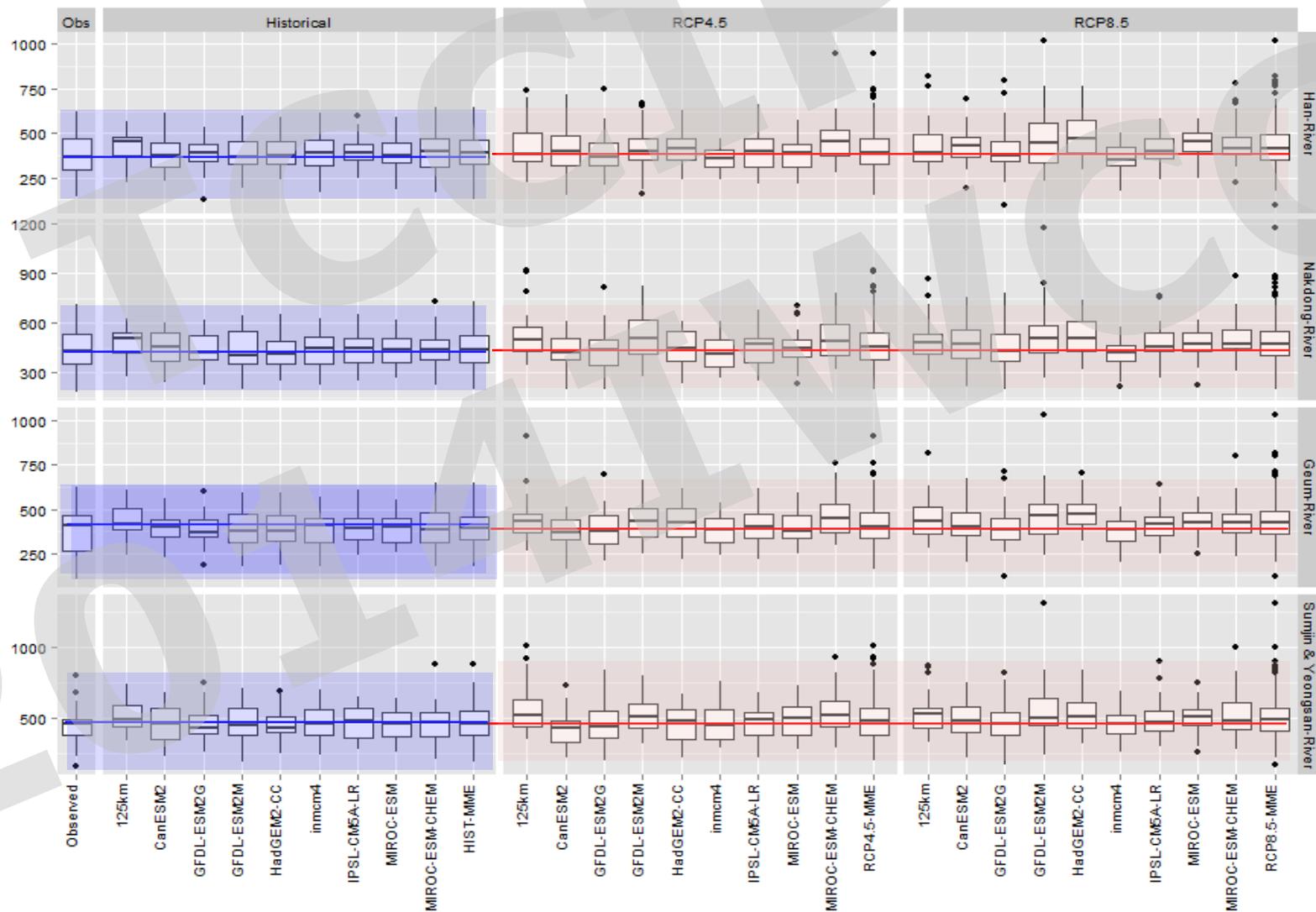
Basins

Number of days daily precipitation $\geq 80\text{mm}$



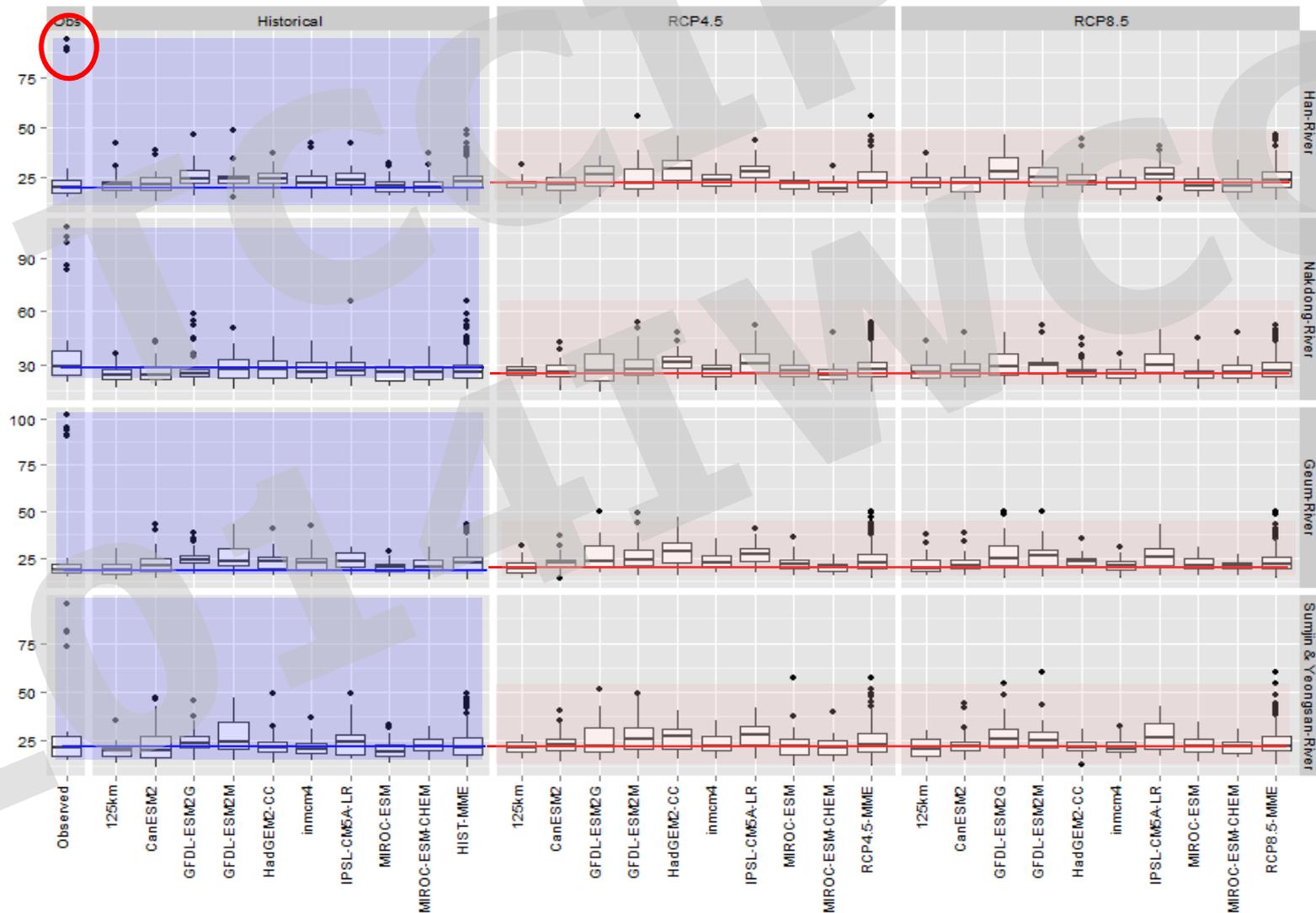
Basins

Dry season (Oct ~ May) Precipitation (mm)

