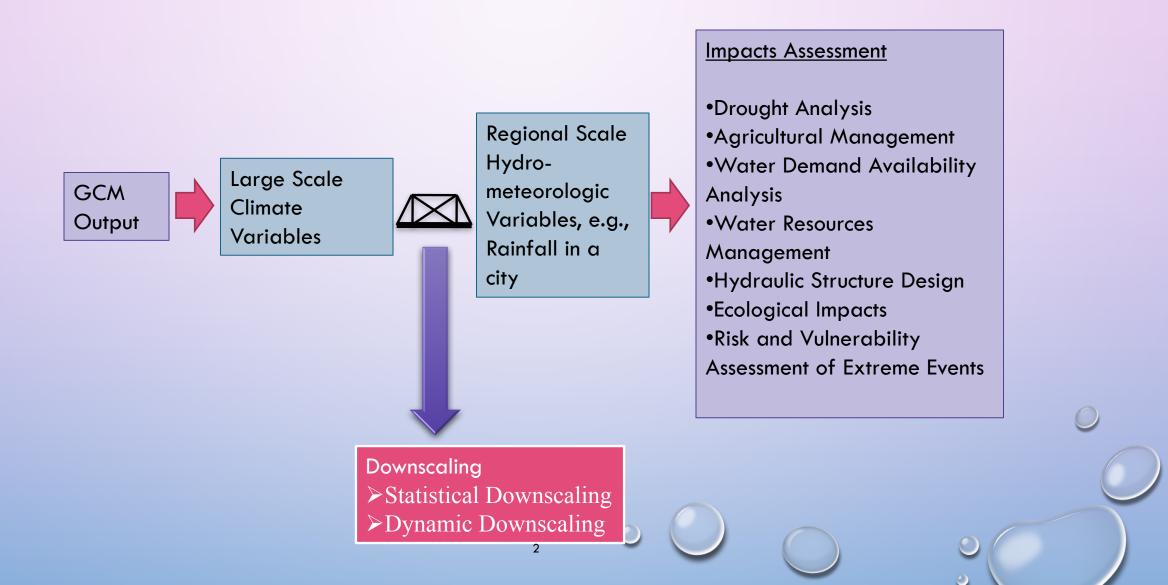
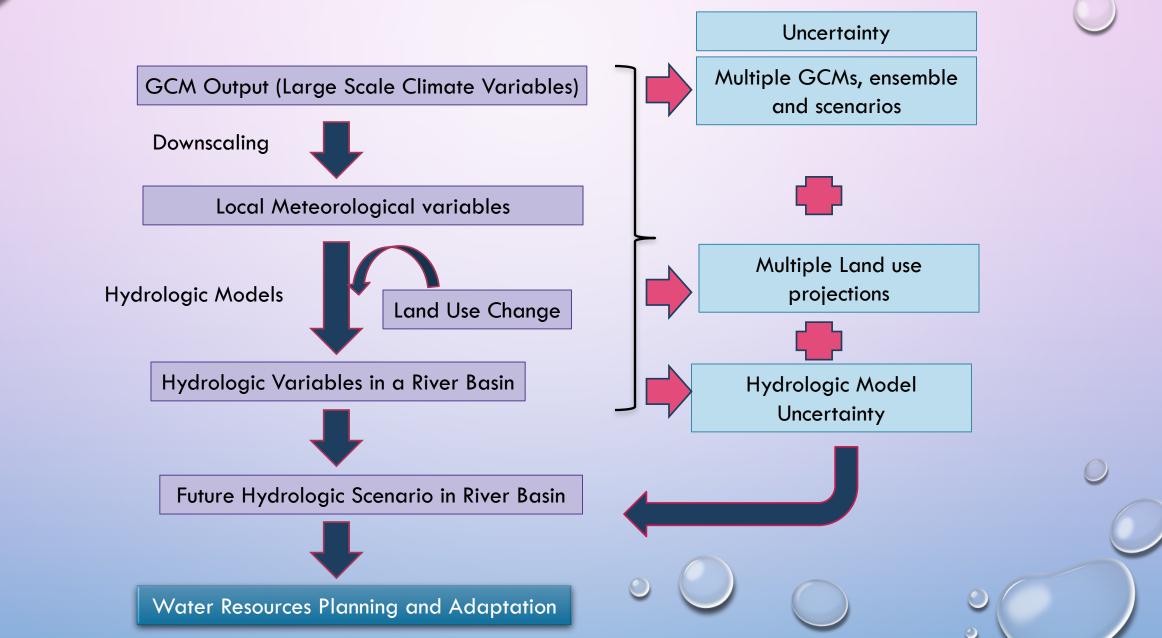
### Applications of Downscaling in Assessing Impacts of Climate Change: Challenges and Unanswered Questions

