

# Alternate methodologies for computing evapotranspiration at a 2-kilometer resolution for Florida

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U.S. Department of the Interior

**U.S. Geological Survey** 

## Outline

- USGS gridded evapotranspiration products
- Alternate methodologies
- Bias analysis at station locations



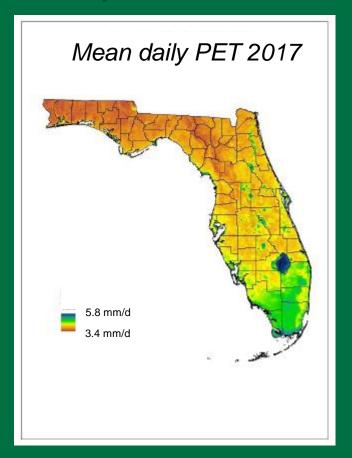
# Contributing researchers

- David Sumner, USGS Caribbean-Florida Water Science Center
- Barclay Shoemaker, USGS Caribbean-Florida Water Science Center
- John Mecikalski, University of Alabama in Huntsville
- Qinglong (Gary) Wu, South Florida Water Management District



# Gridded evapotranspiration products

- Potential and Reference Evapotranspiration
  - ~2-kilometer spatial resolution
    - grid is identical to NEXRAD grid
    - extent of Florida
    - includes water bodies
  - Daily time step
    - 1985-present





# Gridded evapotranspiration products

- Potential evapotranspiration
  - Priestley-Taylor equation
  - f (Rs, Tmax, Tmin, RHmax, Rhmin)
    - Albedo constant value for land and for water
- Reference evapotranspiration
  - Penman-Monteith equation
  - f (Rs, Tmax, Tmin, RHmax, RHmin, Wind)
    - Albedo is constant value (grass)

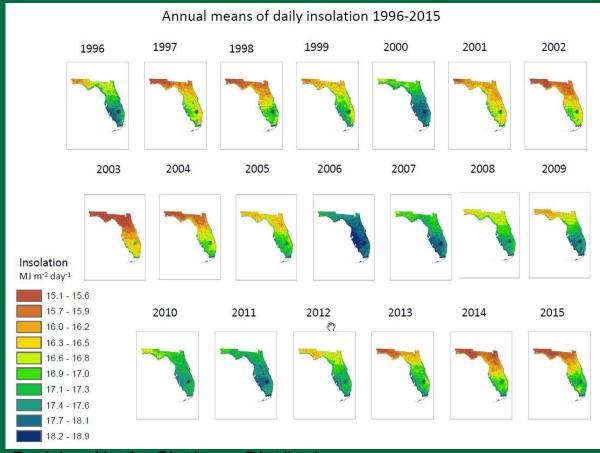


# Gridded evapotranspiration products, continued

Input Solar Radiation

Geostationary Operational Environmental Satellite

(GOES)





# Gridded evapotranspiration products, continued

- Input Meteorological variables:
  - 1985-1995: North American Regional Reanalysis



# North American Regional Reanalysis

- NOAA model used in support of weather forecasting
- 32 km spatial resolution
  - Interpolated using Radial Basis Function (RBF)
- 3 hour time step
  - Min and max of atmospheric variables based on 8 daily values

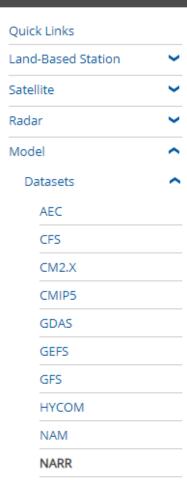




Formerly the National Climatic Data Center (NCDC)... more about NCEI »

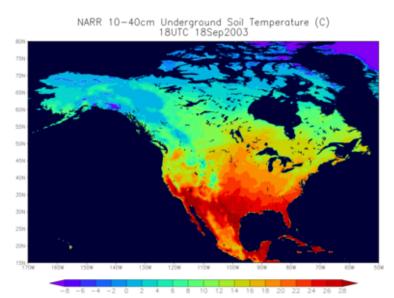
Home Climate Information Data Access Customer Support Contact About Search

Home > Data Access > Model > Datasets > North American Regional Reanalysis (NARR)



