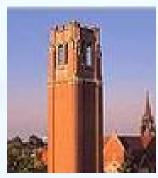
## The University of Florida Water Institute



Wendy Graham, Ph. D., Water Institute Direct orCarl Swisher Eminent Scholar



## 01 Introduction

#### Florida Water and Climate Alliance A Stakeholder-Scientist Network



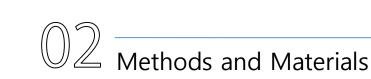


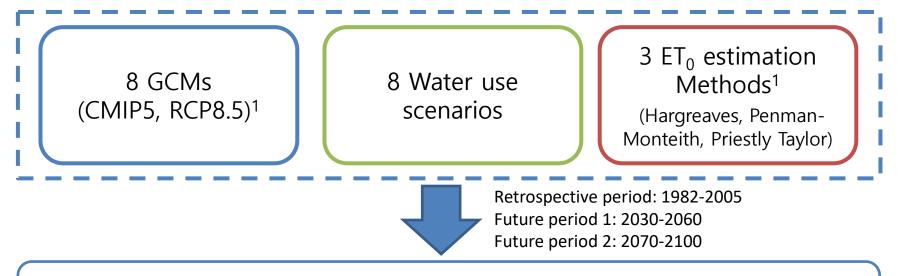
Goal: To increase the regional relevance and usability of climate and sea level rise models for the specific needs of water suppliers and resources managers in Florida.

## 01 Introduction

Tampa Bay Water Project Objectives

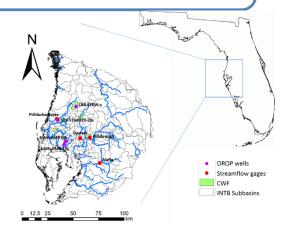
- Evaluate the ability of Global Climate Models (GCMs) to reproduce retrospective temperature and rainfall in the Tampa Bay Region
- Evaluate the ability of GCMs to reproduce retrospective hydrologic behavior when used with the calibrated Integrated Northern Tampa Bay (INTB) Model
- Evaluate changes in hydrology resulting from GCM future climate projections
- Quantify the relative uncertainties of changes in future climate and water use scenarios for water supply planning



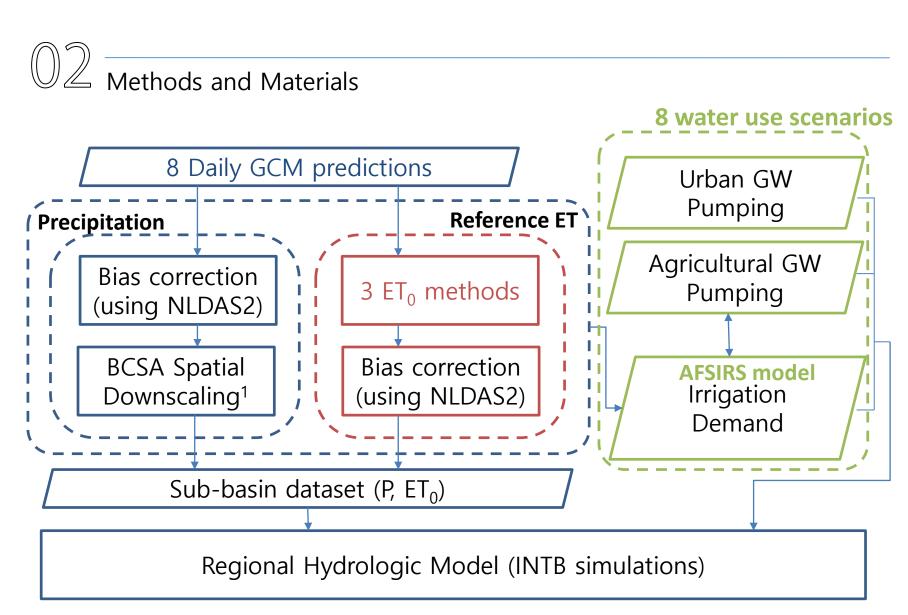


Changes in hydrology, changes in environmental compliance, ability to meet future demand.

- Historic climate data: NLDAS-2 (1/8<sup>th</sup> degree grid)
- Hydrologic model: Integrated Hydrologic Model (IHM)
- Study region: Integrated Northern Tampa Bay (INTB)



Ref: <sup>1</sup>Chang et al (2016)



Variance-based GSA and Tukey's HSD test to evaluate the results.

