

Use of Climate Information in Water Supply Planning

Alison Adams, Ph.D., P.E. CESM-SDWG February 19, 2013



Florida's Largest Regional Public Water Supplier

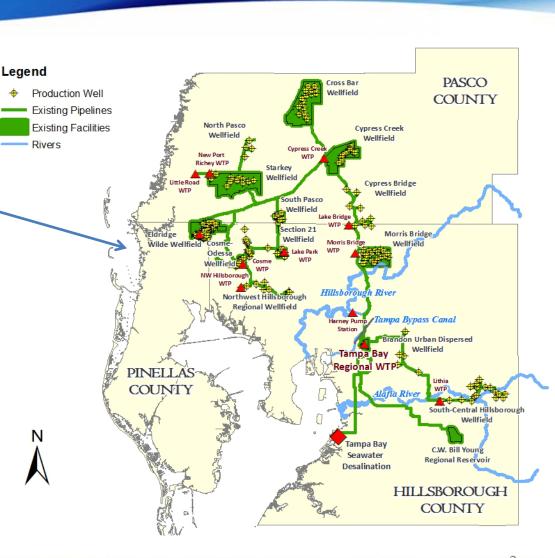


Wholesale drinking water to six governments

2.4 Million Residents

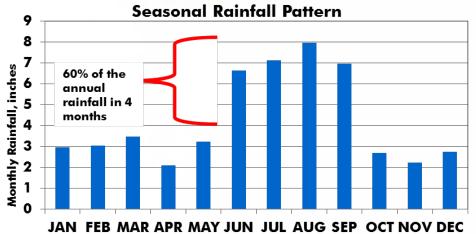
220-250 mgd annual average

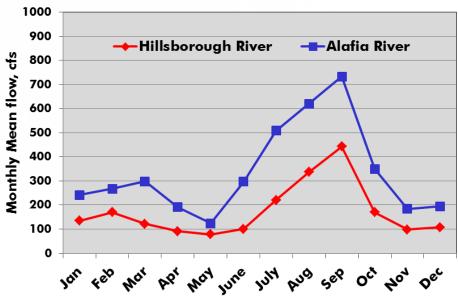
Seasonal to multi-year variable climate





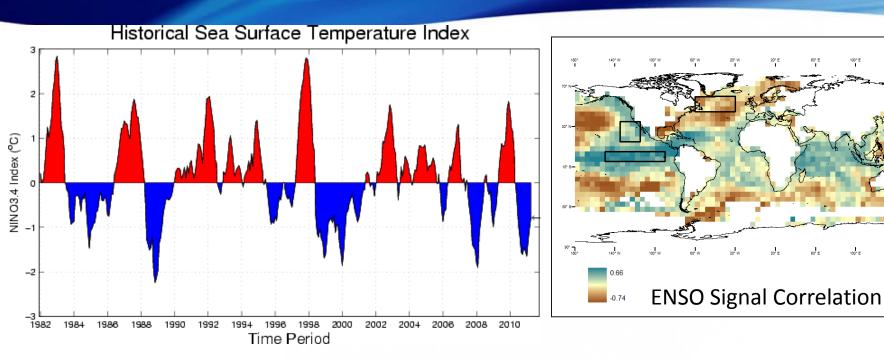
Why Climate Variability is Important







ENSO Affects Local Rainfall Patterns

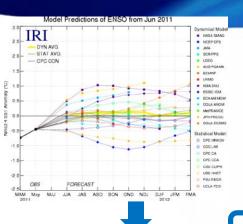


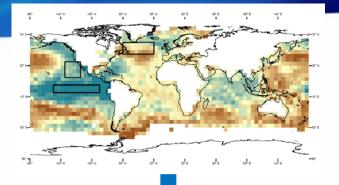
Plant City rainfall probabilities (%) conditional of La Niña

	Below Normal	Normal	Above Normal
DJF	<mark>65</mark>	35	0
JFM	85	12	3
FMA	71	21	8



Tampa Bay Water's Seasonal Outlook





Climate Outlook & Real time observation

Below Above Normal Normal Normal DJF 65 35 0 IFM 85 12 3

Contingency Table



CENTER CARRESTONICA CONTROLLAR CARRESTONICA CONTROLLAR

Conditional Markov Rainfall Model

Rainfall/Runoff Model

Tampa Bay Climate Outlook: March 2010

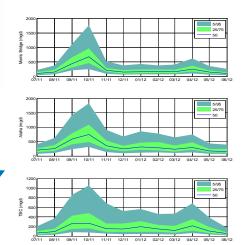
Continued EL Nino Advisory: Both Dynamical and Statistical Models

Climate Outloo.