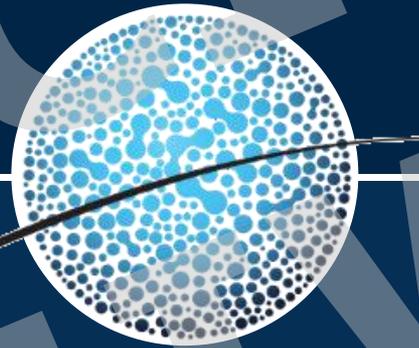


Basins

Maximum dry days = maximum number of consecutive dry days



Basins



Assessment of Climate Change Impact on Agricultural Reservoirs

2014

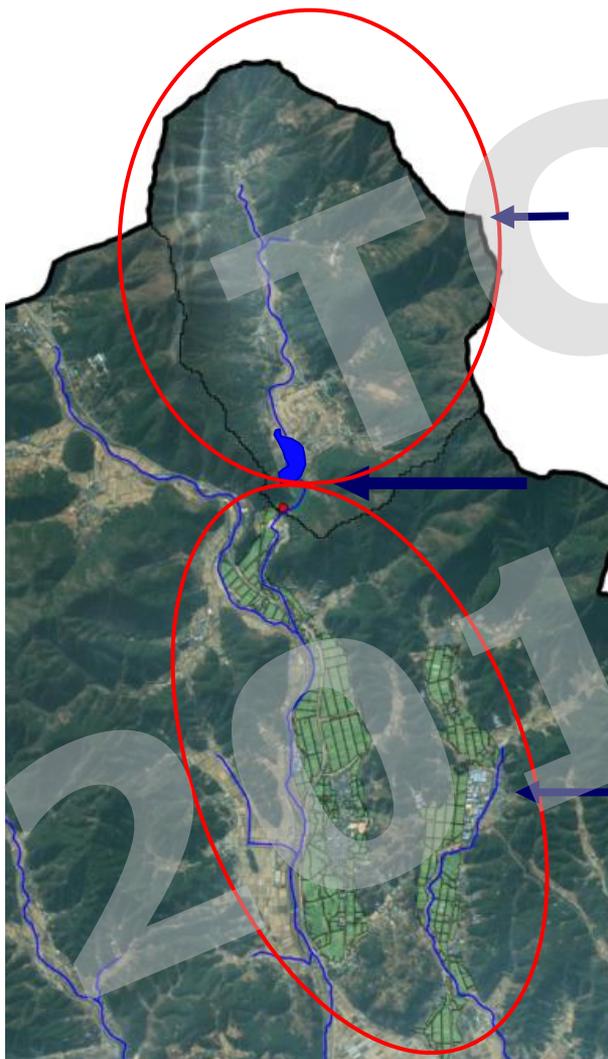
Agricultural reservoirs



Small agricultural reservoirs are vulnerable to climate change

- 62% of total water resources are used for agricultural water (2007)
- 80% of agricultural water are used for paddy irrigation during Apr-Sep.
- 80% of irrigation water for paddy areas are supplied from agricultural facilities
- 56% of total irrigated areas are supplied by agricultural reservoirs

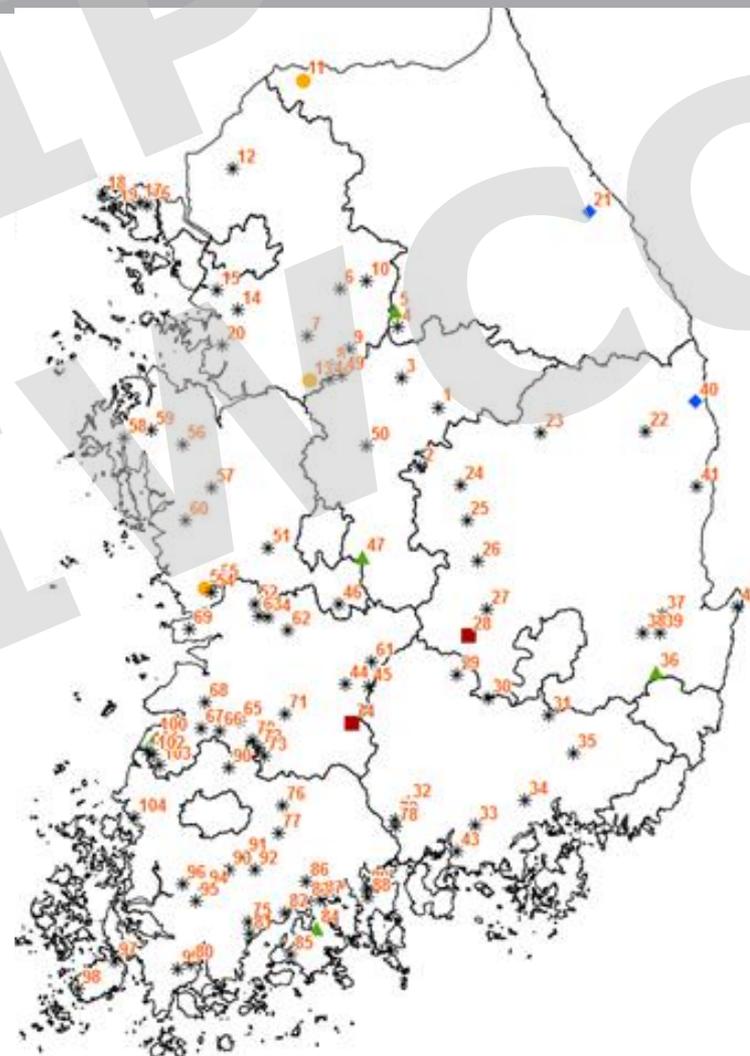
Selected Agricultural Reservoirs



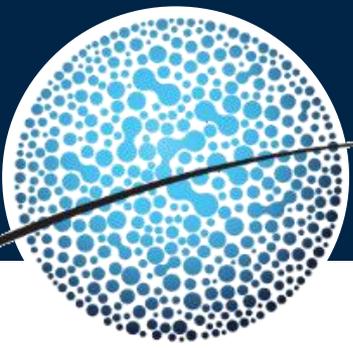
Upstream watershed (Inflow)

Reservoir (storage)

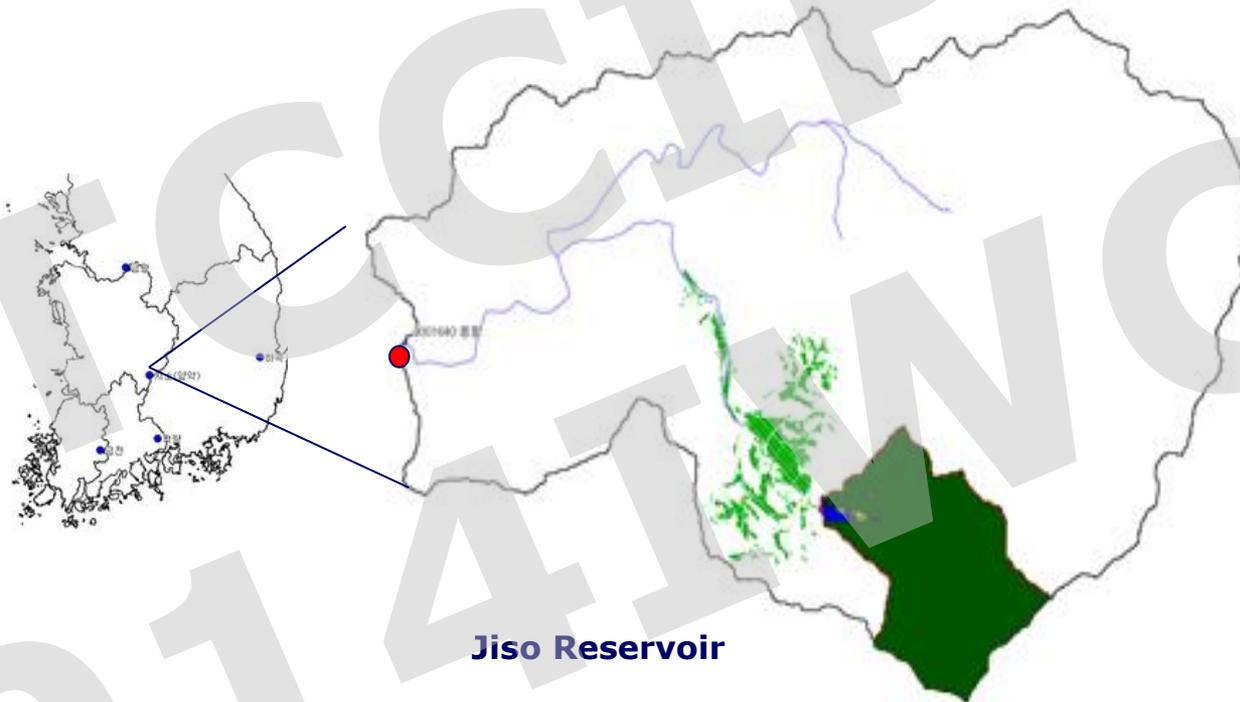
Irrigation (Demand)