### North American Regional Reanalysis (NARR)

The North American Regional Reanalysis (NARR) is a regional reanalysis of North America containing temperatures, winds, moisture, soil data, and dozens of other parameters. Produced by the National Centers for Environmental Prediction (NCEP), the NARR model takes in, or assimilates, a great amount of observational data to produce a long-term picture of weather over North America. The data that are assimilated in order to initialize the model to real-world conditions are temperatures, winds, and moisture from radiosondes as well as pressure data from surface observations. Also included in this dataset are dropsondes, pibals, aircraft temperatures and winds, satellite radiance (a measure of heat) from



A sub-region plot of NARR underground soil temperature (a layer from 10 cm to 40 cm below ground) at 18 UTC on September 18, 2003. This image was produced by downloading one file of NARR data through NOMADS and visualizing with the Grid Analysis and Display System (GrADS).

polar (orbiting Earth) satellites, and cloud drift winds from geostationary (fixed at one location viewing Earth) satellites.

# Gridded evapotranspiration products, continued

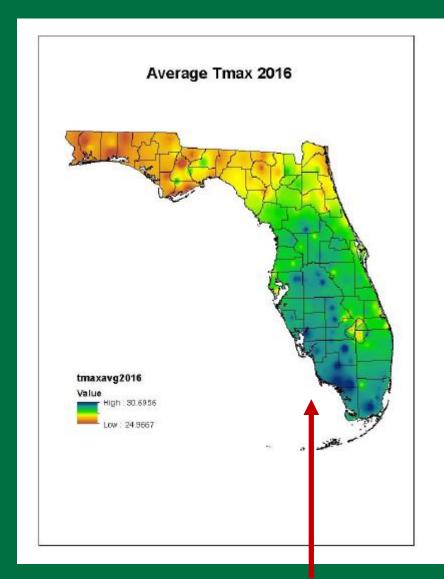
- Input Meteorological variables:
  - 1985-1995: North American Regional Reanalysis

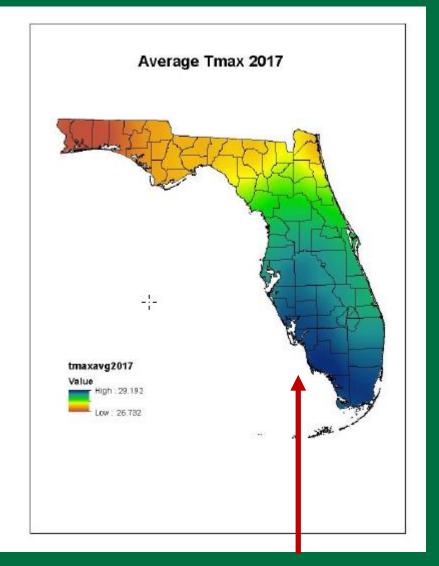


# Gridded evapotranspiration products, continued

- Input Meteorological variables:
  - 1985-1995: North American Regional Reanalysis
  - 1996-2017: interpolated from weather station data
    - FAWN, DBHYDRO, NOAA weather station data
    - 1996-2016: Inverse-Distance Weighted (IDW) interpolation
    - 2017: RBF interpolation









*IDW* 

### http://fl.water.usgs.gov/et/

#### **Evapotranspiration Information and Data**

The U.S. Geological Survey Florida Evapotranspiration Network is a network of 15 data collection sites representing various land cover types, which provide long-term, accurate, and unbiased information that meets the needs of many diverse users. The USGS collects the evapotranspiration data needed by Federal, State, and local agencies for planning and operating water-resources projects and regulatory programs.

The links below allow you to find information and data about Florida's evapotranspiration resources.

#### **Evapotranspiration Data**



#### **Data Collection Sites**

The map shows current and past evapotranspiration data collection sites in Florida; click the map to go to a full size version with links to NWISWeb data.

#### Statewide Evapotranspiration Data - (2km Daily)

#### ET Data County and State 1985-2016

- 1985 1986 1987 1988 1989
- 1991 1992 1993 1994
- 1996 1997 1998 1999
- 2001 2002 2003 2004
- 2006 2007 2008 2009
- 2011 2012 2013 2014
- 2015 2016

#### Information Resources

#### **Evapotranspiration Publications**

- USGS Evapotranspiration Publications for Florida
- Technical Report: Satellite-based Evapotranspiration Estimates Over Florida
- · InTechWeb Report: Use of Visible Geostationary Operational Meteorological Satellite Imagery in Mapping Reference and Potential Evapotranspiration over Florida