All models show a significant El Nino persistence through March-April-May 2010 (see Figure 1 and Table 1, based on updated ENSO forecast on Feb 22, 2010). By April-May-June most of the models shows transitioning to ENSO-Neutral conditions, All models aggree in decreasing sea surface temperature in Spring. Table A.1 in Appends shows the accuracy of these models based on one step prediction skill. The average Nino 3.4 Sea Surface Temperature anomalies for the latest week that data is available shows a + 1.2°C departure, which indicates El Nino conditions.

How did we fare in Januarry and February?

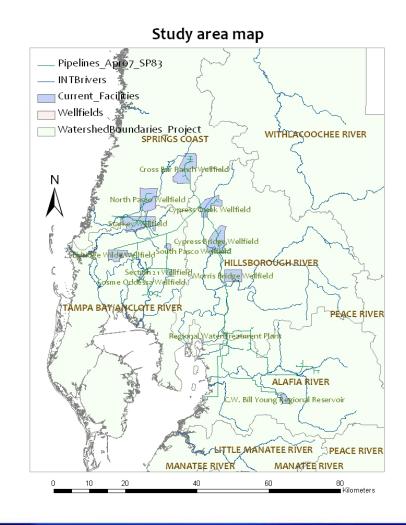
Given that we are in El Nino for the past few month it's appropriate to see how January and
"«bnuary 2010 (the month where the effect of El Nino are felt) rainfall were compared to historical
"stions. Figure 2 presents the three Timpa Bay Water stations (St. Leo, Plant City, and CVC
"we used to model surface flows. The figures show box plots of each stations based
"sta. Black crosses indicate the 2008 data. Magenta crosses are for 2009 and
"se figures show a significant change compared to 2009. In all the cr
""orical median or above it, which shows the effect of El Ni-





Tampa Bay Water's Climate Change Assessment Project

.... In this project we are using dynamically and statistically downscaled climate model output to drive hydrologic models and explore potential impacts of climate variability and climate change on water availability and water allocation decisions

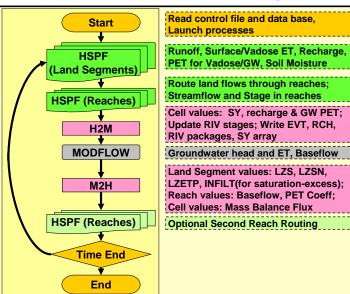




Integrated Hydrologic Model (IHM) Hydrologic Processes

Evapotranspiration Infiltration SURFACE (HSPF Evapotranspiration GROUND WATER COMPONENT (MODFLOW)

IHM Sequential Integration



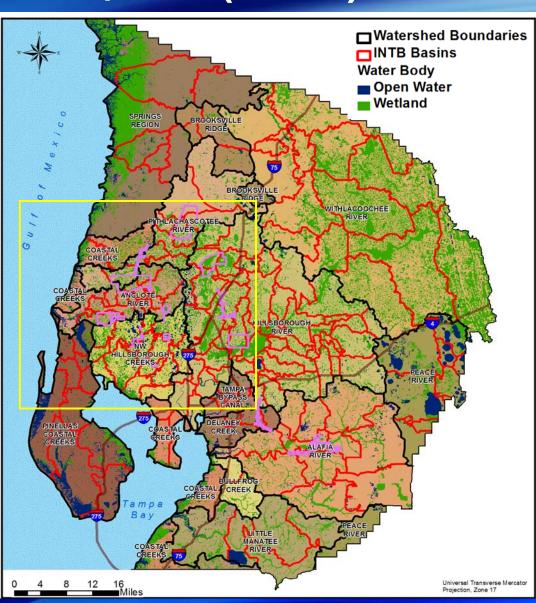


Integrated Northern Tampa Bay Model Surface-Water Component (HSPF)

- Convective Rainfall (4 months)
 - 60% volume / 75% events
 - 1.25-mile event spatial scale
- 65% of basins with 2 mile radius
- Rain input: 300 gauges, 15-min.
- ET 5x seasonal variation

Average Annual Budget 1989-98

		Flux
Budget Term	Percent	(in/yr)
Evap. & Transp.	69	38.0
Stream & Spring Q	21	11.0
Well Pumping	5	3.0
GW Flow to Gulf	3	1.5
SW Pumping	1	0.5
Other GW Outflows	1	0.5
Total	100	54.5