104 reservoirs



Representative Reservoir for Reproducibility/Uncertainty Analysis

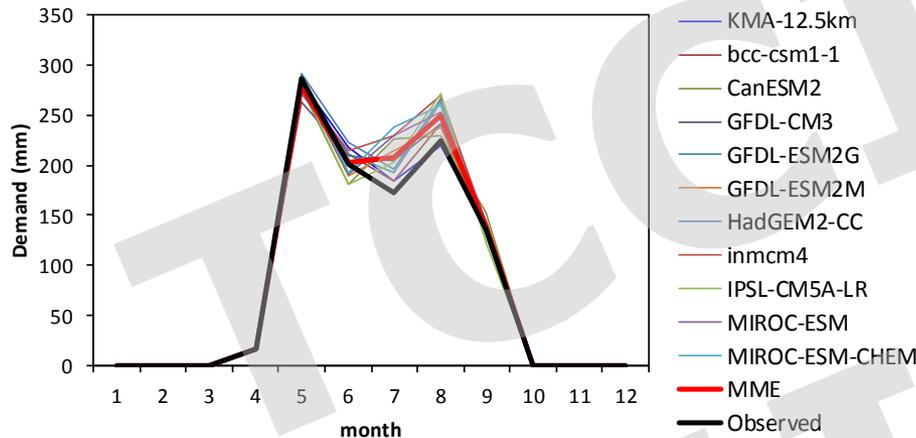


● HOMWRS

- Inflow: 3-Tank model
- Demand: ET + water supply for Sowing, transplanting + loss
- Storage based on water balance

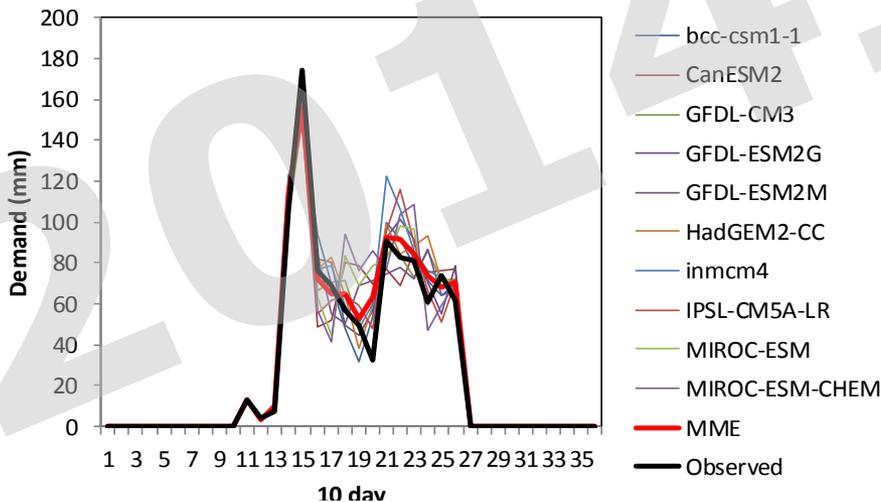
Reproducibility of irrigation demand

Monthly



Models	Demand (mm)	% difference	R ²
Observed	1038		
MME	1094	5.4	0.989
KMA-12.5km	1057	1.9	0.997
bcc-csm1-1	1066	2.8	0.997
CanESM2	1082	4.2	0.975
GFDL-CM3	1077	3.8	0.986
GFDL-ESM2G	1090	5.0	0.987
GFDL-ESM2M	1073	3.4	0.986
HadGEM2-CC	1135	9.4	0.994
inmcm4	1146	10.5	0.976
IPSL-CM5A-LR	1074	3.4	0.976
MIROC-ESM	1116	7.5	0.977
MIROC-ESM-CHEM	1119	7.9	0.97

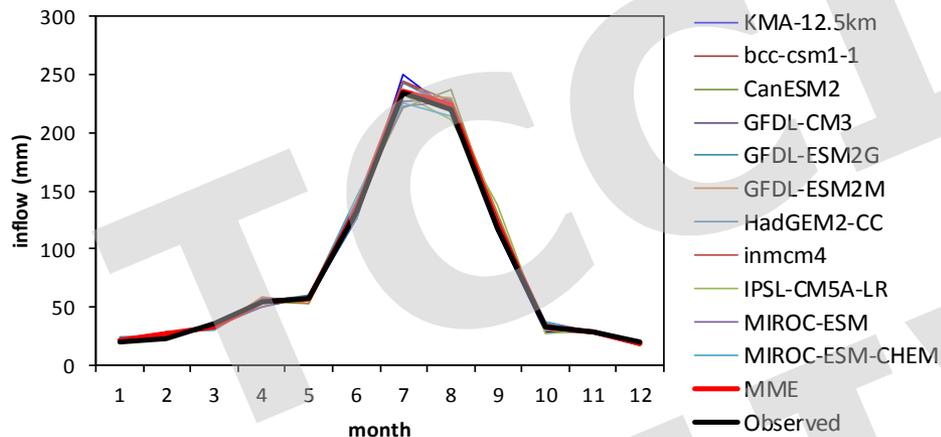
10-day



Models	demand (mm)	% difference	R ²
Observed	1038		
MME	1094	5.4	0.972
KMA-12.5km	1057	1.9	0.962
bcc-csm1-1	1066	2.8	0.961
CanESM2	1082	4.2	0.916
GFDL-CM3	1077	3.8	0.967
GFDL-ESM2G	1090	5.0	0.964
GFDL-ESM2M	1073	3.4	0.939
HadGEM2-CC	1135	9.4	0.967
inmcm4	1146	10.5	0.967
IPSL-CM5A-LR	1074	3.4	0.957
MIROC-ESM	1116	7.5	0.931
MIROC-ESM-CHEM	1119	7.9	0.887

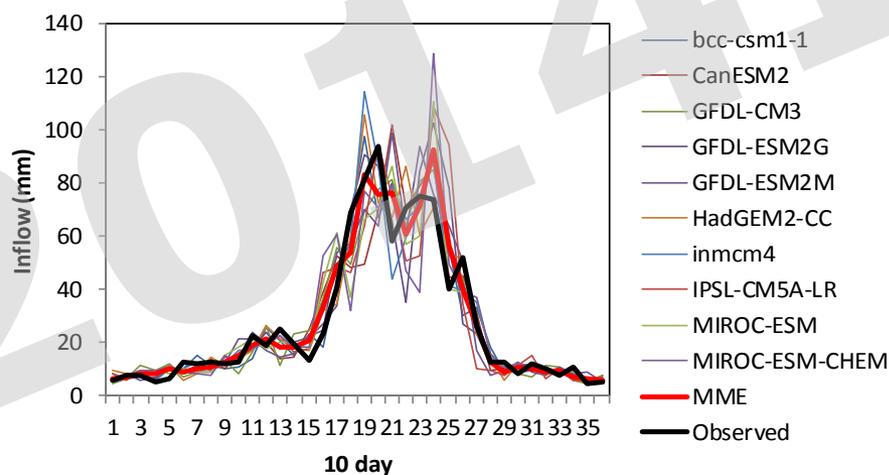
Reproducibility of upstream inflow

Monthly



Models	Inflow (mm)	% difference	R ²
Observed	976		
MME	988	1.3	0.999
KMA-12.5km	980	0.5	0.996
bcc-csm1-1	989	1.4	0.998
CanESM2	982	0.7	0.992
GFDL-CM3	989	1.4	0.996
GFDL-ESM2G	990	1.5	0.998
GFDL-ESM2M	997	2.2	0.996
HadGEM2-CC	987	1.1	0.997
inmcm4	997	2.2	0.998
IPSL-CM5A-LR	992	1.7	0.991
MIROC-ESM	984	0.9	0.996
MIROC-ESM-CHEM	989	1.4	0.995