Subimal Ghosh Department of Civil Engg. and IDP in Climate Studies IIT Bombay

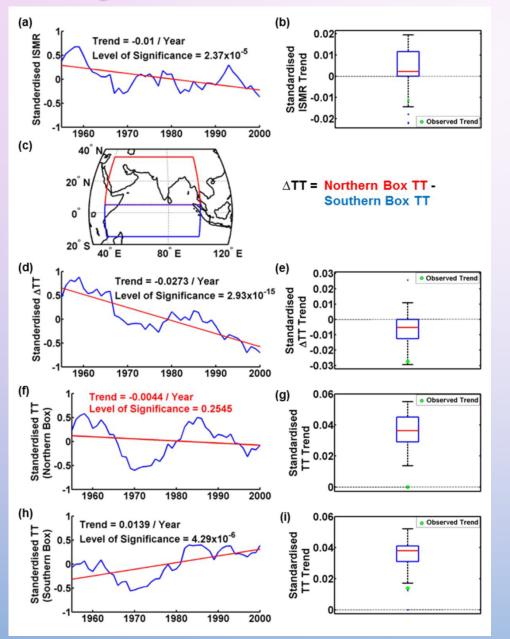
### Downscaling



# General Framework



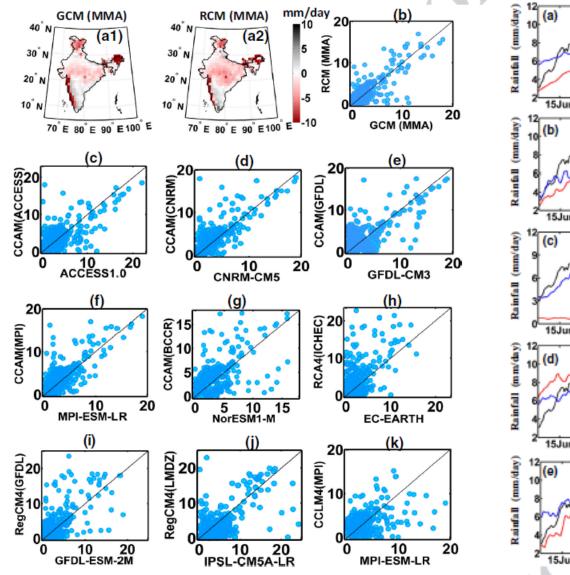
### Can GCMs simulate observed trend?

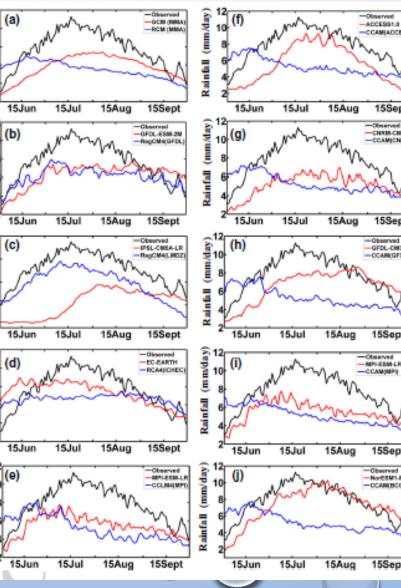


Even if you consider Multi-Model Average → it will give opposite trend

Saha et al. (2014), GRL

# CORDEX Models are great but may need Improvements





#### Singh et al. (2016) **Climate Dynamics**

COAMLACCESS

15Sept

-Observed

CNRM-CMS COMPONENT

15Sept

-Observed

GFDL-CIMD

OCANVGED!