Current Research outline

1. Statistical downscaling

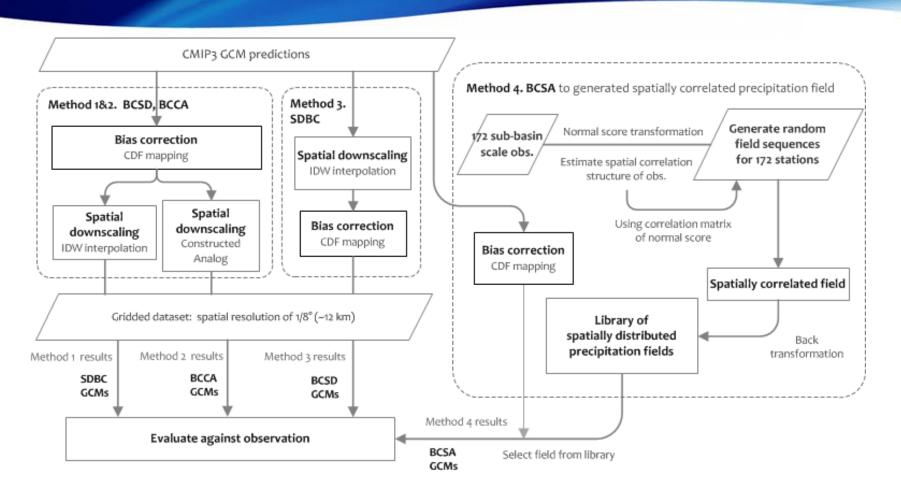
- Comparative evaluation of 4 methods (BCSD_daily, BCCA, SDBC, BCSA)
 - Ready to submit Hydrology and Earth System Science
- Hydrologic simulation
 - Submitting to ASABE transaction

2. Evaluation of downscaled reanalysis data

- R1+MM5 (Hwang et al., 2011)
- R2+RSM (Stefanova et al., 2011)
- ERA40+RSM (Stefanova et al., 2011)
- 20CR+RSM (DiNapoli and Misra, 2012)
- Submitting to JAWRA
- 3. Uncertainty of Bias-correction in climate change impact assessment