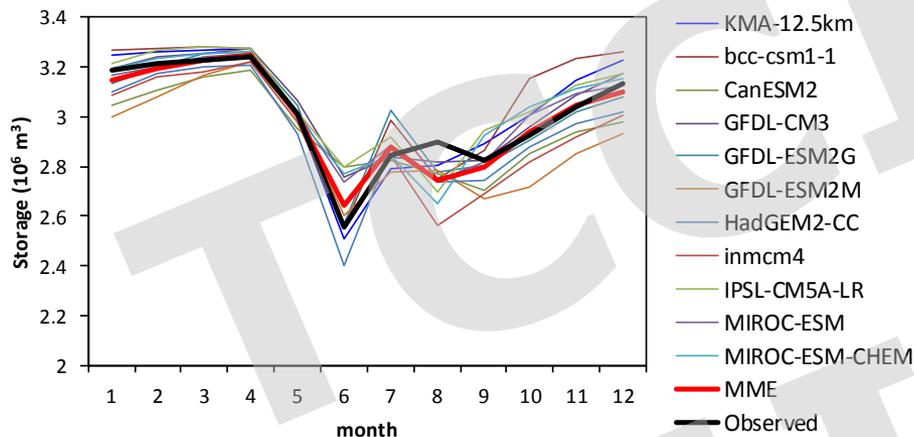
10-day



Models	Inflow (mm)	% difference	R ²
Observed	976		
MME	988	1.3	0.913
KMA-12.5km	980	0.5	0.891
bcc-csm1-1	989	1.4	0.887
CanESM2	982	0.7	0.804
GFDL-CM3	989	1.4	0.881
GFDL-ESM2G	990	1.5	0.836
GFDL-ESM2M	997	2.2	0.823
HadGEM2-CC	987	1.1	0.911
inmcm4	997	2.2	0.906
IPSL-CM5A-LR	992	1.7	0.739
MIROC-ESM	984	0.9	0.779
MIROC-ESM-CHEM	989	1.4	0.678

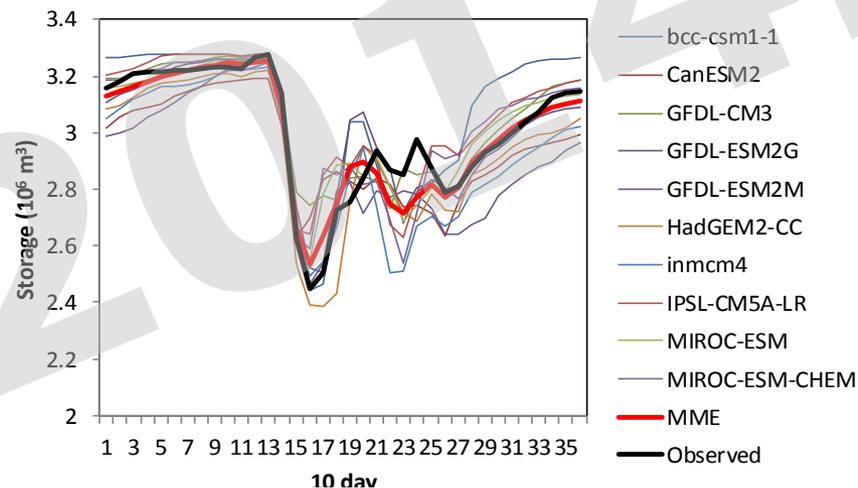
Reproducibility of reservoir storage level

Monthly



Models	Storage (10^6 m^3)	% difference	R ²
Observed	3.009		
MME	2.997	-0.4	0.93
KMA-12.5km	3.036	0.9	0.95
bcc-csm1-1	3.077	2.3	0.84
CanESM2	2.943	-2.2	0.76
GFDL-CM3	3.041	1.1	0.91
GFDL-ESM2G	3.006	-0.1	0.89
GFDL-ESM2M	2.901	-3.6	0.84
HadGEM2-CC	2.937	-2.4	0.96
inmcm4	2.920	-3.0	0.84
IPSL-CM5A-LR	3.061	1.7	0.77
MIROC-ESM	3.025	0.5	0.87
MIROC-ESM-CHEM	3.036	0.9	0.74

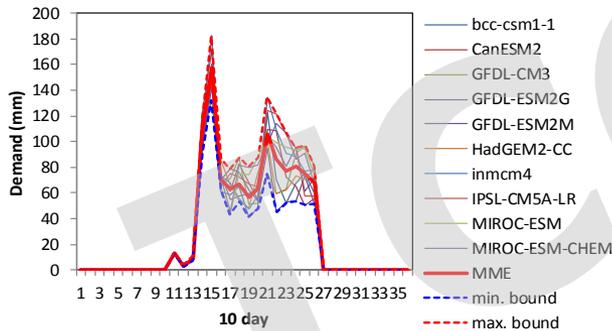
10-day



Models	Storage (10^6 m^3)	% difference	R ²
Observed	108		
MME	108	-0.4	0.922
KMA-12.5km	109	0.9	0.92
bcc-csm1-1	111	2.3	0.818
CanESM2	106	-2.2	0.767
GFDL-CM3	109	1.1	0.871
GFDL-ESM2G	108	-0.1	0.857
GFDL-ESM2M	104	-3.6	0.841
HadGEM2-CC	106	-2.4	0.923
inmcm4	105	-3.0	0.82
IPSL-CM5A-LR	110	1.7	0.789
MIROC-ESM	109	0.5	0.856
MIROC-ESM-CHEM	109	0.9	0.745

Uncertainties in RCP8.5 scenario (2011~2040)

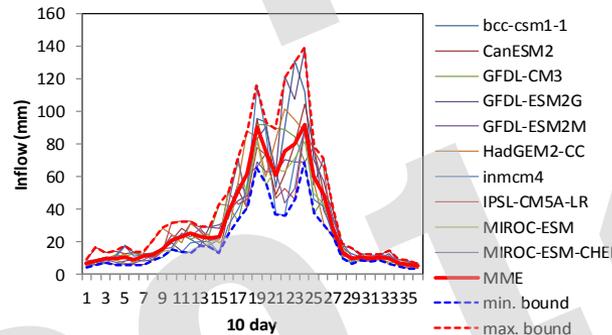
Demand



Models	Demand (mm)	% Change
MME	1099	
bcc-csm1-1	1026	-6.7
CanESM2	1117	1.6
GFDL-CM3	1051	-4.4
GFDL-ESM2G	979	-11.0
GFDL-ESM2M	1079	-1.9
HadGEM2-CC	1062	-3.4
inmcm4	1195	8.7
IPSL-CM5A-LR	1196	8.7
MIROC-ESM	1197	8.9
MIROC-ESM-CHEM	1210	10.0
KMA-12.5km	997	-9.3

