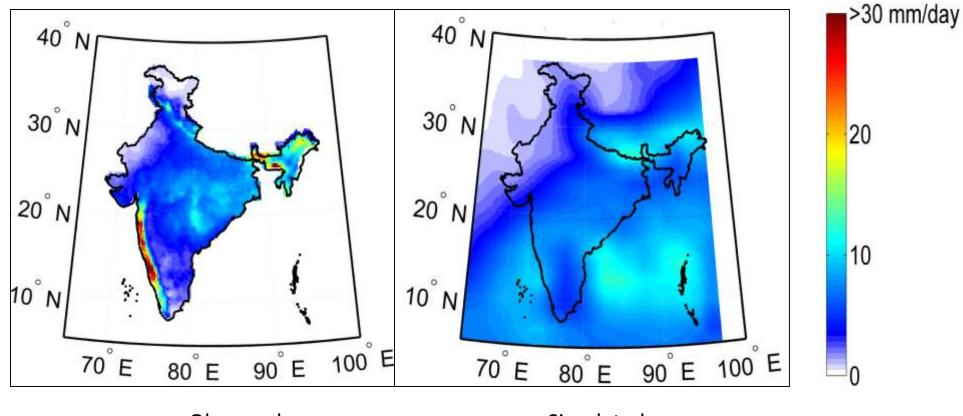
Statistical Downscaling for hydro-climatic projections with CMIP5 Simulations to assess Impact of Climate Change

IIT Bombay, IISc Bangalore, IIT Gandhinagar, IIT Kanpur and IIT Guwahati

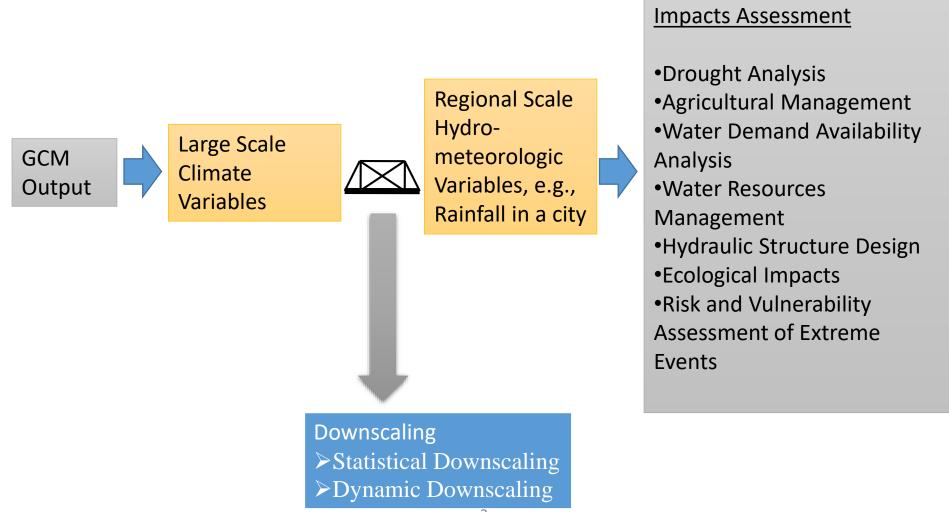
#### GCM Simulations for Indian Monsoon Rainfall



Observed

Simulated

# Downscaling



# Statistical Downscaling

- Data Driven Approach
- Computationally in-expensive
- Based on relationship between large scale climate variables and local scale desired variables
- Basic Assumption: this statistical relationship will not change in altered condition

# Statistical Downscaling

- Philosophy:
  - GCMs can not simulate rainfall very well as rainfall is a regional scale phenomena
  - But GCMs simulate well some of the large scale variables, which affects rainfall.
  - Those large scale variables: Predictor
  - Rainfall: Predictand
  - Derive and apply the relationship between predictor and predictand