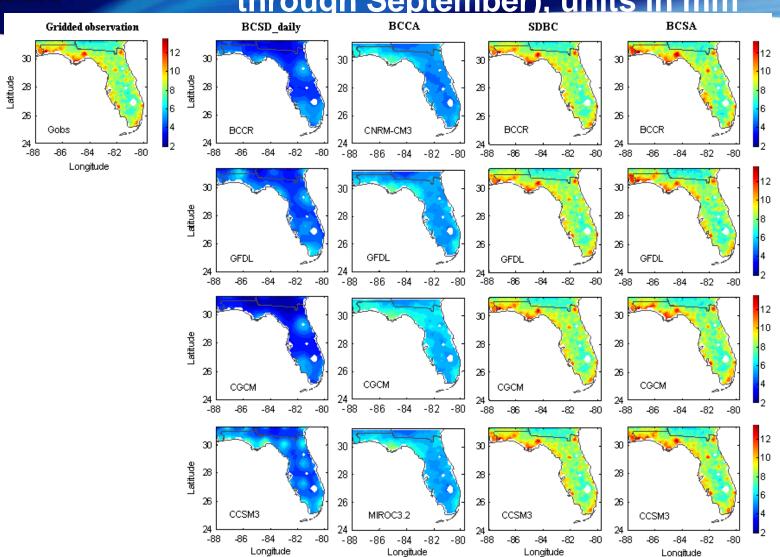


Methodology



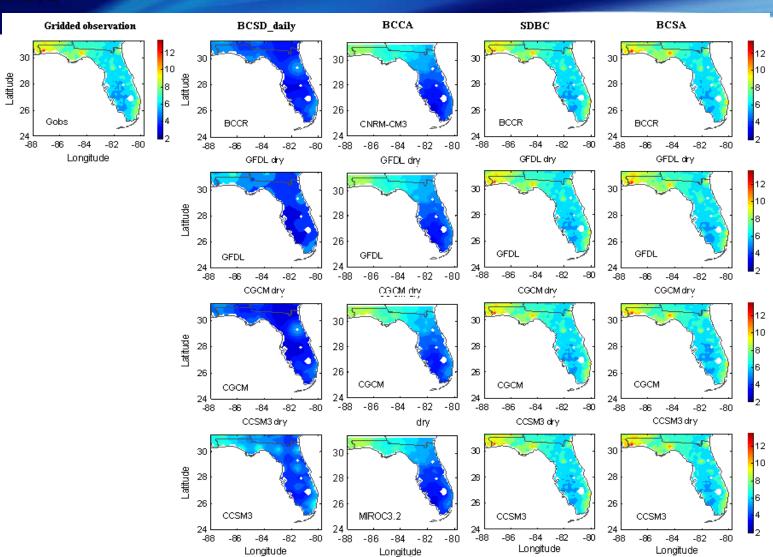


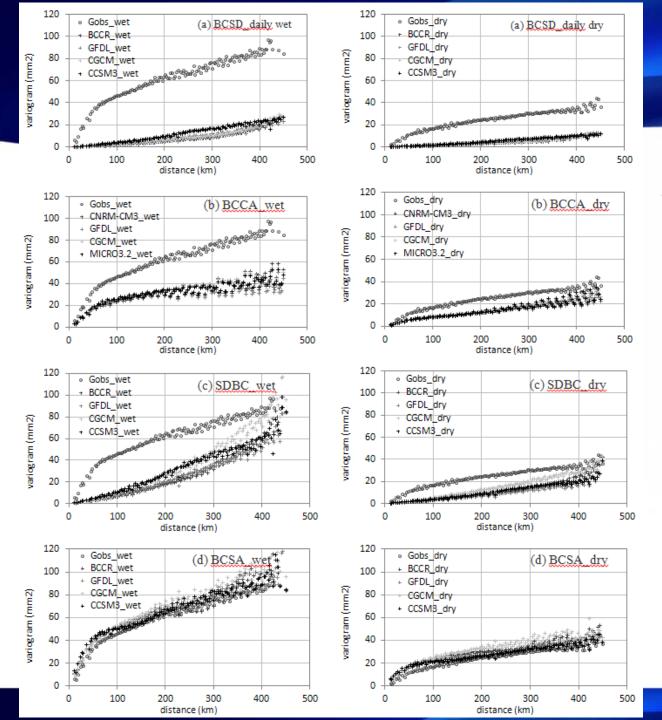
Spatial distribution of the temporal standard deviation for wet season (June through September), units in mm





The same but for dry season (October through May)





Spatial variability (Variograms)



Dynamical Downscaling

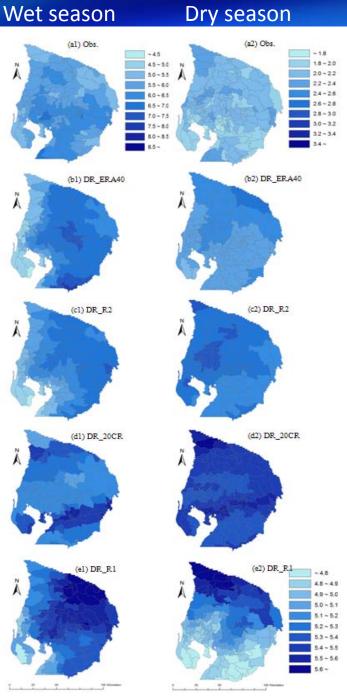
- Assessment of the utility of dynamically-downscaled regional reanalysis data to predict streamflow in west central Florida
 - Reanalysis data robust proxy of historic atmospheric observations
 - Verifying accurate prediction of historic climatic and hydrologic behavior using reanalysis data is an essential first step before using retrospective and future GCM projections to predict potential hydrologic impacts of future climate change



Spatial distribution of daily mean precipitation

- Study period from 1989 to 2001
 - 1. R1+MM5 (Hwang et al., 2011)
 - 2. R2+RSM (Stefanova et al., 2011) 1979-2001
 - ERA40+RSM (Stefanova et al., 2011 1979-2001
 - 4. 20CR+RSM (DiNapoli and Misra, 2012)
 1903-2008

IHM calibration/verification period 1989-2006

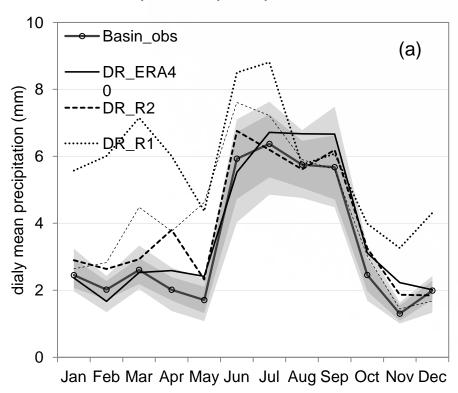




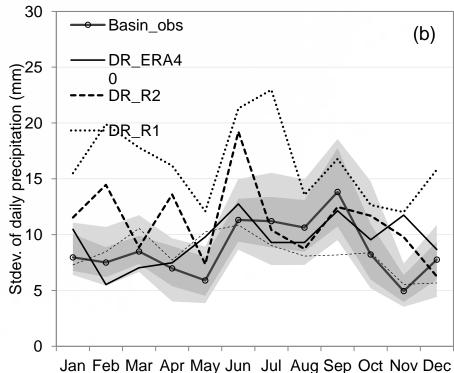
Comparison of the mean annual cycles of (a) monthly mean and (b) standard deviation of daily precipitation.