15Sept

MPI-ESM-LR

15Sept

Observed

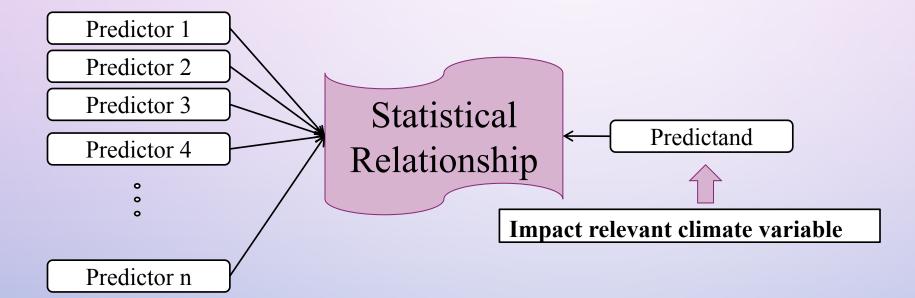
NorESM1-III

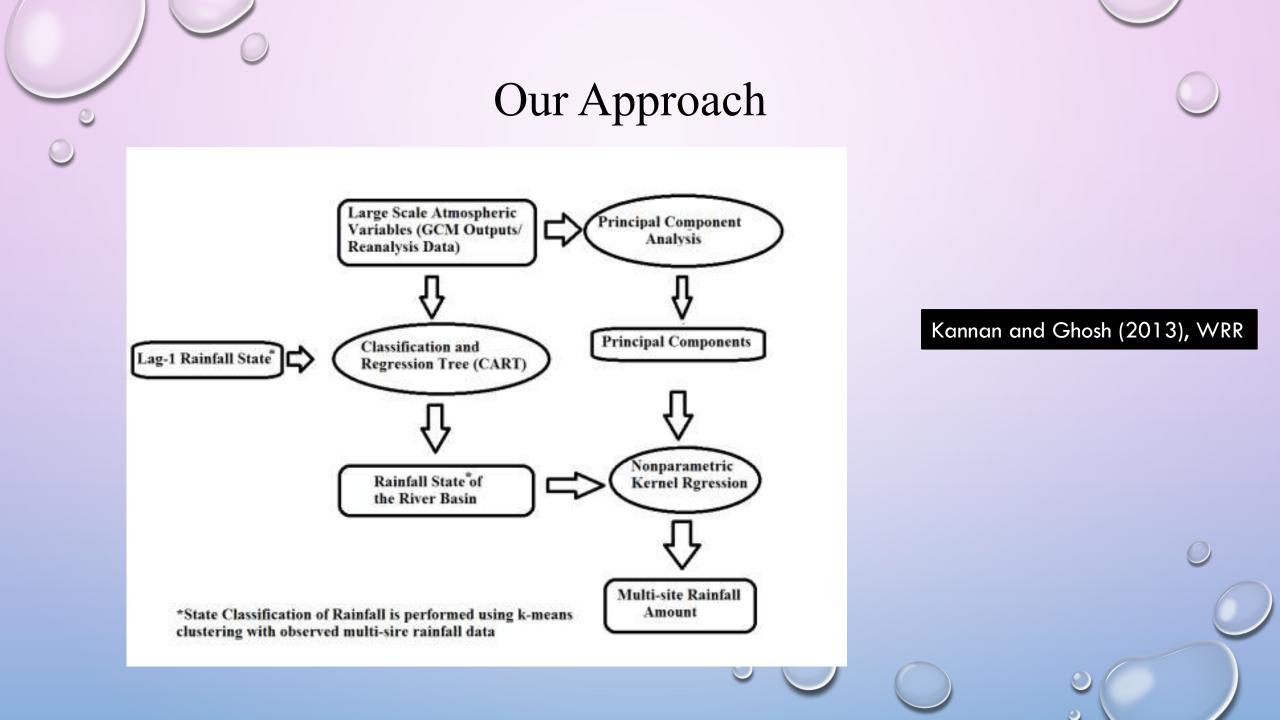
CCAMBCCS.

15Sept

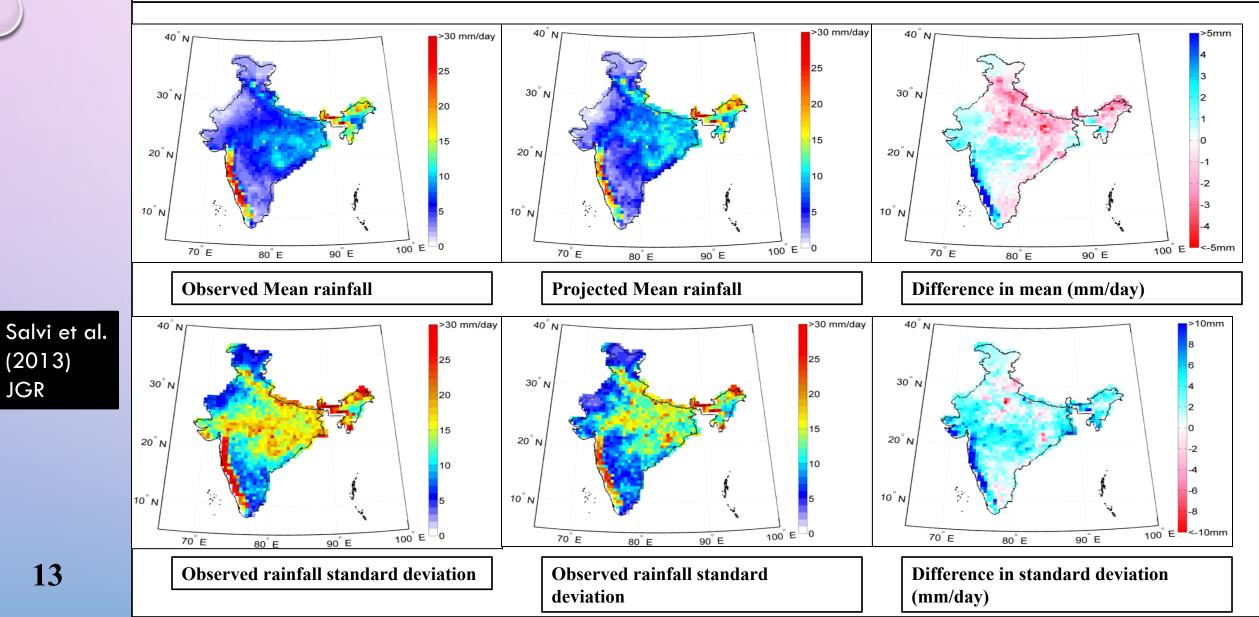
### What about statistical downscaling?