-11.0% ~ 10.0%

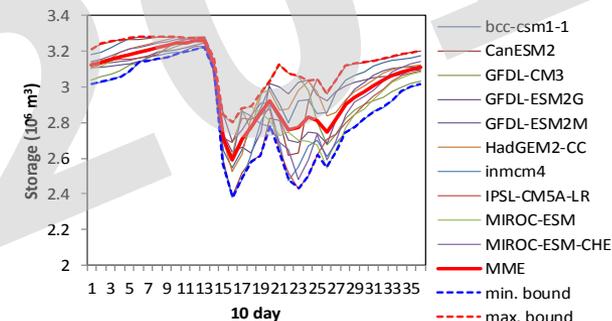
Inflow



Models	Inflow (mm)	% Change
MME	1078	
bcc-csm1-1	1192	10.6
CanESM2	1088	0.9
GFDL-CM3	1083	0.4
GFDL-ESM2G	1198	11.1
GFDL-ESM2M	1052	-2.5
HadGEM2-CC	1125	4.3
inmcm4	954	-11.5
IPSL-CM5A-LR	1009	-6.5
MIROC-ESM	1003	-6.9
MIROC-ESM-CHEM	993	-7.9
KMA-12.5km	1153	7.0

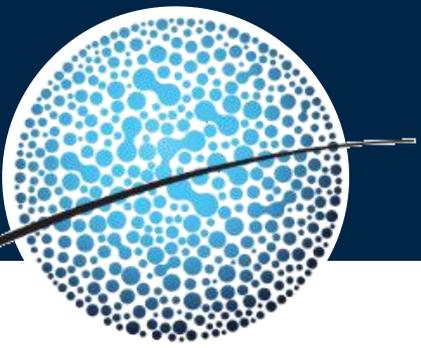
-11.5% ~ 11.0%

Storage

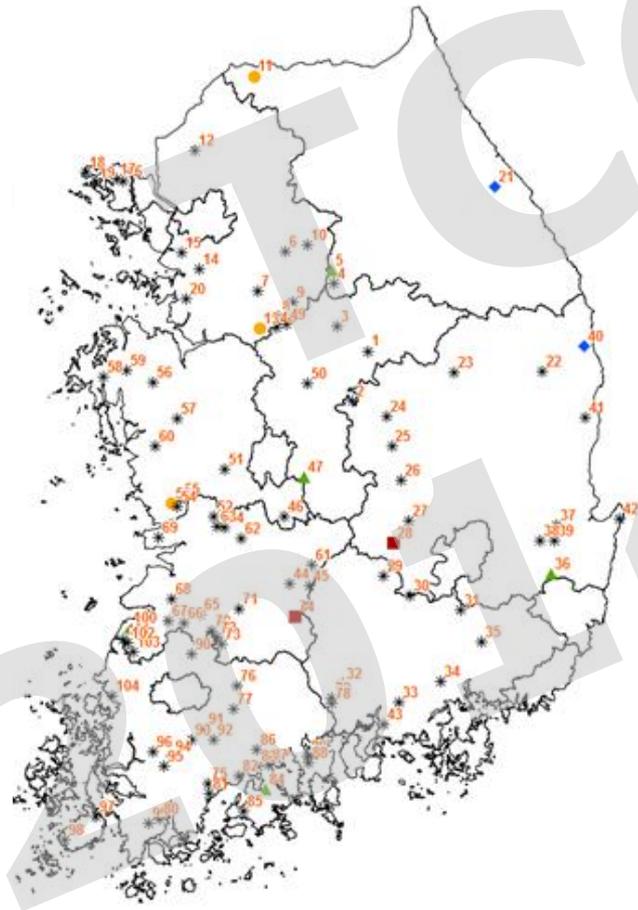


Models	Storage (10 ⁶ m ³)	% Change
MME	3.003	
bcc-csm1-1	3.062	2.0
CanESM2	3.042	1.3
GFDL-CM3	3.011	0.3
GFDL-ESM2G	3.027	0.8
GFDL-ESM2M	3.004	0.0
HadGEM2-CC	3.045	1.4
inmcm4	2.887	-3.9
IPSL-CM5A-LR	2.953	-1.7
MIROC-ESM	2.968	-1.2
MIROC-ESM-CHEM	2.980	-0.8
KMA-12.5km	3.066	2.1

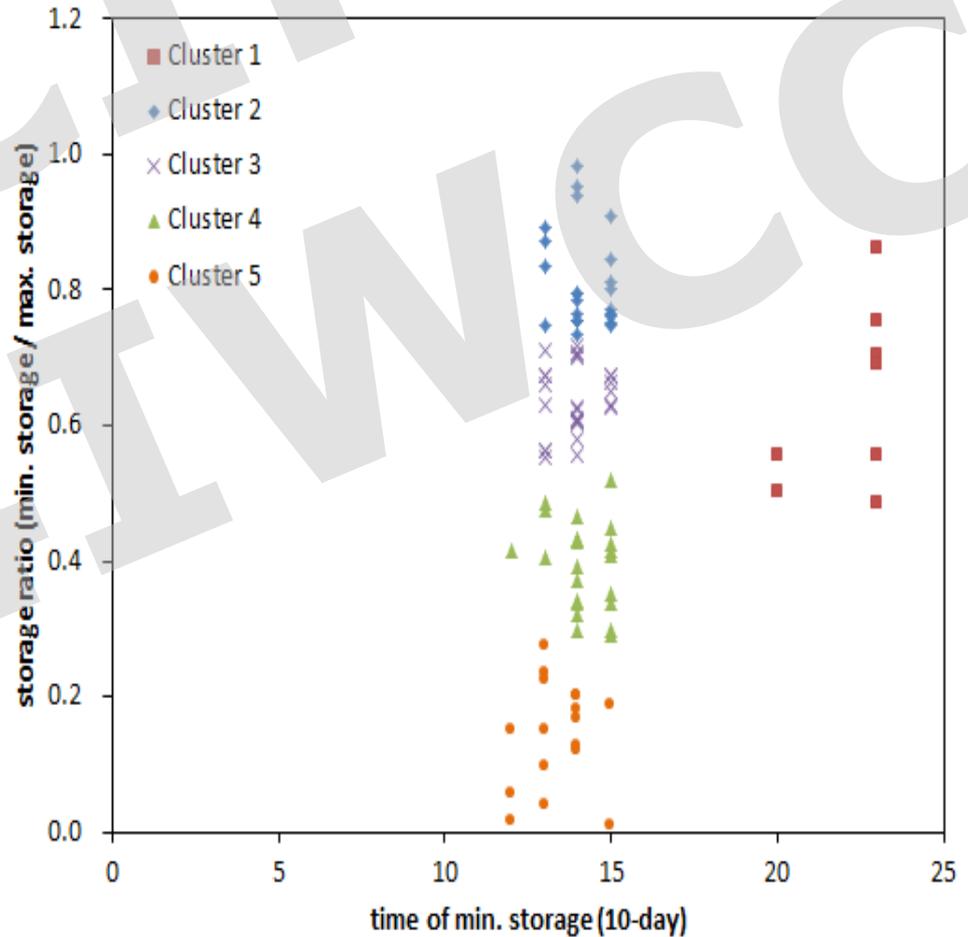
-3.9% ~ 2.1%



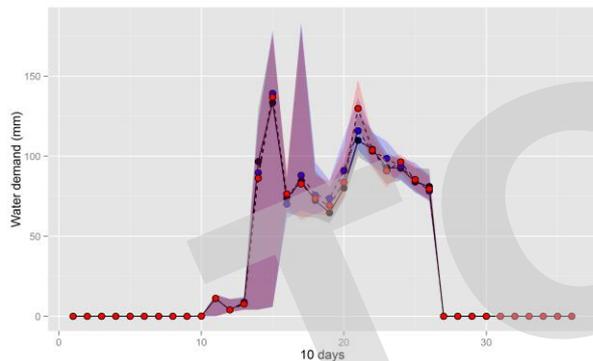
Assessment of Climate Change Impact on Agricultural Reservoirs