Raw results

monthly mean precipitation

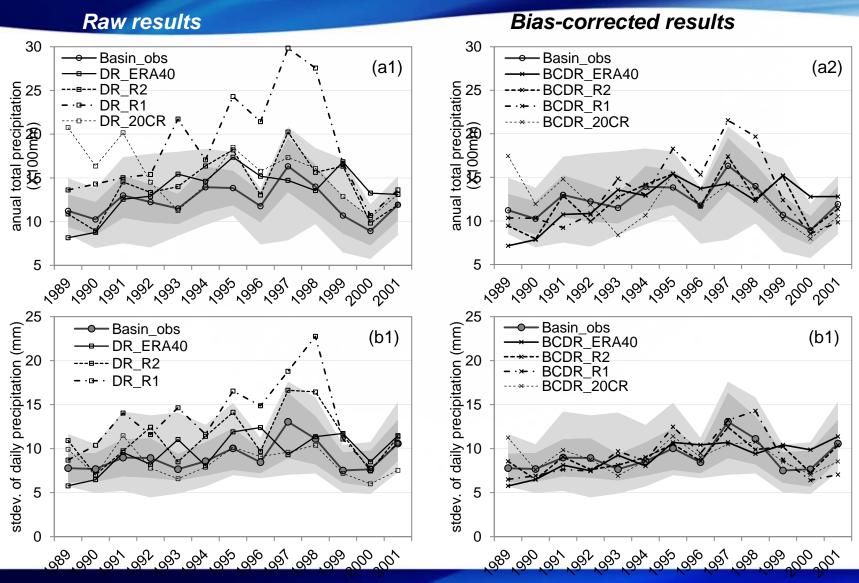


standard deviation of daily precipitation



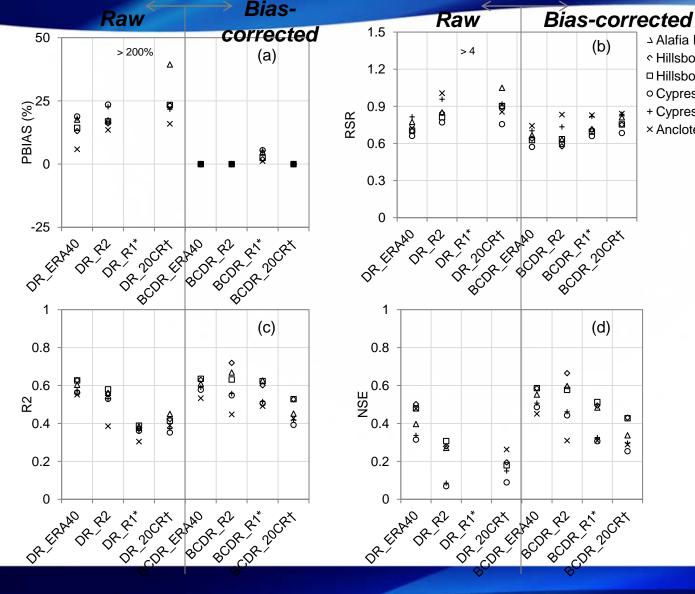


Comparison of time series of (a) annual total precipitation and (b) standard deviation of daily precipitation over the year