### Standard Approach

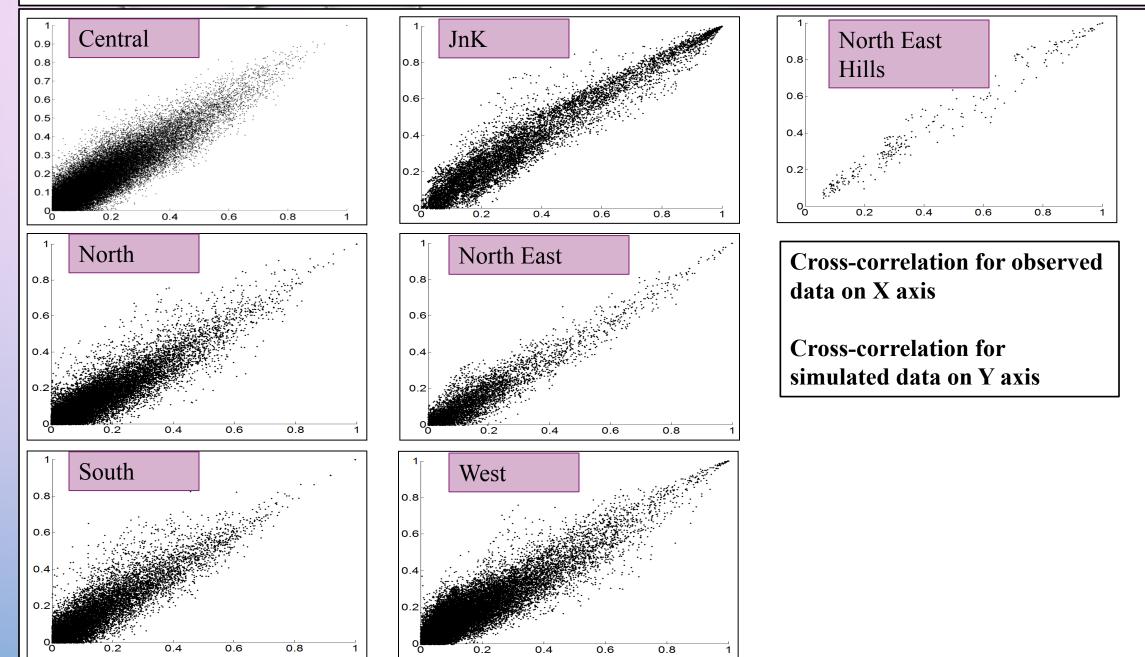




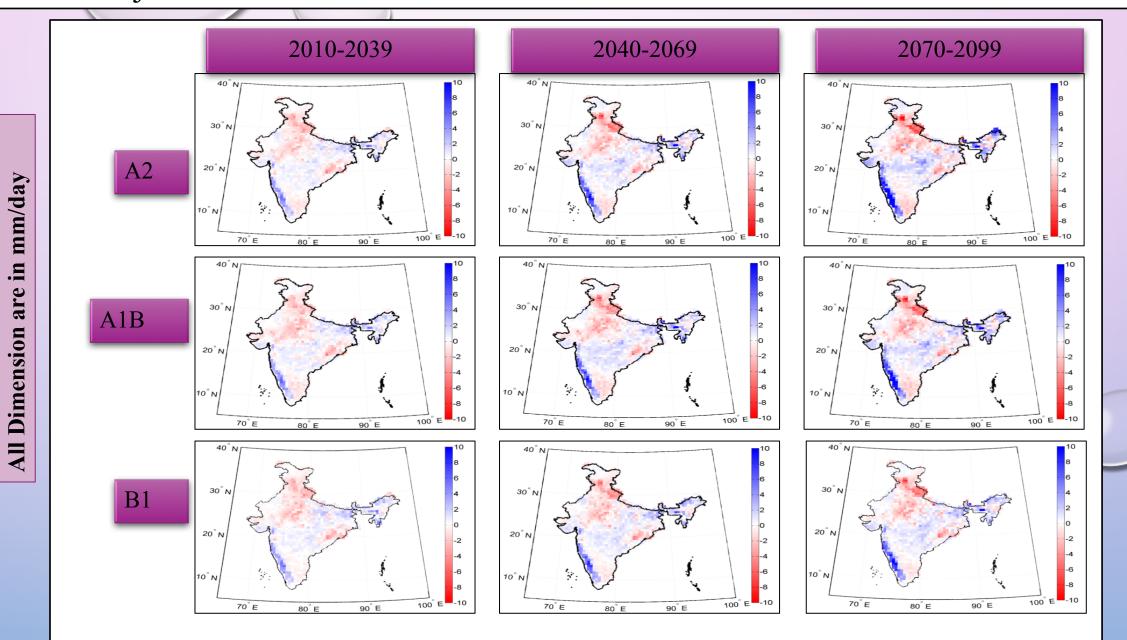
# Historic Period Simulations: Comparison of Statistical Properties (1971-2000)



### Zone-Wise Cross-correlation plots



### **Future Projections**



### Do they work in non-stationary climate?

#### **Design of Experiments**

**Experiment Series 1** Basis-Criteria based training (**TR**) and testing (**TE**) period selection

**Base Experiment (TR-RAN-TE-RAN):** Rainfall projections obtained using randomly selected training and testing period from 1951-2005.

#### **Criteria based Selection:**

Criteria 1: Conventional way

(1) Train with first 30 years (chronological) and test for remaining **(TR-CH-TE-CH)**.

Criteria 2: Hypothetical Climate change scenario

 Train with relatively cold (cooler) years and test for relatively hot (warmer) years (TR-C-TE-H) from 1951-2005.

(2) Train with non-ENSO years and test for ENSO years (TR-nonEN-TE-EN) from 1951-2005.