#### Other USGS Evapotranspiration Resources

USGS Nevada Water Science Center - Evapotranspiration Studies

#### Need More Information?

Contact the Florida <u>Evapotranspiration Specialists</u>



#### Statewide Evapotranspiration Data - (2km Daily)

| File              | Size       | File              | Size      | File             | Size       | File              | Size        |
|-------------------|------------|-------------------|-----------|------------------|------------|-------------------|-------------|
| Alachua 2016      | 6,412,654  | Flagler 2016      | 3,095,175 | <u>Lake 2016</u> | 7,680,098  | Pinellas 2016     | 1,699,415   |
| Baker 2016        | 4,067,558  | Franklin 2016     | 3,453,638 | Lee 2016         | 5,296,903  | Polk 2016         | 13,176,262  |
| Bay 2016          | 4,884,410  | Gadsden 2016      | 3,250,057 | Leon 2016        | 4,781,970  | Putnam 2016       | 5,581,810   |
| Bradford 2016     | 1,904,923  | Gilchrist 2016    | 2,119,279 | Levy 2016        | 7,213,457  | SantaRosa 2016    | 6,660,772   |
| Brevard 2016      | 6,739,281  | Glades 2016       | 6,585,205 | Liberty 2016     | 5,782,353  | Sarasota 2016     | 3,744,114   |
| Broward 2016      | 8,119,784  | Gulf 2016         | 3,811,593 | Madison 2016     | 4,652,583  | Seminole 2016     | 2,138,572   |
| Calhoun 2016      | 3,745,912  | Hamilton 2016     | 3,159,156 | Manatee 2016     | 5,137,157  | StJohns 2016      | 4,269,590   |
| Charlotte 2016    | 4,579,016  | Hardee 2016       | 4,290,226 | Marion 2016      | 10,940,972 | StLucie 2016      | 4,045,293   |
| Citrus 2016       | 4,072,675  | Hendry 2016       | 8,068,759 | Martin 2016      | 4,522,732  | Sumter 2016       | 3,996,311   |
| Clay 2016         | 4,220,191  | Hernando 2016     | 2,960,137 | Monroe 2016      | 6,500,986  | Suwannee 2016     | 4,737,152   |
| Collier 2016      | 13,295,931 | Highlands 2016    | 7,724,319 | Nassau 2016      | 4,278,366  | Taylor 2016       | 6,872,260   |
| Columbia 2016     | 5,041,943  | Hillsborough 2016 | 7,028,943 | Okaloosa 2016    | 6,237,374  | <u>Union 2016</u> | 1,512,231   |
| Dade 2016         | 13,269,070 | Holmes 2016       | 3,039,802 | Okeechobee 2016  | 6,017,924  | Volusia 2016      | 8,200,274   |
| DeSoto 2016       | 4,243,340  | IndianRiver 2016  | 3,094,198 | Orange 2016      | 6,607,163  | Wakulla 2016      | 4,093,822   |
| Dixie 2016        | 4,600,821  | Jackson 2016      | 6,180,173 | Osceola 2016     | 10,313,256 | Walton 2016       | 6,811,257   |
| <u>Duval 2016</u> | 5,627,581  | Jefferson 2016    | 4,040,008 | PalmBeach 2016   | 14,322,146 | Washington 2016   | 4,097,488   |
| Escambia 2016     | 4,435,212  | Lafayette 2016    | 3,283,216 | Pasco 2016       | 5,006,794  | Florida 2016      | 376,950,271 |

The above files are compressed, tab-delimited tables of numeric data that are generally software independent. Tab-delimited data can be imported into a variety of GIS, database or spreadsheet software packages.

Each compressed County archive file contains one data file having tab delimited columns of data, which include the following fields:

#### Column Definition

- Date of data representation (Year Month Day as yyyymmdd)
- 2 Latitude of Pixel value (Decimal degrees)
- 3 Longitude of Pixel value (Decimal degrees)
- 4 Pixel ID number
- 5 Potential ET(mm/day)
- 6 Reference ET(mm/day)
- 7 Solar Radiation Daily Insolation (MegaJoules/sq meter/day)
- 8 Maximum Relative Humidity for day (%)
- 9 Minimum Relative Humidity for day (%)
- 10 Maximum Temperature for day (C)
- 11 Minimum Temperature for day (C)
- 12 Wind Speed (meters/second)

#### Notes concerning the current data sets:

Missing values are represented by the number -9999.900.