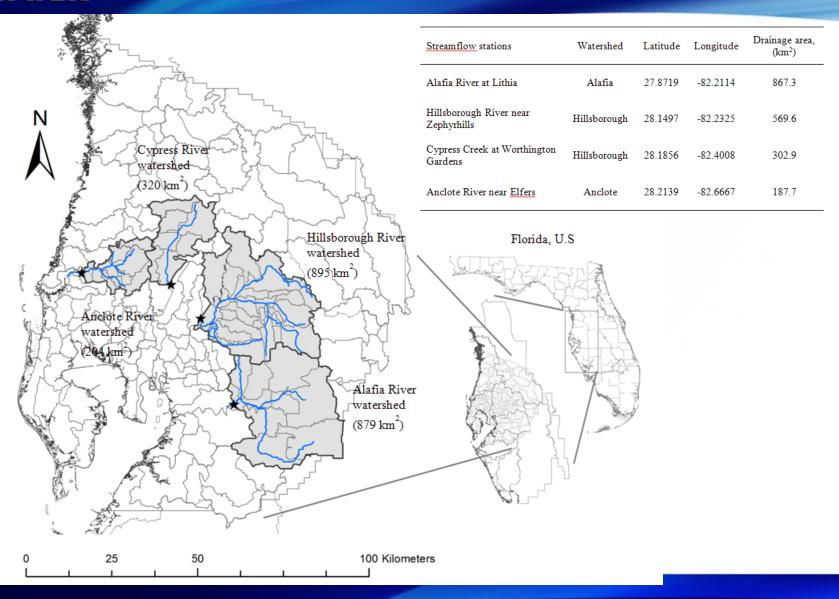


Comparison of error statistics of monthly areal precipitation predictions



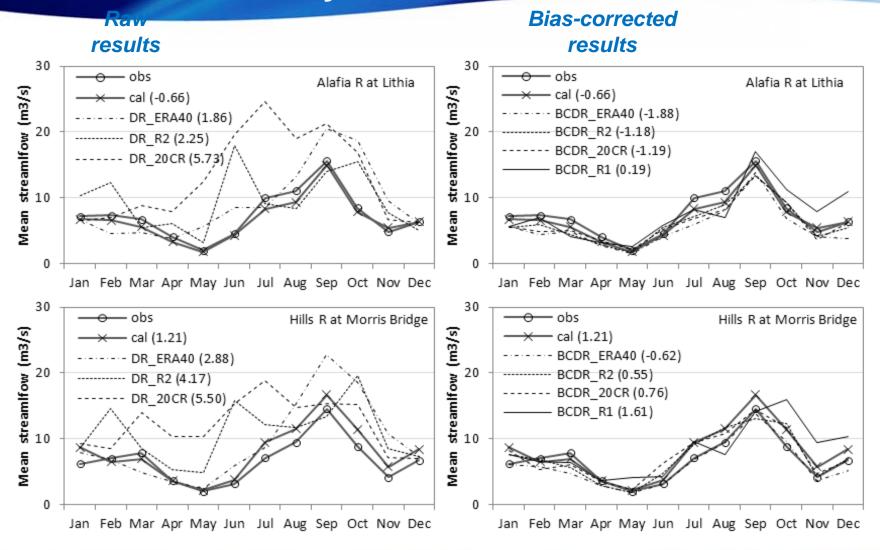
- △ Alafia R at Lithia
- Hillsborough R at Morris Bridge
- □ Hillsborough R near Tampa
- Cypress Creek at Worthington
- + Cypress Creek near Sulfur spring
- × Anclote R near Elfers

Hydrologic implication





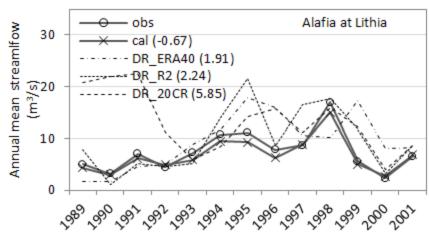
Comparison of observed vs. simulated mean monthly streamflow

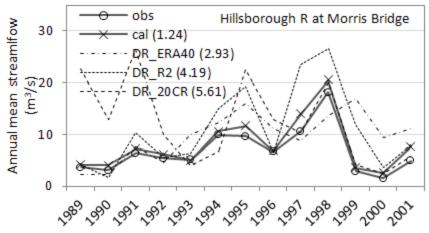


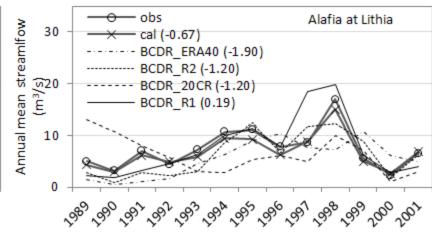


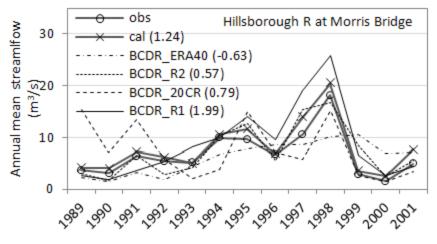
Comparison of observed vs. simulated annual time series