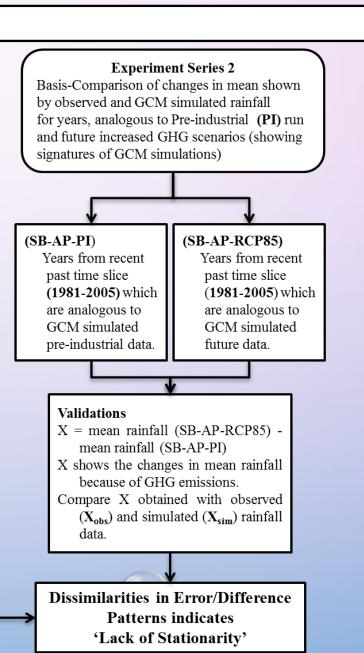
Criteria 3: Reverse Climate Change scenario

 Train with relatively hot (warmer) years and test for relatively cold (cooler) years (TR-H-TE-C) from 1951-2005.

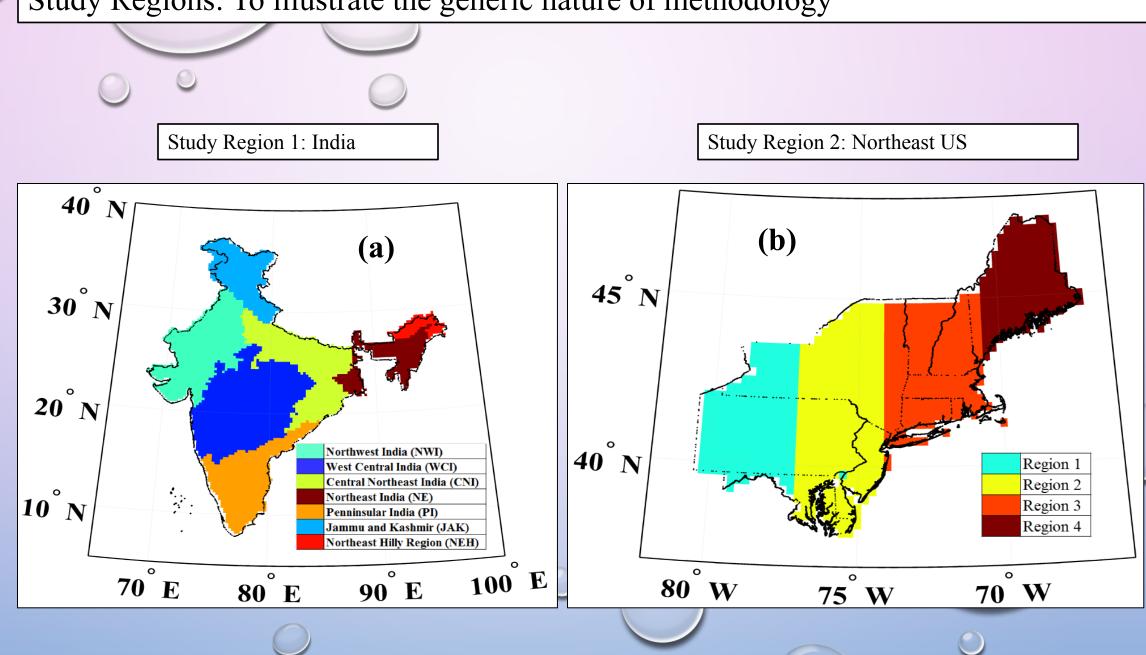
(2) Train with ENSO years and test for non-ENSO years (TR-EN-TE-nonEN) from 1951-2005.

#### Validations

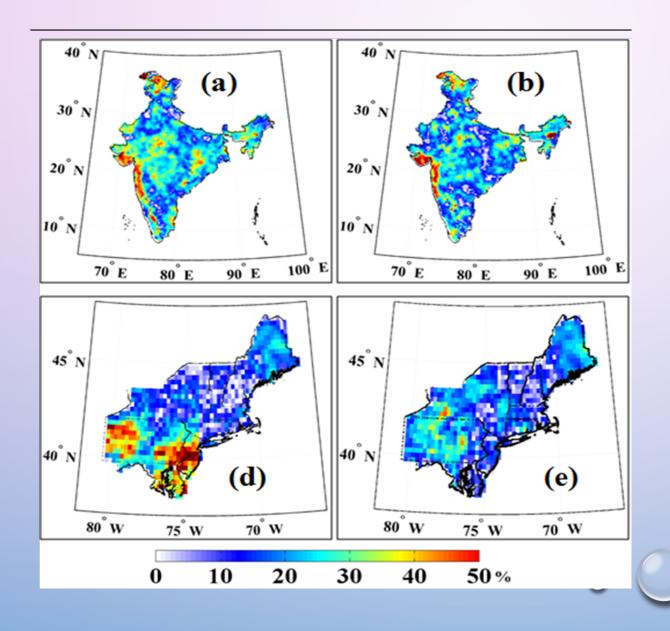
Magnitude and spatial pattern of Root Mean Square Error (RMSE) for **TR-RAN-TE-RAN** is compared with that for the set of experiments from series 1.