# Proposed Methodology

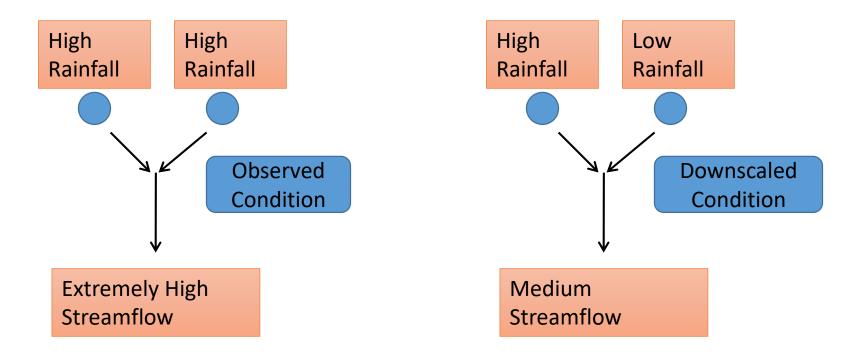
- Kernel Regression Based Statistical Downscaling (IIT Bombay)
- ANN based Statistical Downscaling (IITK and IITG)
- Bias Correction Spatial Disaggregation (IITGN)
- Uncertainty Modeling (IISc Bangalore)

# Steps in Statistical Downscaling

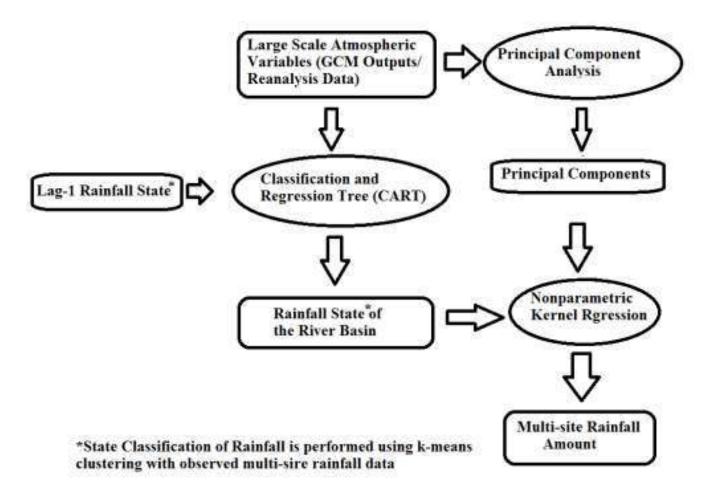
- Deriving relationship between predictor and predictand
- Apply the relationship to the GCM simulated predictors to project predictand

# Multi-site Daily Downscaling

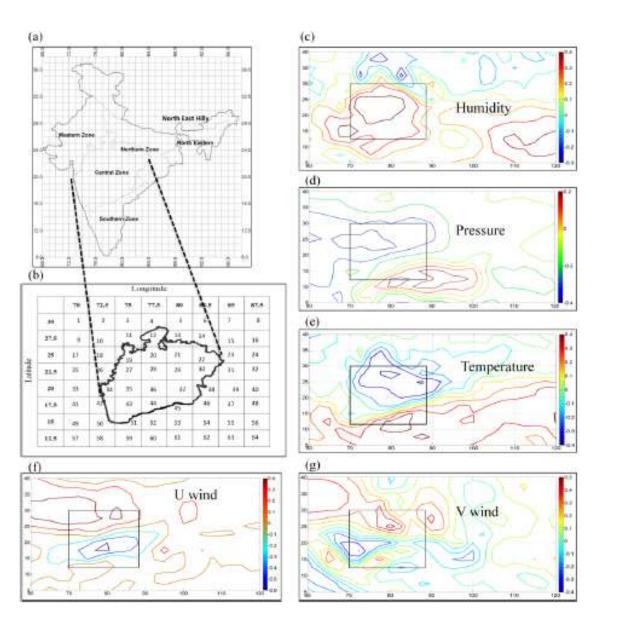
- Challenges
  - Cross correlation
  - Variability for daily scale