Raw results







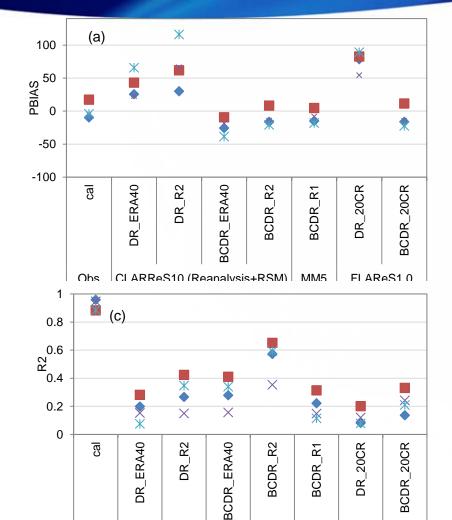


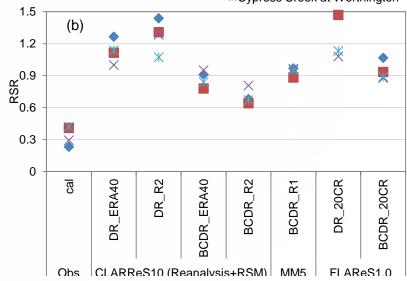


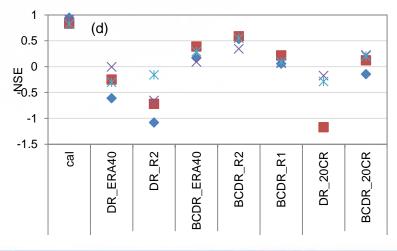
Comparison of error statistics of monthly streamflow simulations for each target station; (a) PBIAS, (b) RSR, (c) R², and (d) NSE Alafia R at Lithia



- Hillsborough R at Morris Bridge
- ×Anclote R near Elfers
- **X**Cypress Creek at Worthington









Uncertainty of Bias-correction in climate change impact assessment



CLAREnCE10 data

http://coaps.fsu.edu/CLARReS10/index.shtml

- 3 GCMs + Regional Spectral Model (RSM)
 - CCSM, HadCM3, and GFDL (not available yet)
- Spatial resolution (10kmx10km) over southeastern US
- Variables
 - hourly Prec., humidity, wind speed, roughness, etc.
 - daily Tmax/min data
 - Daily bias-corrected Prec. data are available
- Retrospective simulation period
 - 1968-2000
- Future simulation (AR4 A2 scenario)
 - 2038-2070

Bias-correction (BC) Methodology

4 Future Bias Correction methods

- Correct using historic bias amount corresponding the 'magnitude' of future prediction (CDFm, Wood et al)
- Correct using historic bias amount corresponding the 'Percentile' of future prediction (EDCDFm, Li et al., 2010)
- Correct using historic bias percentage corresponding the 'magnitude' of future prediction (CDFm_%bias)
- Correct using historic bias percentage corresponding the 'Percentile' of future prediction (EDCDFm_%bias)

3 methods for CDF development

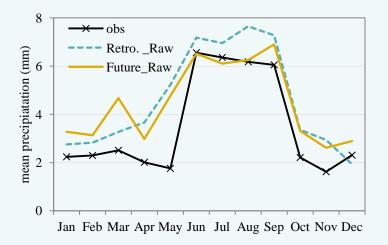
- Monthly CDF (≈30data)
- 2. CDF for moving window (\pm 30 days, 61data)
- CDF for moving window (± 15 days, 31data)

Total 12 combination of methodologies!