Study Regions: To illustrate the generic nature of methodology



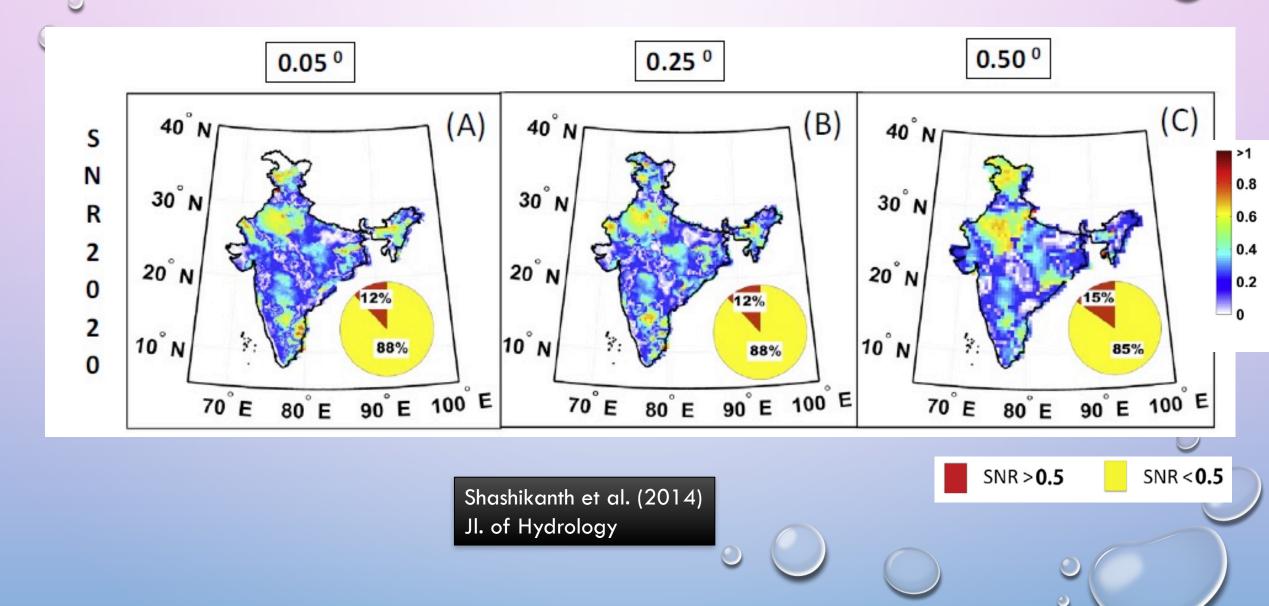
### Violation on Non-stationarity



### Salvi et al., 2016, Clim Dyn

 $\bigcirc$ 

# Do they provide any signal of changes?



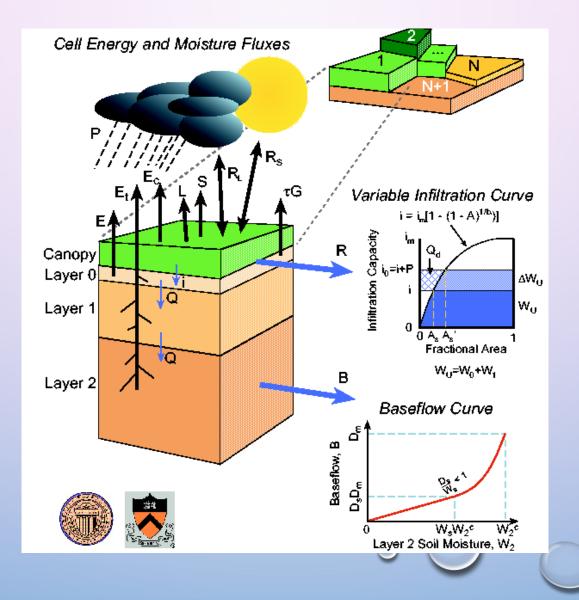




# Assessing Hydrological Impacts



### Variable Infiltration Capacity Model



### Hydrological Parameterization in the background of Climate Uncertainty

-20

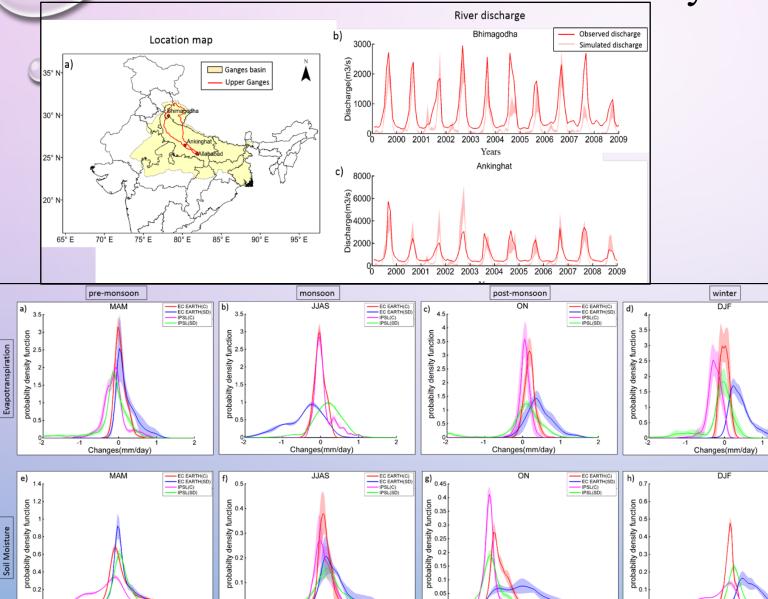
-15 -10 -5 0 5 10

Changes(mm/day)

EC EARTH(C) EC EARTH(SC IPSL(C) IPSL(SD)

EC EARTH(C) EC EARTH(SI

IPSL(C) IPSL(SD)



-15

-10

0 5 10

Changes(mm/day)

0

Changes(mm/day)