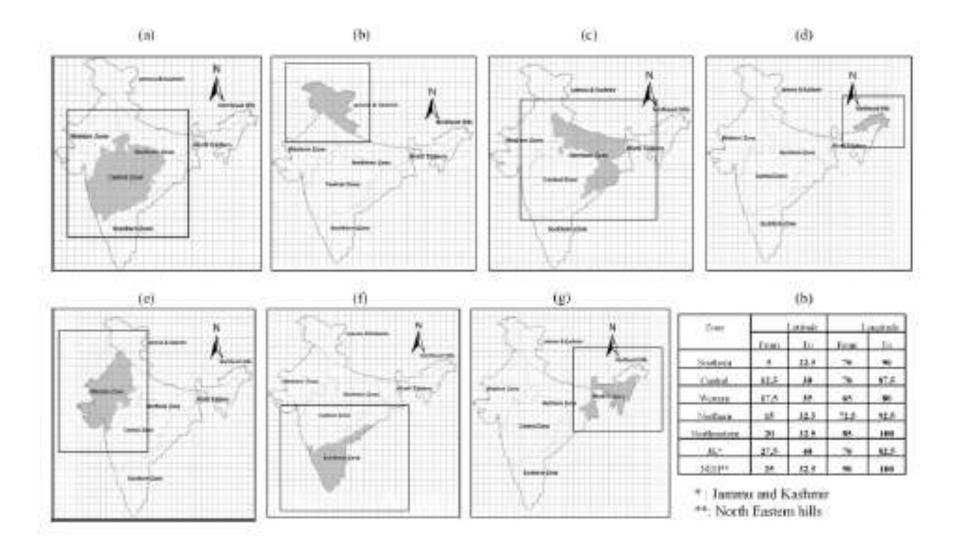
### Kernel Regression Based Approach



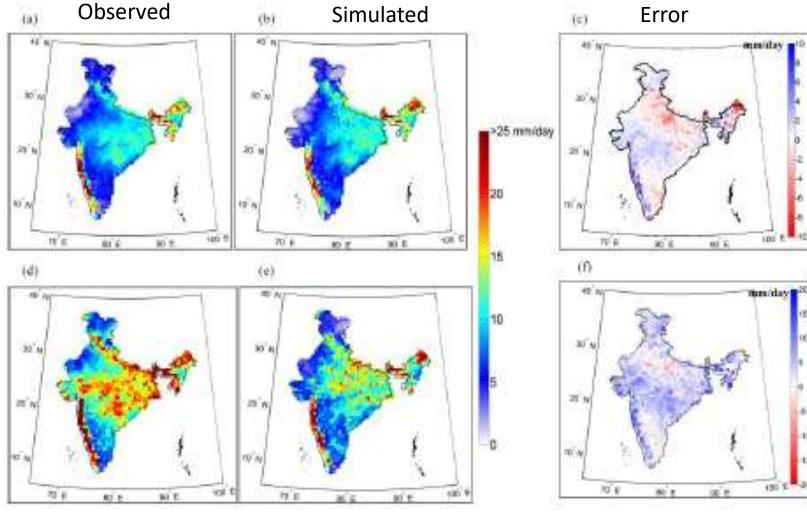
### Application to All India



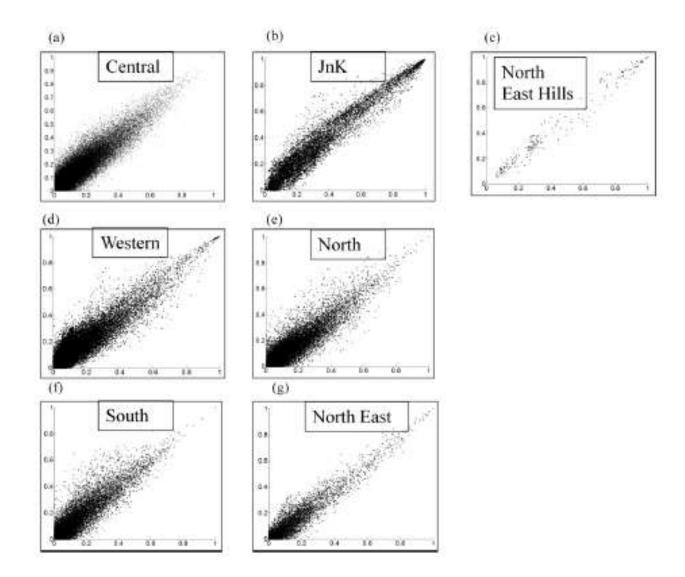
#### Predictor: spatial domain



# Mean and Standard deviation of simulated data



### Zone-wise cross correlation



#### **Generalized Regression Artificial Neural Network GRNN**

INPUT

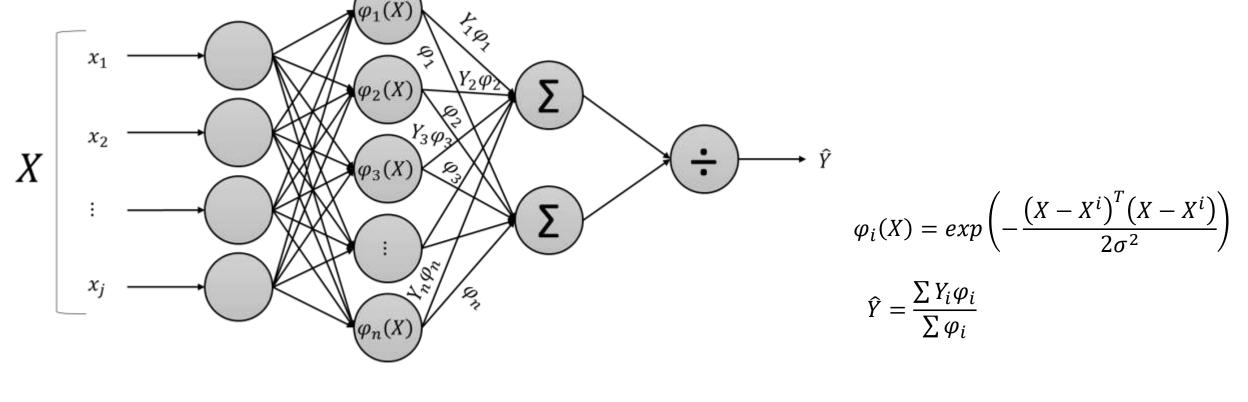
NEURONS

✓ GRNN is a probabilistic-based ANN Model for performing regression.

PATTERN

NEURONS

- ✓ Output is estimated using weighted average of the output of training dataset, where the weight is calculated using the Euclidean distance between the training data and test data.
- If the distance is large then the weight will be very less and if the distance is small it will put more weight to the output.



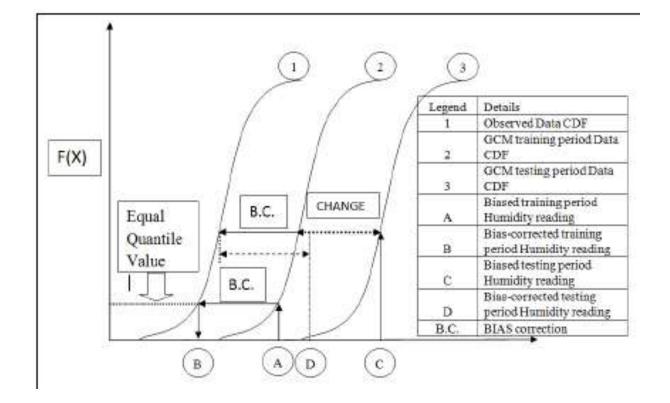
OUTPUT

NEURONS

SUMMATION

NEURONS

### BCSD (Bias Correction)



# **Uncertainty Modeling**

- Multi-model averaging
  - Equal Weights to all the model
- Bayesian Model Averaging
  - Two criteria
    - Performance for historic period
    - Convergence for future period
  - Weights based on the criteria
  - Weighted averaging

#### Selected GCMs and rainfall predictors

#### **Sets of GCMs:**

- ✓ CCCMA CanESM2
- ✓ CNRM CM5
- ✓ MPI ESM MR
- ✓ MPI ESM LR
- ✓ BNU ESM

#### List of Predictors:

- TAS (Near surface air temperature)
- T850 (Air temperature at 850hpa)
- T500 (Air temperature at 500hpa)
- UAS (Eastward near surface wind velocity)
- U850 (Eastward wind velocity at 850hpa)
- VAS (Northward near surface wind velocity)
- V850 (Northward wind velocity at 850hpa)
- Q850 (Specific humidity at 850hpa)
- PSL (Sea level air pressure)
- Z500 (Geopotential height at 500hpa)