Methods and Materials

Bias Correction Stochastic Analog (BCSA) Downscaling Technique<sup>1</sup>

#### What we did

Developed a new statistical downscaling method (BCSA) and compared it to existing methods (BCSD, SDBC, BCCA)

#### Why we did it

Existing statistical downscaling methods did not reproduce spatiotemporal rainfall characteristics in Florida well

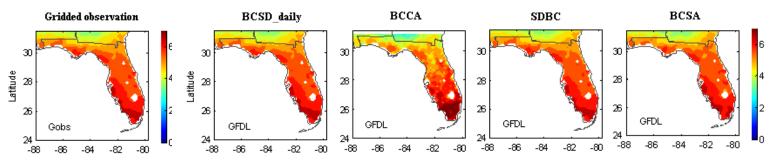
#### What we found...

Choice of statistical downscaling method matters in Florida!

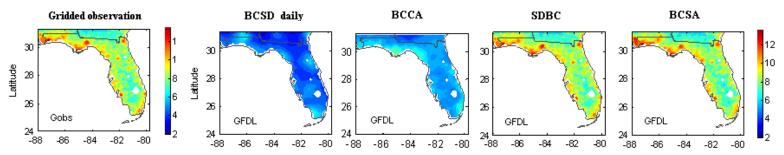
Note: Even dynamic downscaling requires bias correction before use in hydrologic models !