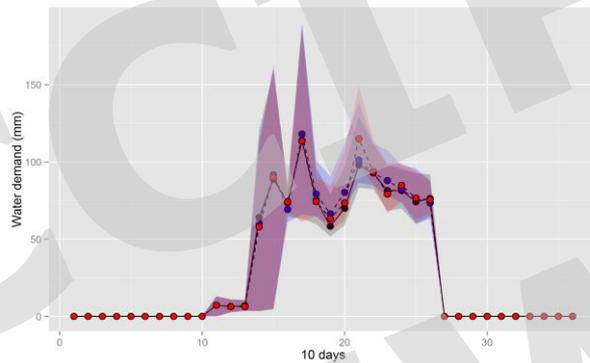
104 reservoirs



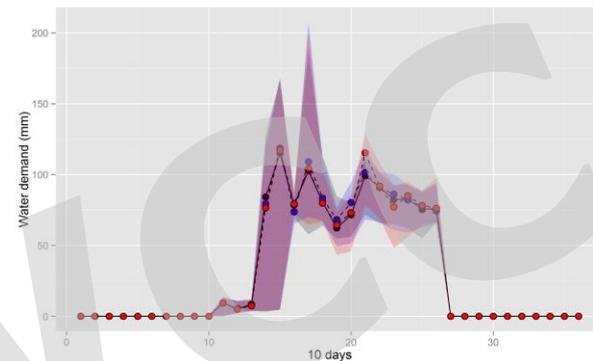
Irrigation demand projection (2011~2040)



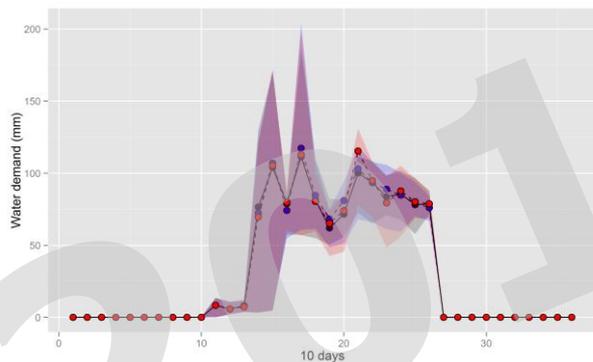
Cluster 1



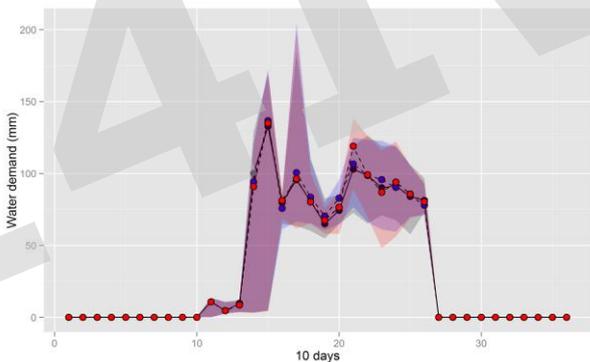
Cluster 2



Cluster 3



Cluster 4

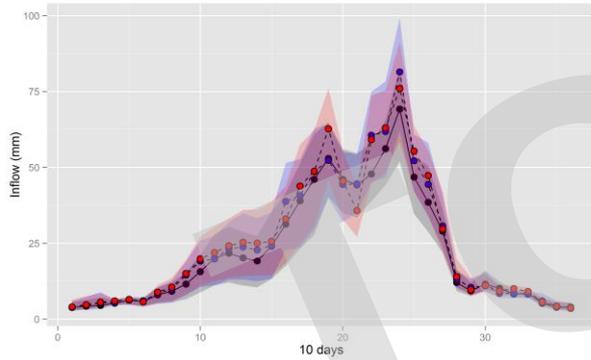


Cluster 5

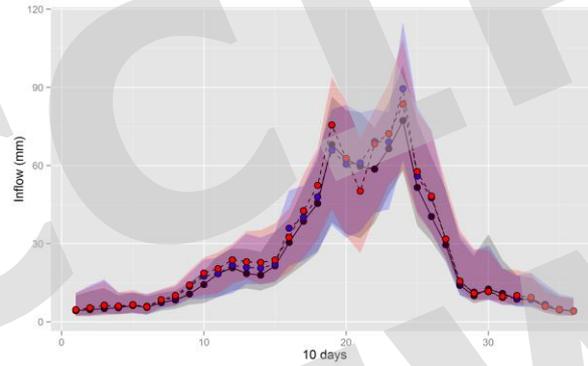
Clusters		Historical	RCP4.5	RCP8.5
1	Inflow(mm)	1194	1226	1218
	% change		2.7	2.0
2	Inflow(mm)	1069	1099	1092
	% change		2.8	2.1
3	Inflow(mm)	1126	1148	1140
	% change		2.0	1.3
4	Inflow(mm)	1128	1153	1145
	% change		2.2	1.5
5	Inflow(mm)	1202	1223	1217
	% change		1.8	1.3

1.3% ~ 2.1%

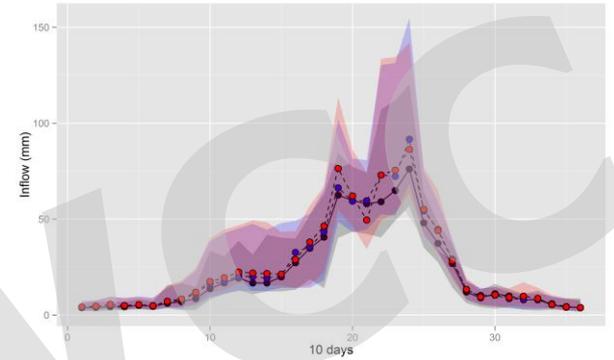
Inflow projection (2011~2040)



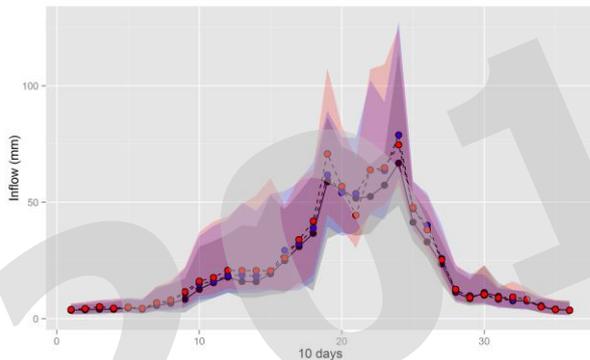
Cluster 1



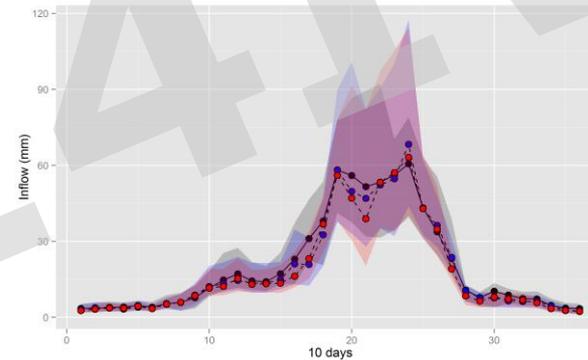
Cluster 2



Cluster 3



Cluster 4

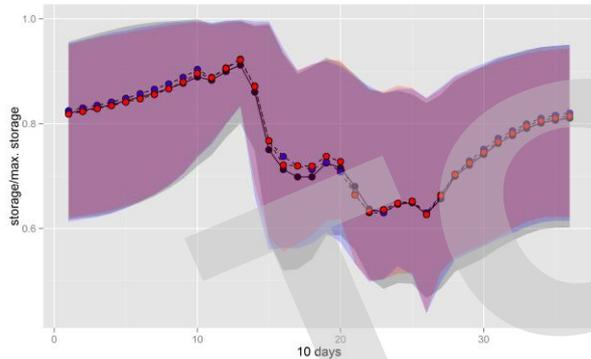


Cluster 5

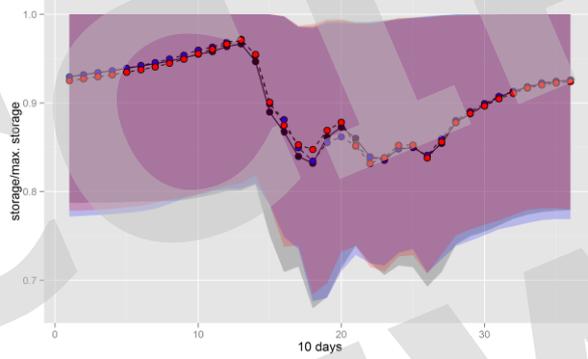
Clusters		Historical	RCP4.5	RCP8.5
1	Inflow(mm)	799	871	885
	% change		9.0	10.8
2	Inflow(mm)	876	940	959
	% change		7.3	9.5
3	Inflow(mm)	817	902	919
	% change		10.4	12.5
4	Inflow(mm)	742	815	829
	% change		9.8	11.7
5	Inflow(mm)	722	796	806
	% change		10.2	11.5