-10

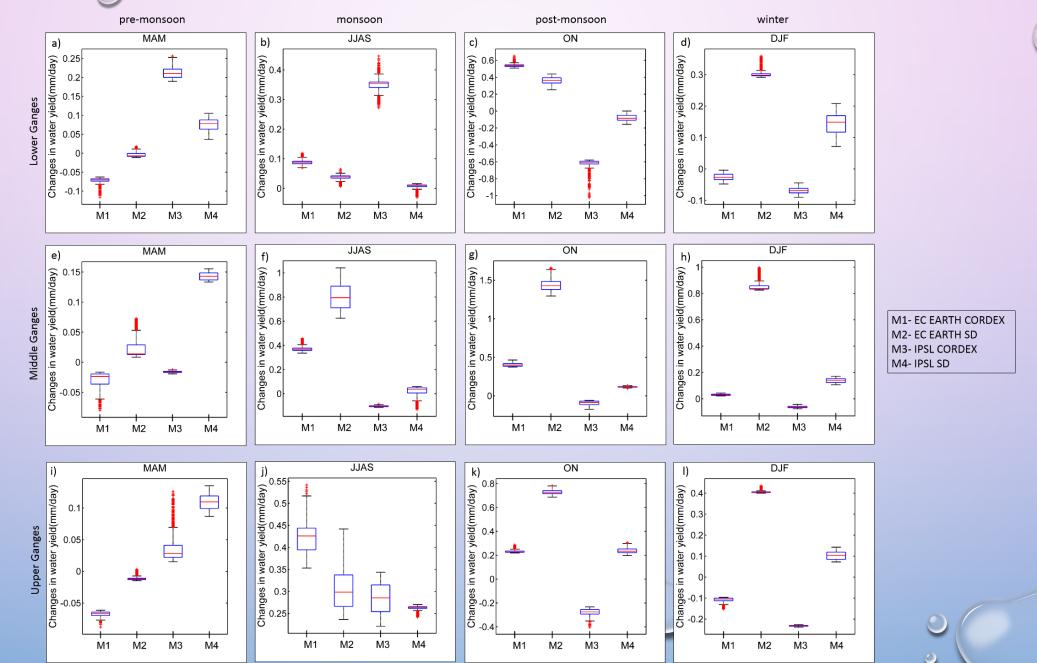
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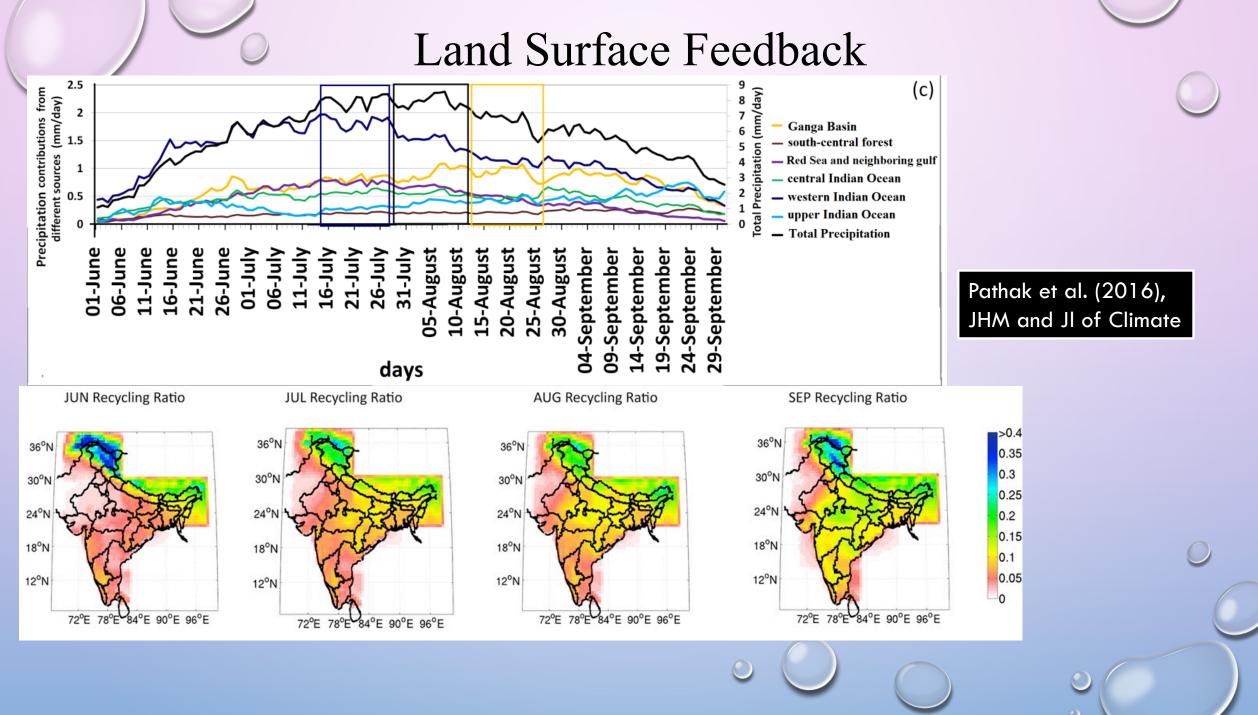
Changes(mm/day)

Given the climate uncertainty, you may not even require any hydrological parameterization?