## 03 Results: Comparison of Downscaling Methods

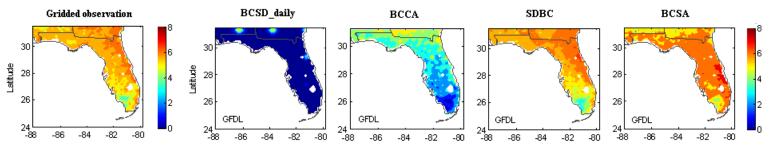
#### Wet season average daily rainfall



#### Wet season standard deviation of daily rainfall

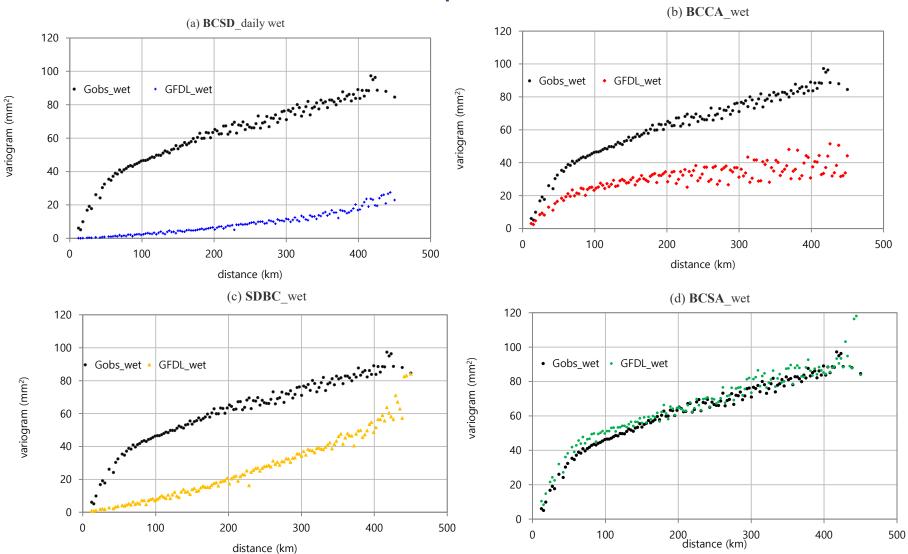


#### Wet season frequency of wet spells >5 days





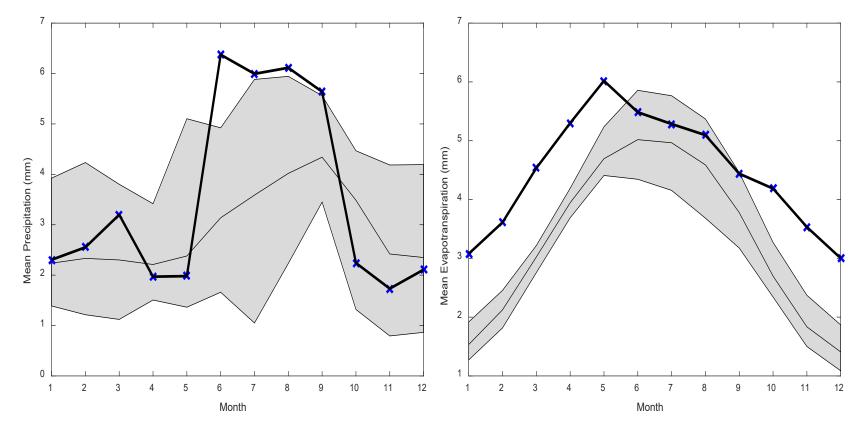
Wet season spatial structure





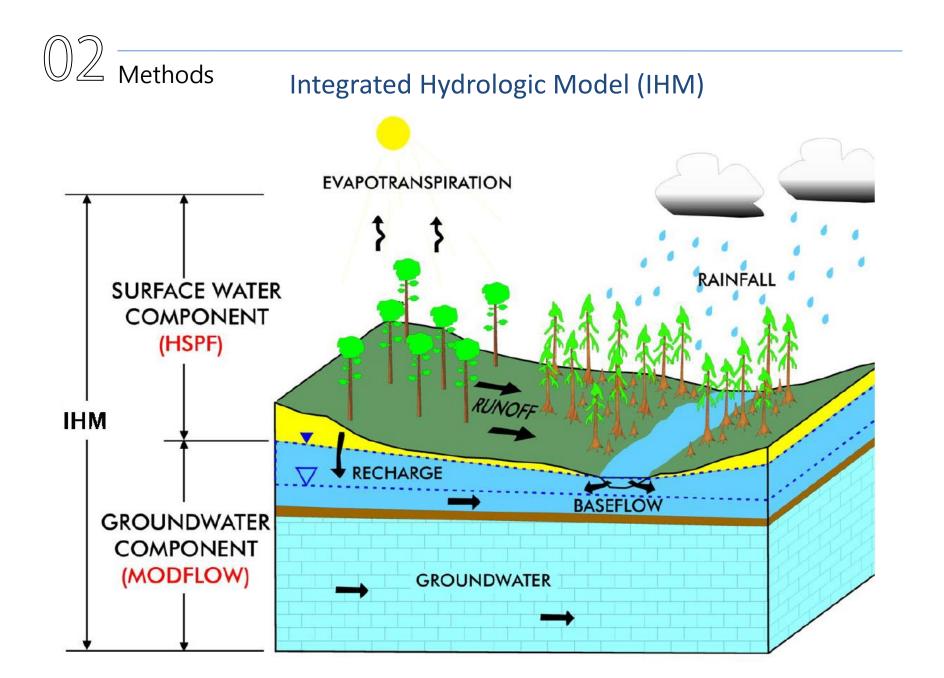
 $03 \overline{\text{Results}}$ 

#### Comparison of raw and downscaled, bias-corrected retrospective P and ET<sub>0</sub>



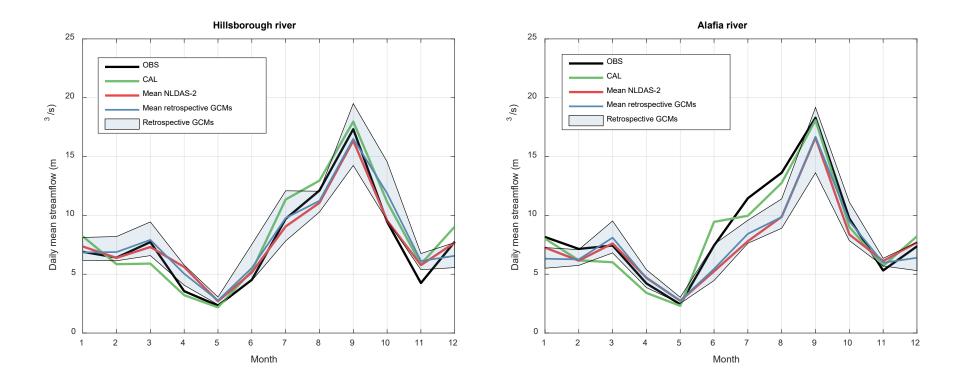
All GCMs must be bias corrected before use with INTB !!