#### Daily data quality codes for January 1 through December 31

| Daily Quality Codes | Size |  |
|---------------------|------|--|
| Quality Codes 2016  | 905  |  |

Each compressed Daily Quality Code file contains one data file having tab delimited columns of data, which include the following fields:

| Column | Definition |
|--------|------------|
|        |            |

- 1 Date (Year Month Day as yyyymmdd))
- 2 Quality Code (A value of 1, 2, 3, or 4)

Based on the quality of Solar data for that day

( 1 = Good quality, 2 = Usable data, 3 = Uncertain or unverifiable quality, 4 = Unusable or missing)

Information regarding the methodology used in the ET computations are detailed in the InTechWeb Report:

Use of Visible Geostationary Operational Meteorological Satellite Imagery in Mapping Reference and Potential Evapotranspiration over Florida

**≈**0505

Metadata file describing this year's GOES ET process (XML format): GOES ET metadata 2016

Preliminary Information-Subject to Revision. Not for Citation or Distribution.



Statewide Evapotranspiration Data - (2km Daily)

#### Statewide Evapotranspiration Data - (2km Daily)

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| Columbia 2016     | 5,041,943  | Hillsborough 2016 | 7,028,943 | Okaloosa 2016    | 6,237,374  | <u>Union 2016</u> | 1,512,231   |
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| DeSoto 2016       | 4,243,340  | IndianRiver 2016  | 3,094,198 | Orange 2016      | 6,607,163  | Wakulla 2016      | 4,093,822   |
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12 wind Speed (meters/second)

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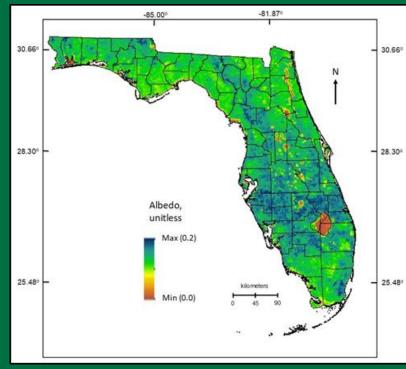
Metadata file describing this year's GOES ET process (XML format): GOES ET metadata 2016

Preliminary Information-Subject to Revision. Not for Citation or Distribution.



## Alternative methodologies

- "Blue sky" albedo from Moderate Resolution Imaging Spectroradiometer (MODIS) on the NASA Terra satellite
  - 5 year averages of daily values
  - <u>2011-15</u>
  - Product unavailable in 2017
- PET calculations





## Alternative methodologies, continued

- Meteorological data from:
  - 1. FAWN, DBHYDRO and NOAA weather station data
  - 2. North American Regional Reanalysis
  - 3. Weather Research and Forecasting Model



# Weather Research and Forecasting Model

- 2 km spatial resolution
  - Grid aligns with PET, RET, NEXRAD grids
- 1 hour temporal resolution

Preliminary Information-Subject to Revision. Not for Citation or Distribution.



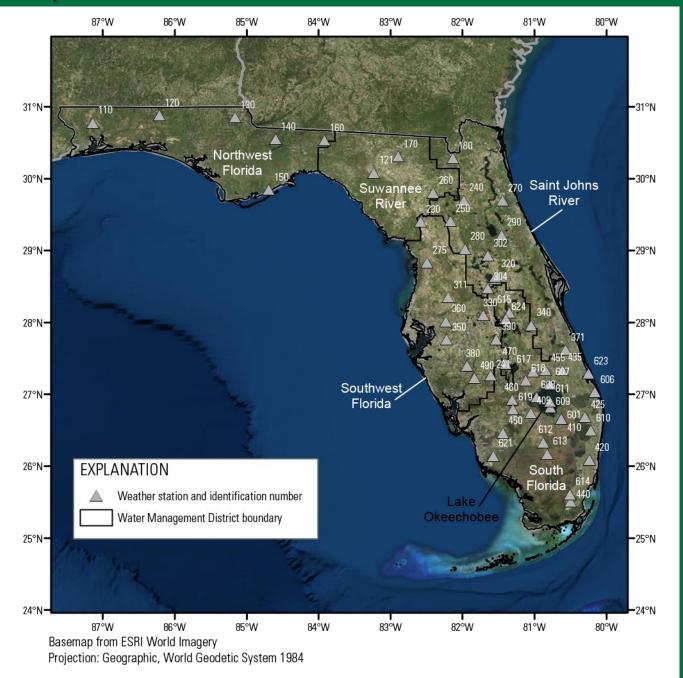


# Bias analysis at weather station locations

- Bias in NARR and WRF data as alternative sources of meteorological data
  - Calendar year 2017
  - 57 weather stations
    - contained sufficient data to compute RET
    - wind was limiting variable that excluded stations



#### Preliminary Information-Subject to Revision. Not for Citation or Distribution.





Preliminary Information-Subject to Revision. Not for Citation or Distribution.