9.5% ~ 12.5%

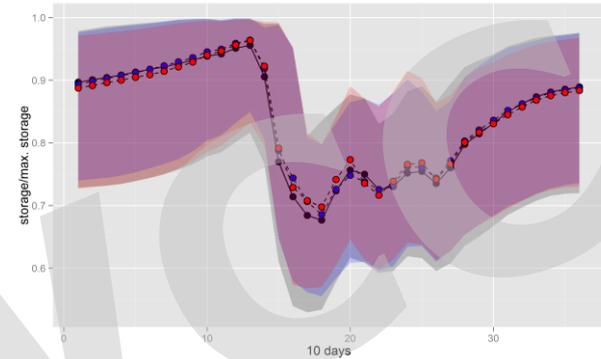
Storage ratio projection (2011~2040) (current storage / total storage)



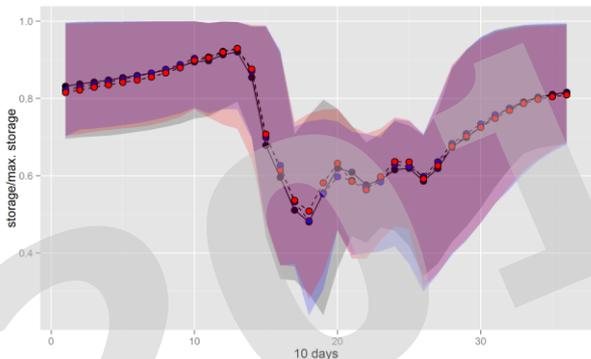
Cluster 1



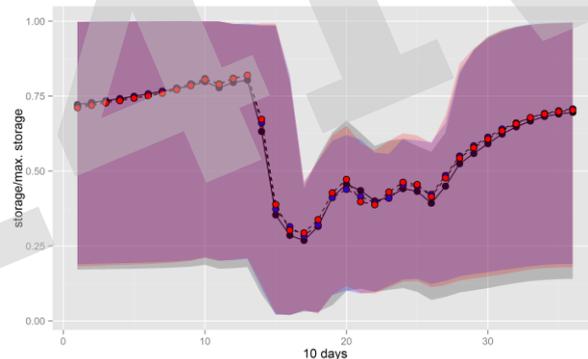
Cluster 2



Cluster 3



Cluster 4



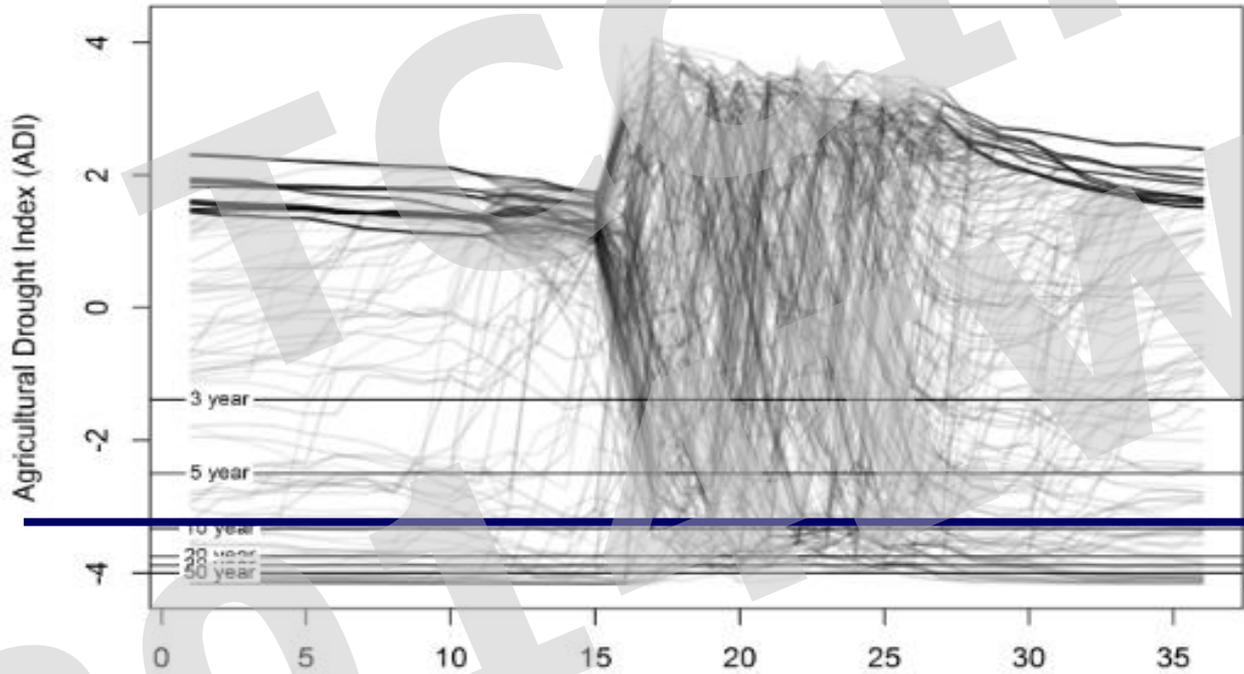
Cluster 5

Clusters		Historical	RCP4.5	RCP8.5
1	Storage ratio	0.77	0.78	0.78
	% change		0.9	0.5
2	Storage ratio	0.90	0.90	0.90
	% change		0.2	0.1
3	Storage ratio	0.84	0.84	0.84
	% change		0.5	0.3
4	Storage ratio	0.74	0.74	0.74
	% change		0.4	0.3
5	Storage ratio	0.59	0.60	0.60
	% change		1.6	1.6

0.1% ~ 1.6%

Agricultural drought index based on storage level

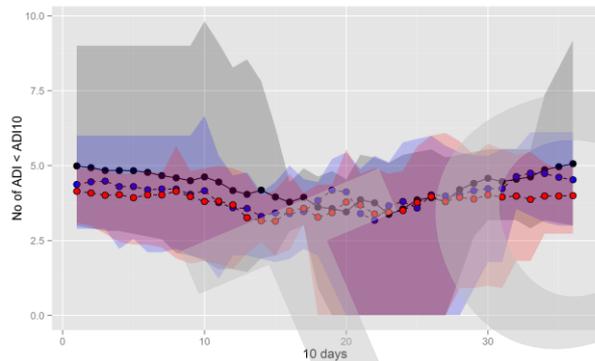
- Storage → Agricultural Drought Index (ADI)



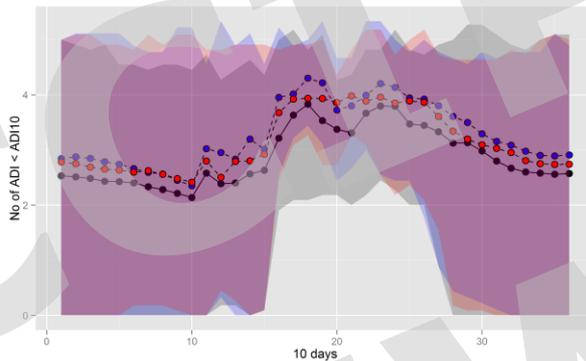
10-year frequency

Frequency(yr)	~3	3	5	10	20	30	50	50~
$P(X \leq x)$		0.33	0.20	0.10	0.05	0.03	0.02	
ADI		-1.39	-2.50	-3.33	-3.75	-3.89	-4.00	
Classification	Normal	Moderate drought	Severe drought		Extreme drought			