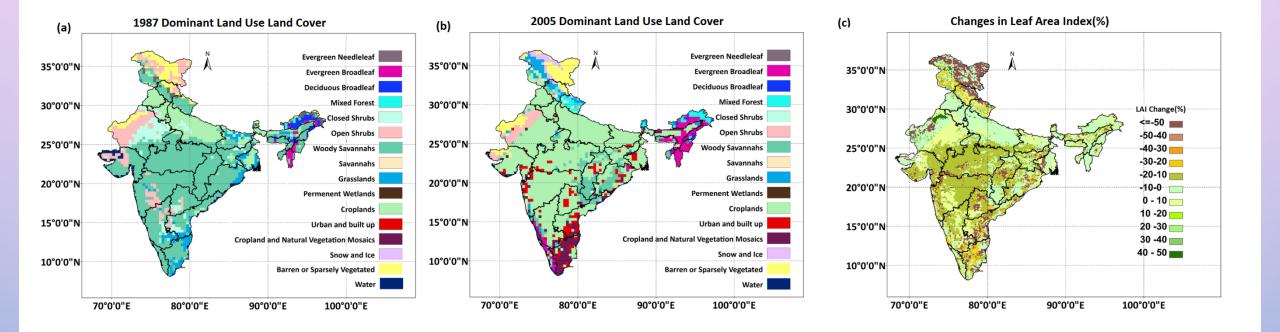
> Joseph et al. (2016), Revised manuscript under preparation for JH

# Hydrological Parameter Calibration lost its value? ③

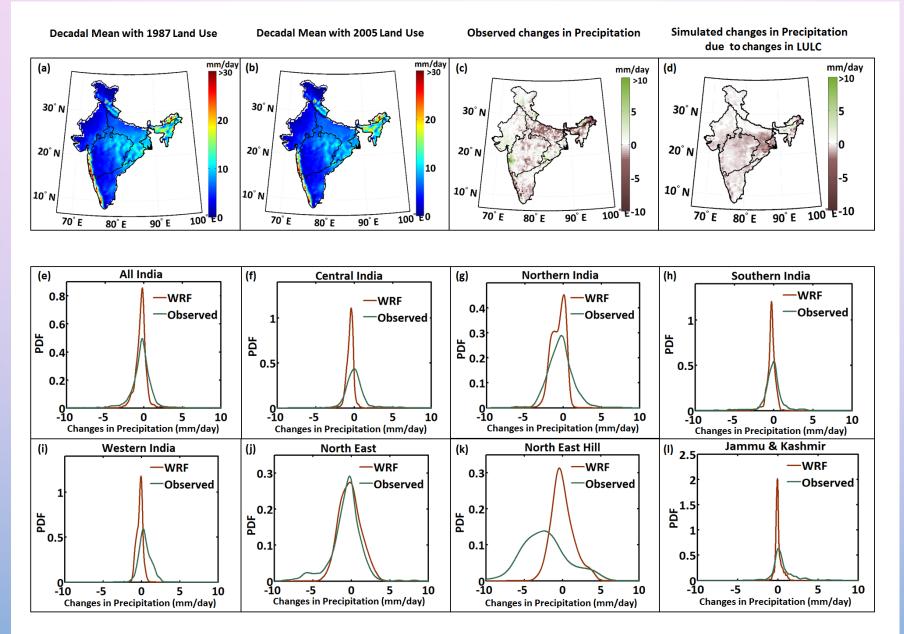




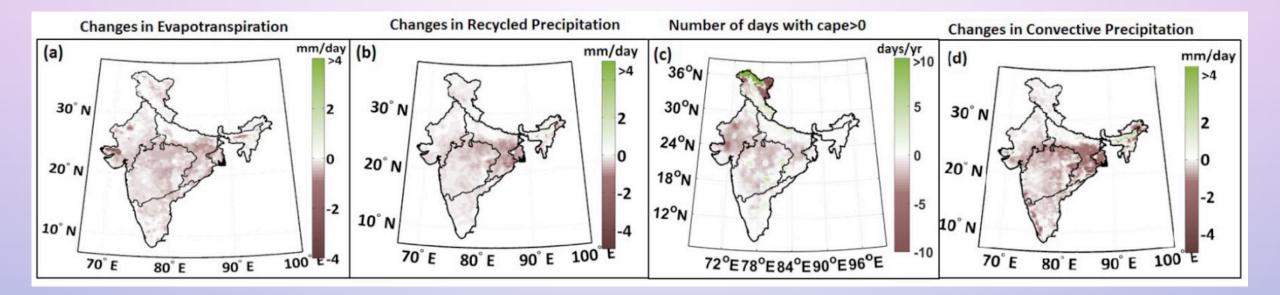
### LULC Changes in India



### Impacts of LULC: Changes Partially Consistent with Observed Changes



### Processes



Paul et al. (2016), Scientific Reports (Nature Publishing Group)

# **Open Questions**

- We use the word "Uncertainty", but are we really "Confident" enough in providing regional projections with different changing sign and magnitudes to water resources managers for developing adaptation strategies?
- Shouldn't we consider feedback from regional LULC changes and growing urbanization?
- We are worried about 'stationarity' in statistical downscaling relationship; but with a high bias in Dynamic downscaling, we have to use bias correction... are regional biases stationary?





# Thank you