### Cross plots for FAWN 110: model (NARR and WRF) on x-axis, observed on y-axis

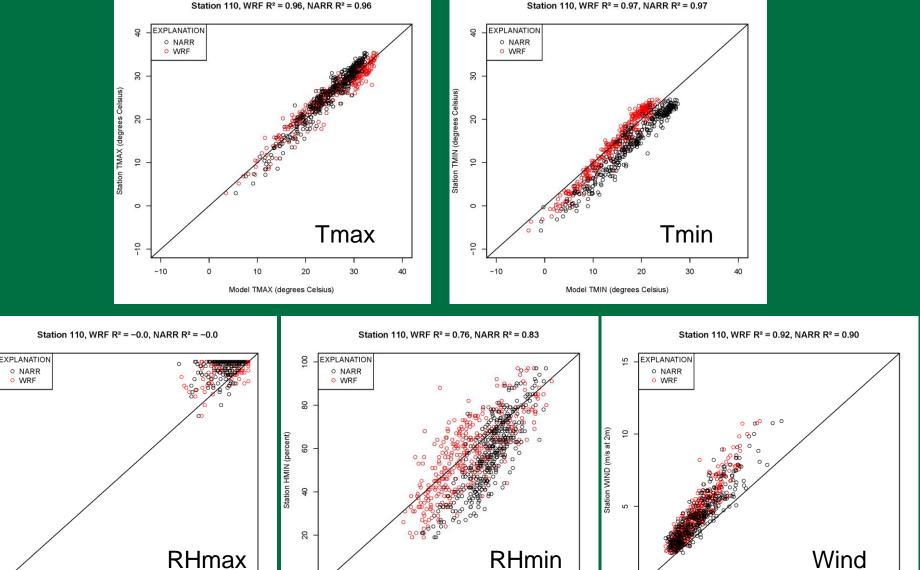
O NARR

20

Model HMAX (percent)

100

WRF



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Model HMIN (percent)

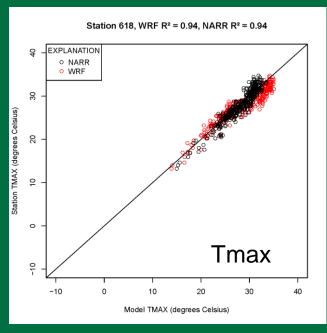
100

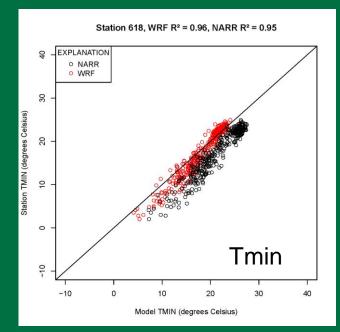
10

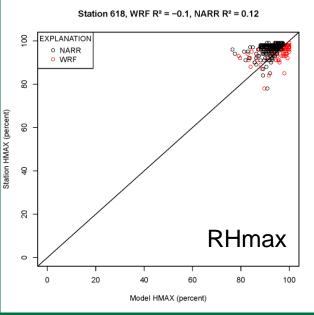
Model WIND (m/s at 2m)