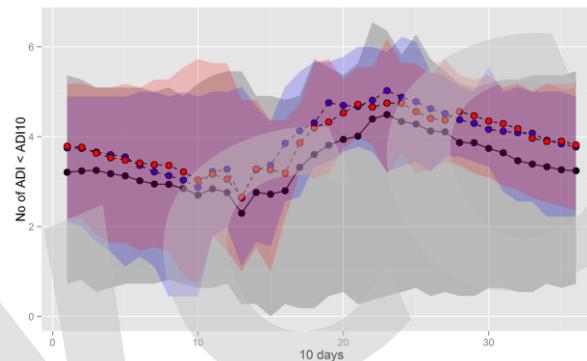
ADI projection (2011~2040)



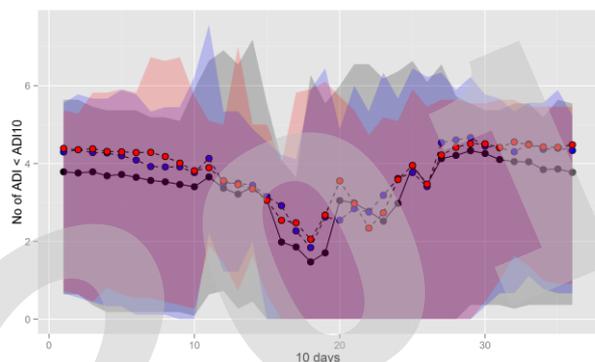
Cluster 1



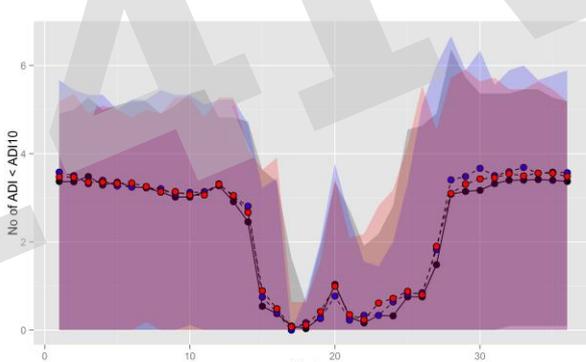
Cluster 2



Cluster 3



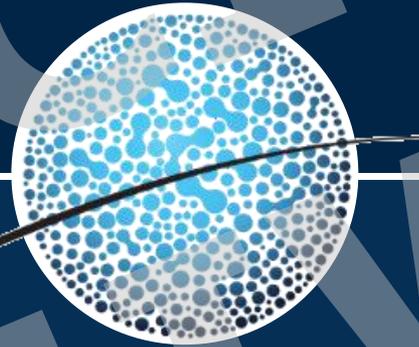
Cluster 4



Cluster 5

Clusters		Historical	RCP4.5	RCP8.5
1	No of ADI < ADI ₁₀	154.6	145.0	136.0
	% change		-6.2	-12.0
2	No of ADI < ADI ₁₀	103.7	118.1	113.0
	% change		13.8	8.9
3	No of ADI < ADI ₁₀	122.9	141.4	139.8
	% change		15.1	13.8
4	No of ADI < ADI ₁₀	122.3	135.0	136.4
	% change		10.4	11.5
5	No of ADI < ADI ₁₀	80.4	84.7	84.4
	% change		5.3	4.9

-12.0% ~ 13.8%



Concluding Remarks

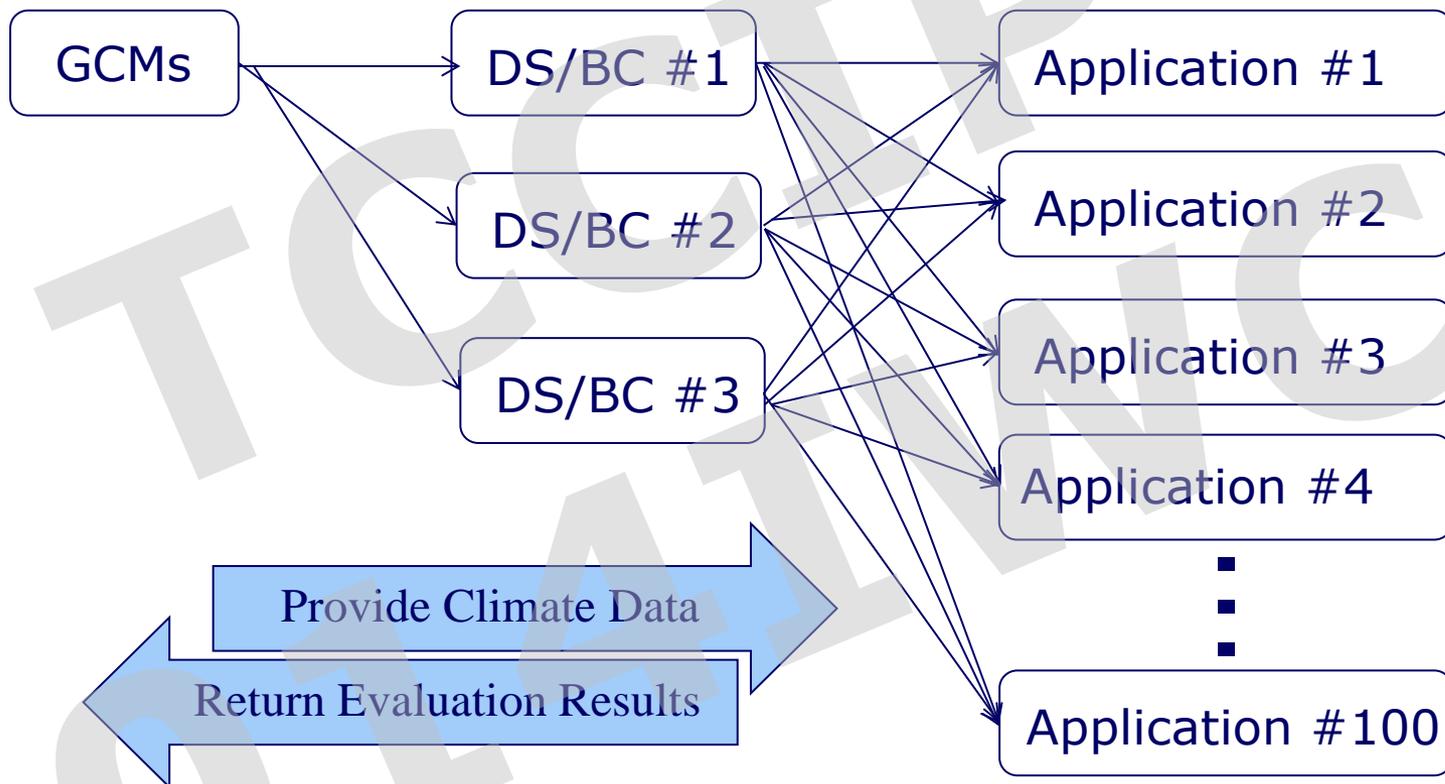


Within a climate change impact assessment

- Upstream Inflow, irrigation demand, and water storage level increased for the future period.
- However, frequency of agricultural drought based on the reservoir storage level showed the increase trends.
- Precipitation shows the higher uncertainty compared to the other weather variables.
- Reproducibility-related information for the downscaled/bias-corrected scenario data should be provided.
- Multi-model ensemble (MME) is necessary in order to consider uncertainty.



- **Process-based evaluation of GCMs and DS methods is necessary**



- **Concerns in Downscaling**

- Spatial/variable coherence, Hourly data, Grid data