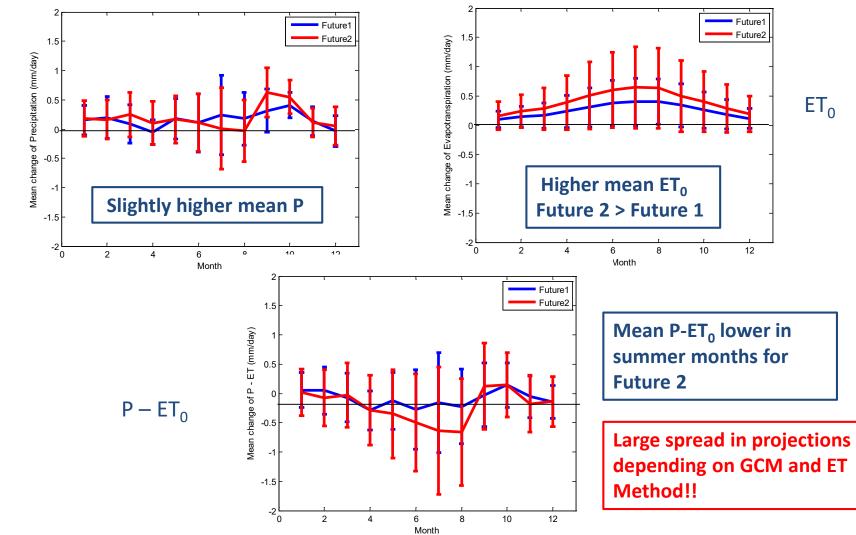
#### Comparison of retrospective monthly streamflow predicitions



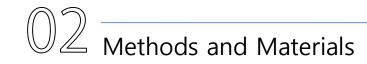
**Retrospective streamflow OK after P and ET<sub>0</sub> bias-correction** 



#### Projected future monthly change in P, ET<sub>0</sub> and (P-ET<sub>0</sub>)

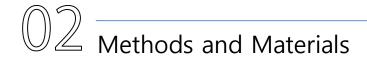


Ρ



### 2045 Water Use Scenarios: Demand Assumptions

- Tampa Bay Water Urban demand projected to be ~280 MGD in 2045 assuming active water conservation.
- Climate-driven agricultural irrigation water demand estimated by the AFSIRS model using GCM future projections
- Land use change not considered. Increases in demand assumed to be the result of intensification on existing lands



## 2045 Water Use Scenarios

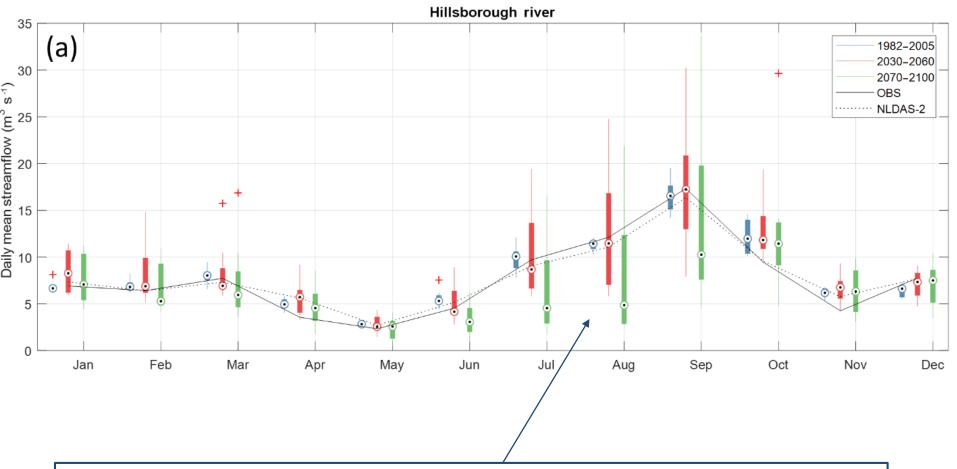
|  | Groundwater Pumping for Retrospective Climate<br>(MGD) |                                      |   |                        |  |
|--|--|--------------------------------------|---|------------------------|--|
| Scenario   | Irrigation<br>Demand                                   | Agricultural<br>Pumping <sup>1</sup> | Urban<br>Pumping <sup>2</sup><br>(CWF) <sup>3</sup> | Total<br>GW<br>Pumping |  |
| No groundwater pumping                                   | 0  | 0                                    | 0   | 0                      |  |
| No urban groundwater pumping                             | 163  | 192                                  | 0   | 192                    |  |
| No agricultural groundwater pumping                      | 0  | 0                                    | 356(90)   | 356                    |  |
| Agricultural adaptation (25% reduction) <sup>4</sup>     | 129  | 152                                  | 356(90)   | 508                    |  |
| Business as Usual  | 163  | 192                                  | 356(90)   | 548                    |  |
| Increase agricultural demand (25% increase) <sup>4</sup> | 203  | 239                                  | 356(90)   | 595                    |  |
| Increase CWF groundwater pumping (45%)                   | 163  | 192                                  | 396(130)  | 588                    |  |
| Increase all urban groundwater pumping (45%)             | 163  | 192                                  | 514(130)  | 706                    |  |