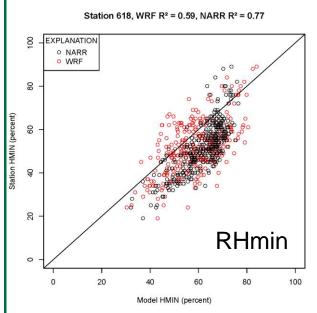
20

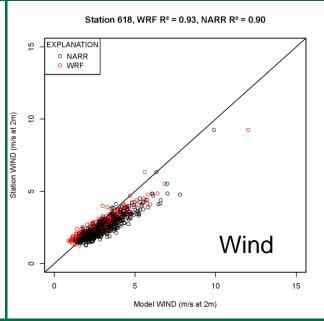
### Cross plots for DBHYDRO 618



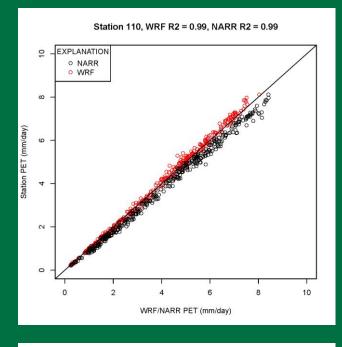


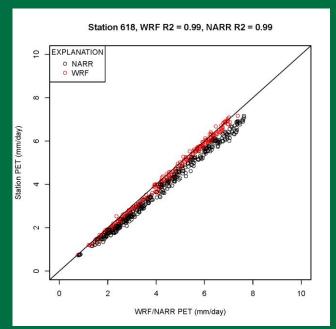


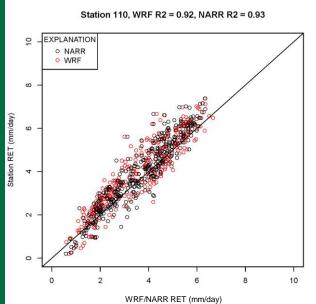


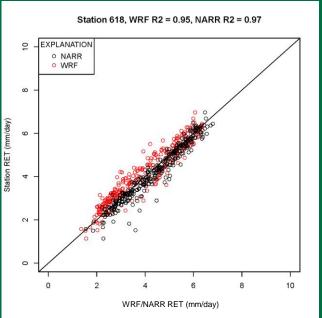


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RET

PET

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## Bias in NARR ET at station locations

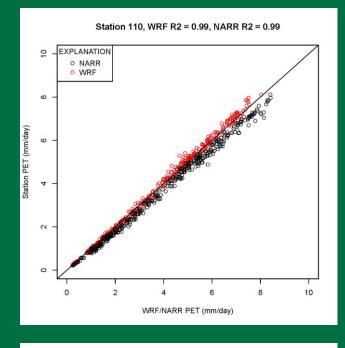
|             | Evapotranspira | ation based on | NARR evano     | NARR evapotranspiration |           | as        | After bias correction |           |  |
|-------------|----------------|----------------|----------------|-------------------------|-----------|-----------|-----------------------|-----------|--|
| Time Period | weather statio | n observations | Tivilit e tapo |                         |           |           |                       |           |  |
| (2017)      | Potential      | Reference      | Potential      | Reference               | Potential | Reference | Potential             | Reference |  |
|             | (millin        | neters)        | (millin        | neters)                 | (millin   | neters)   | (millin               | neters)   |  |
| Annual      | 1547           | 1568           | 1661           | 1535                    | 114       | -33       | -2                    | -1        |  |
| January     | 56             | 87             | 64             | 84                      | 8         | -3        | 1                     | -4        |  |
| February    | 77             | 98             | 86             | 92                      | 9         | -6        | 1                     | -5        |  |
| March       | 118            | 139            | 130            | 130                     | 12        | -9        | 2                     | -6        |  |
| April       | 155            | 165            | 166            | 155                     | 11        | -9        | 1                     | -5        |  |
| May         | 192            | 190            | 206            | 180                     | 13        | -10       | 1                     | -3        |  |
| June        | 171            | 144            | 180            | 145                     | 9         | 1         | -2                    | 5         |  |
| July        | 195            | 164            | 206            | 165                     | 10        | 1         | -2                    | 6         |  |
| August      | 191            | 161            | 201            | 164                     | 10        | 3         | -2                    | 8         |  |
| September   | 156            | 139            | 162            | 138                     | 7         | -1        | -4                    | 3         |  |
| October     | 113            | 119            | 120            | 121                     | 7         | 2         | -2                    | 4         |  |
| November    | 72             | 88             | 81             | 88                      | 8         | 1         | 1                     | 1         |  |
| December    | 51             | 74             | 60             | 71                      | 8         | -2        | 1                     | -4        |  |
|             |                |                |                |                         |           |           |                       |           |  |