#### **Selection Criterion for GCMs and Predictors:**

- Availability of data for the desired period (Historical-1951-2005; Future-2006-2100 for RCP4.5 and RCP8.5)
- Simulation capacity of GCM for a particular predictor
  - Correlation of predictor with the predictand

#### Status

Work completed:

- Downscaling of 5 GCMs (kernel regression) for Precipitation and Temperature
- Development of data portal: under progress (IITB products will be available during next month)

Recruitment:

• Partially done at the institutes

Purchase

• Partially completed

#### Other activities if time permits

• Development of an mobile APP based on downscaled future projections corresponding to different future scenarios

• Possibility of including the projections in WRIS through MoWR



#### Portal- http://www.civil.iitb.ac.in/climate/index.html

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Regional Climate Projections in India with Statistical Downscaling			
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#### Welcome to Regional Climate Projections in India with Statistical Downscaling

If you are using data from Kernel Regression based Statistical Downscaling please cite the following papers-

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> Salvi, K., S. Kannan, and S. Ghosh (2013), High-resolution multisite daily rainfall projections in India with statistical downscaling for climate change impact assessment, J. Geophys. Res. Atmos., 118, 3557-3558, http://dx.doi.org/10.1002/jgrd.50280.

> Kannan, S., and S. Ghosh (2013), A nonparametric Kernel regression model for downscaling multisite daily precipitation in the Mahanadi basin, Water Resour. Res., 49, http://dx.doi.org/10.1002/wrcr.20118.

"This work is funded by Indian National Committee on Climate Change (INCCC), Ministry of Water Resources, Government of India"

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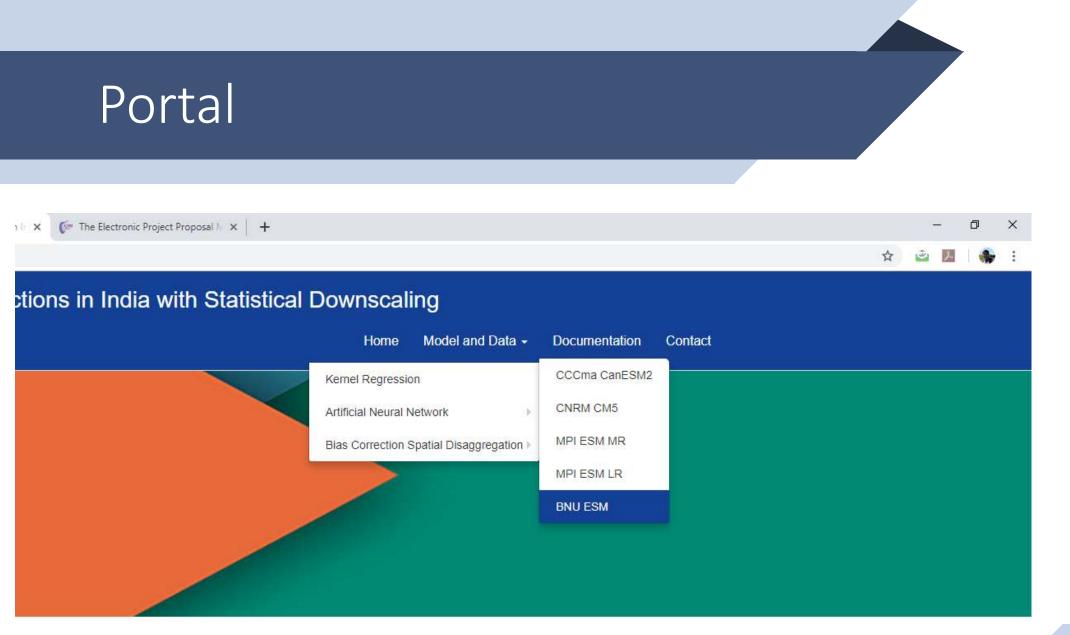
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Contact Us | 022 2576 7319

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### Portal

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# Thank You