- <sup>1</sup>Agricultural = agricultural & recreational
- <sup>2</sup>Urban = public water supply & industrial
- <sup>3</sup>CWFs: Consolidated wellfields
- <sup>4</sup>redcution/increase over climate driven demand



Results

#### Retrospective versus future mean daily streamflow (Business as usual scenario, Hargreaves ET)



Lower mean streamflow in summer months. Future 2 < Future 1. Large spread in future projections!





Results

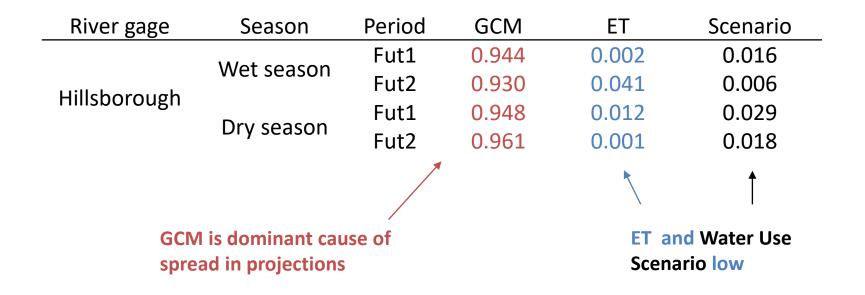
#### Retrospective versus future mean daily groundwater level (Business as usual scenario, Hargreaves ET)





## **Global Sensitivity Analysis**

#### The first order sensitivity index of change in streamflow





## **Global Sensitivity Analysis**

#### The first order sensitivity index of change in groundwater level

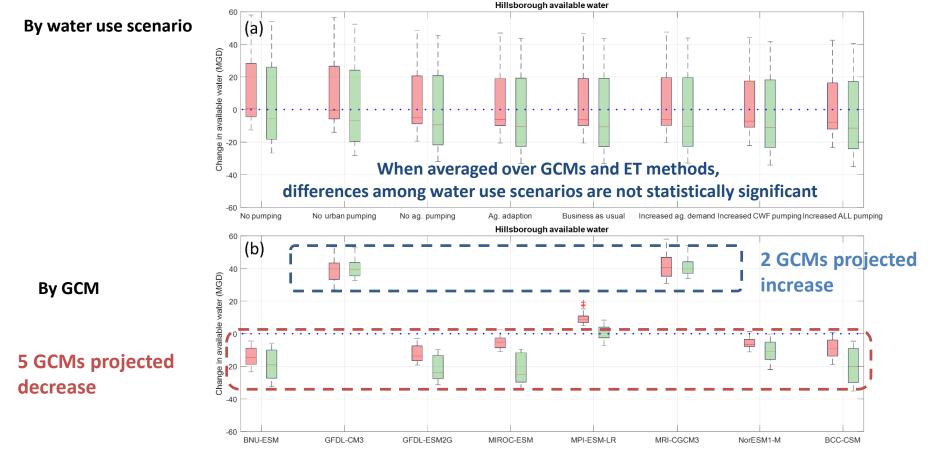
GCM and water use scenario are both significant sources of spread

| OROP well   | Season     | Period | GCM   | MET   | Scenario |
|-------------|------------|--------|-------|-------|----------|
| NWH-RMP-08s | Wet season | Fut1   | 0.442 | 0.005 | 0.501    |
|             |            | Fut2   | 0.576 | 0.004 | 0.278    |
|             | Dry season | Fut1   | 0.475 | 0.007 | 0.435    |
|             |            | Fut2   | 0.550 | 0.002 | 0.288    |