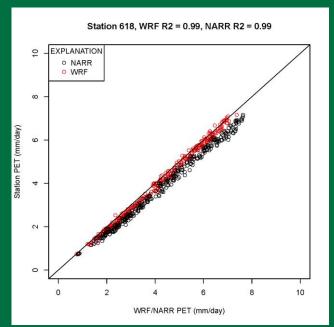


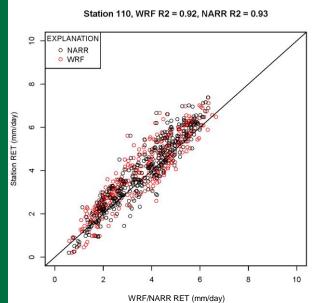
## Bias in WRF ET at station locations

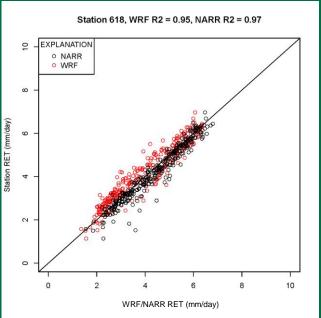
| Time Period | Evapotranspiration based on weather station observations |           | WRF evapotranspiration |           | Ві            | as        | After bias correction |           |
|-------------|--|-----------|------------------------|-----------|---------------|-----------|-----------------------|-----------|
| (2017)      | Potential  | Reference | Potential              | Reference | Potential     | Reference | Potential             | Reference |
|             | (millimeters)  |           | (millimeters)          |           | (millimeters) |           | (millimeters)         |           |
| Annual      | 1547   | 1568      | 1552                   | 1434      | 5             | -133      | -2                    | -1        |
| January     | 56   | 87        | 59                     | 71        | 3             | -16       | 1                     | -4        |
| February    | 77   | 98        | 80                     | 80        | 3             | -18       | 2                     | -7        |
| March       | 118  | 139       | 121                    | 116       | 3             | -24       | 3                     | -12       |
| April       | 155  | 165       | 158                    | 142       | 3             | -23       | 3                     | -13       |
| May         | 192  | 190       | 193                    | 170       | 1             | -20       | 2                     | -9        |
| June        | 171  | 144       | 170                    | 146       | -1            | 2         | -1                    | 12        |
| July        | 195  | 164       | 192                    | 164       | -3            | 0         | -2                    | 11        |
| August      | 191  | 161       | 188                    | 160       | -3            | -1        | -2                    | 10        |
| September   | 156  | 139       | 151                    | 134       | -5            | -5        | -5                    | 6         |
| October     | 113  | 119       | 112                    | 111       | -1            | -8        | -2                    | 3         |
| November    | 72   | 88        | 74                     | 76        | 1             | -11       | -1                    | 0         |
| December    | 51   | 74        | 53                     | 64        | 2             | -10       | 0                     | 2         |
|             |  |           |                        |           |               |           |                       |           |











RET

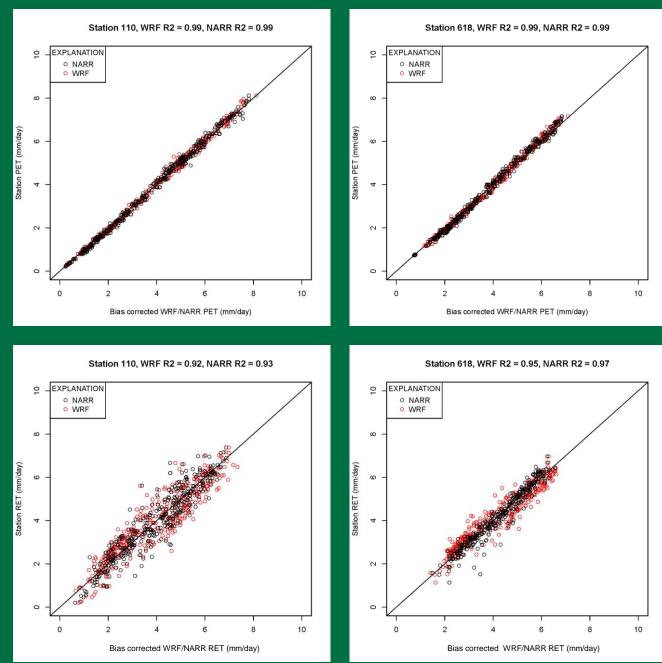
PET

Preliminary Information-Subject to Revision. Not for Citation or Distribution.

### Cross plots for bias corrected PET and RET

PET

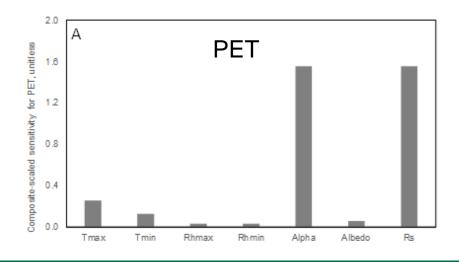
RET

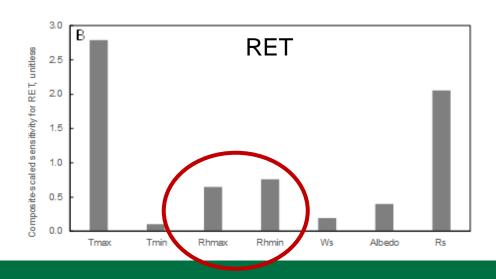


Preliminary Information-Subject to Revision. Not for Citation or Distribution.