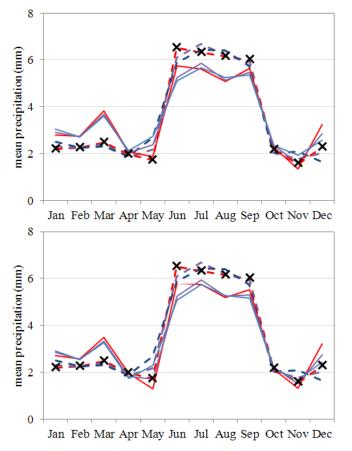
1. Mean daily precipitation

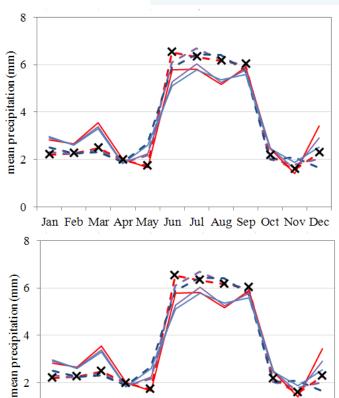
Raw results

HadCM3+RSM



Bias-corrected results





Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec

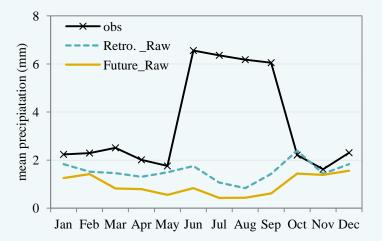


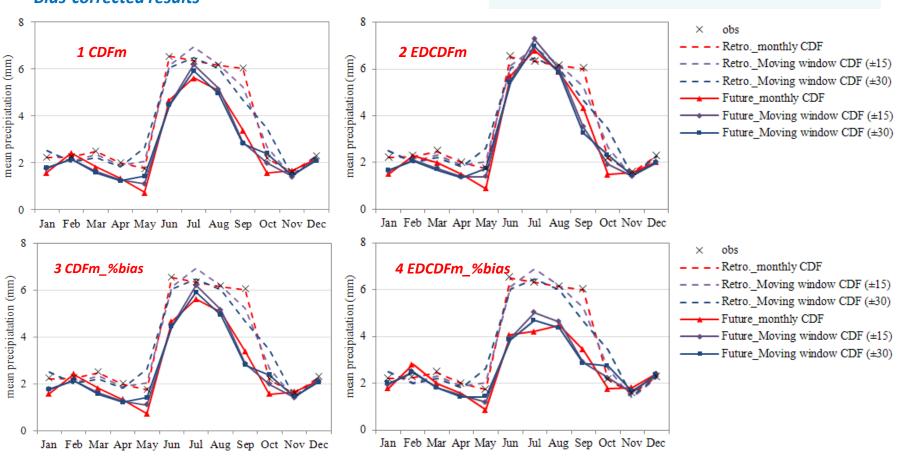
Retro. monthly CDF

1. Mean daily precipitation

Raw results

CCSM+RSM

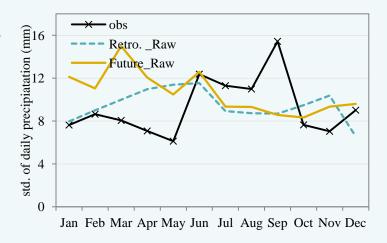


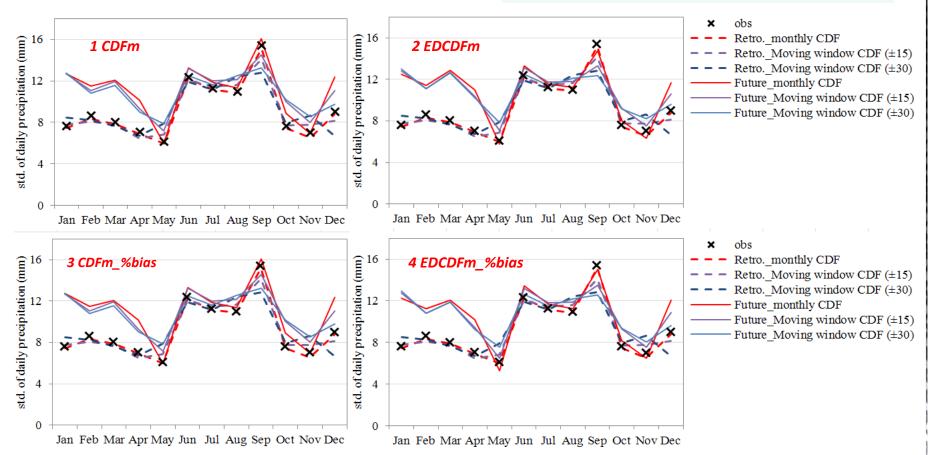


2. Std. of daily precipitation

Raw results

HadCM3+RSM

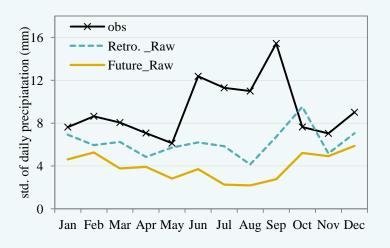


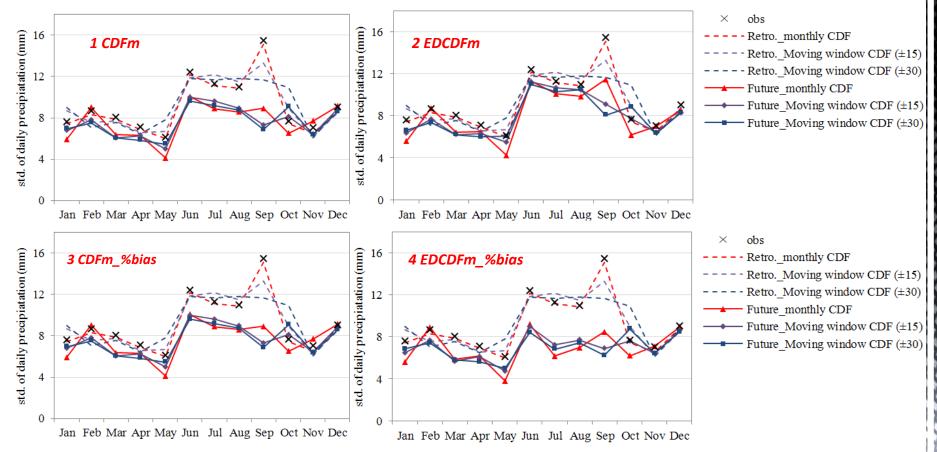


2. Std. of daily precipitation

Raw results

CCSM+RSM







• QUESTIONS??