**B** Results

### Change in available water from the Hillsborough River



When averaged over water use scenarios and ET method, Differences among GCMs are statistically significant

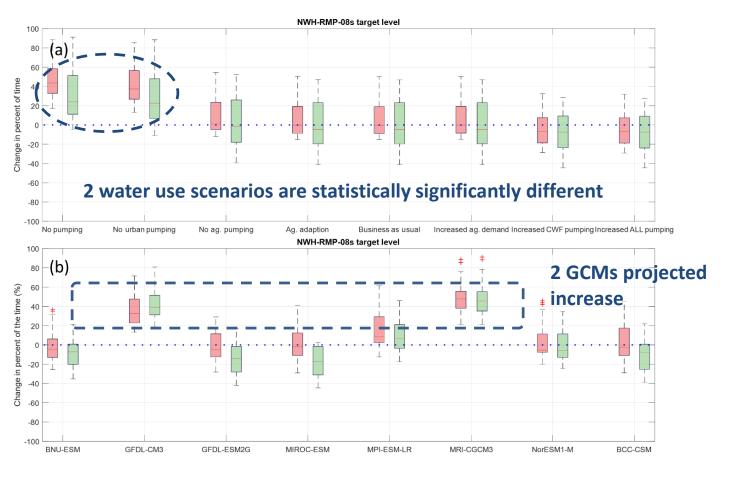


## Results

### Change in percent of the time that GW is above target level

By water use scenario

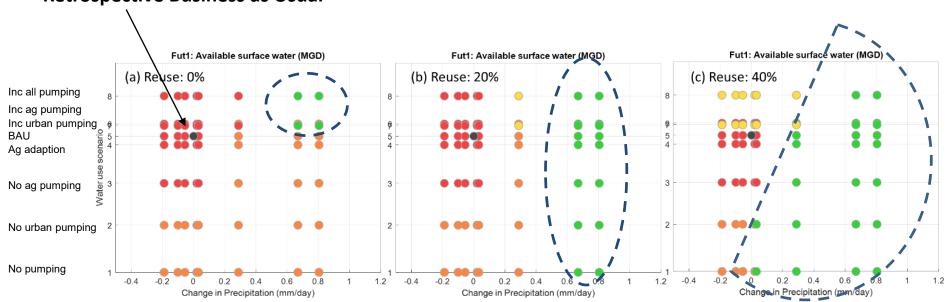
Reduction in municipal pumping leads to increase in time GW is above target at this well



By GCM

#### Differences among GCMs are statistically significant

## Scenario Discovery Analysis<sup>1</sup>: Ability to meet Tampa Bay Water projected 2045 demand and meet environmental regulations



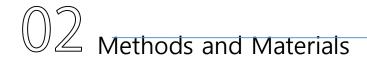
#### **Retrospective Business as Usual**

- Meets 2045 demand and maintains compliance with groundwater regulations
- Meets 2045 demand but does not maintains compliance with regulations
- Does not meet 2045 demand but maintains compliance with regulations
- Does not meet 2045 demand or maintains compliance regulations



#### Take home messages

- Downscaling technique matters! Even dynamically downscaled GCMs must be bias corrected.
- Projected changes in streamflow and groundwater level vary depending on choice of GCM, ET method, water use scenario and RCP scenario
- The uncertainty attributed to GCMs was the dominant factor influencing different future streamflow projections.
- The uncertainty attributed to GCM and water use scenario both contributed to significant differences in future groundwater level projections.
- For the three ET methods shown here, uncertainty is relatively small.
- Climate models projected significantly different changes in streamflow and groundwater. 5 to 6 GCMs among 8 projected decreases in water availability.
- Even with active conservation and dramatic increases in use of reclaimed water current regulations on water withdrawals may have to adapt to future climate.



#### **Evapotranspiration Method Selection**

- What we did
  - Evaluated uncertainty in P, T, and ET<sub>0</sub> projections using Global Sensitivity Analysis and Monte Carlo Filtering
- Why we did it
  - To develop an appropriate ensemble of GCMs, ET methods and RCP trajectories for evaluating future climate change
- What we found
  - Choice of ET method matters!
  - Evaluate impacts of future projections over an ensemble of GCMs and a variety of ET methods and RCP trajectories

# 2070-2100 Change in Annual P-ET<sub>0</sub> by ET method (averaged over GCMs and RCPs)