# Sensitivity of PET and RET to input meteorological variables

- Based on methodology of Hill (1998)
- Dimensionless index



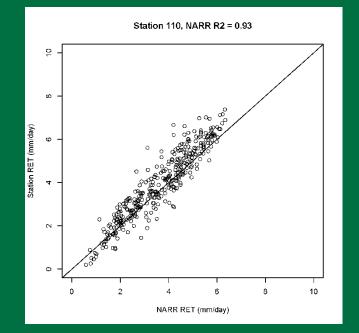


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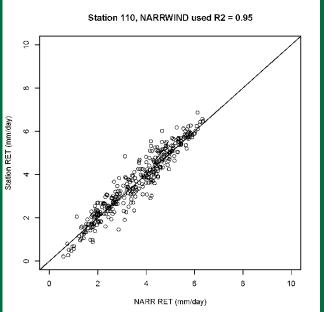
Hill, M.C., 1998, Methods and guidelines for effective model calibration: U.S. Geological Survey Water-Resources Investigations Report 98-4005, 90 p.

### Cross plots for PET and RET for FAWN 110 using NARR meteorological data



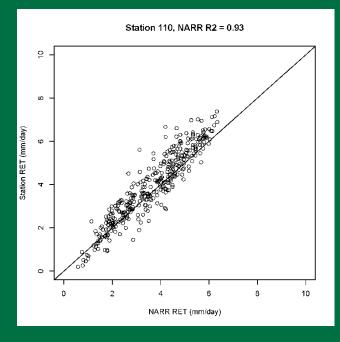
**Effect** 

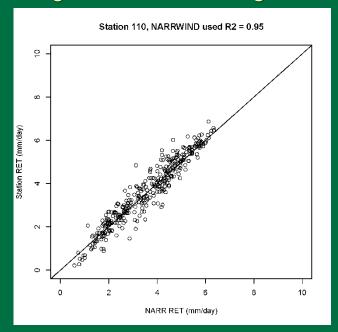
of Wind



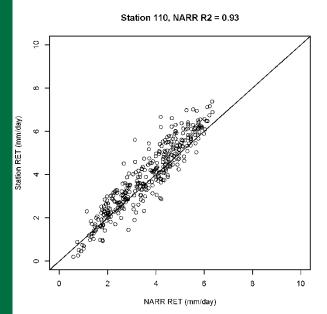
### Cross plots for PET and RET for FAWN 110 using NARR meteorological data

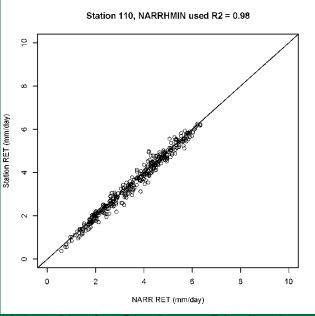
Effect of Wind





Effect of RHmin



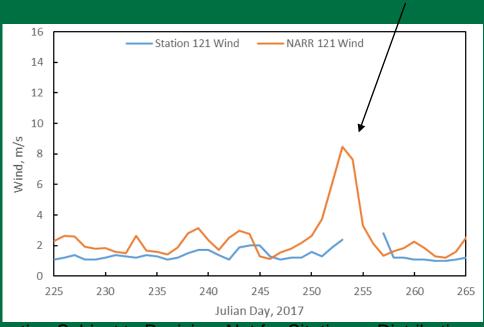


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## Conclusions

- MODIS data improves spatial representation of albedo
- Radial basis functions improves spatial interpolation
- NARR and WRF output are an alternative to weather station data for computing PET
  - can be used to fill missing data

No data at station during Hurricane Irma





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## Conclusions

- Bias in NARR removed by linear model
  - Linear bias correction does not fully remove WRF bias
- Remaining bias primarily due to variability in minimum relative humidity
  - Random scatter, so might be difficult to correct
  - How accurate are RH sensors?
  - How accurate is RH of weather models?
  - Is variability due to sub-grid scale variability?





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