

# Development of Future Depth-Duration-Frequency Curves (DDF) for the SFWMD

#### Michelle M. Irizarry-Ortiz (mirizarry-ortiz@usgs.gov) U.S. Geological Survey Caribbean-Florida Water Science Center

Joint Cooperative USGS-SFWMD Project in collaboration with the Sea Level Solutions Center at Florida International University. Key project staff: John F. Stamm – USGS Ana Carolina Maran – SFWMD Anupama John – FIU Jayantha Obeysekera – FIU

U.S. Department of the Interior U.S. Geological Survey

This information is preliminary and is subject to revision. It is being provided to meet the need for timely best science. The information is provided on the condition that neither the U.S. Geological Survey nor the U.S. Government shall be held liable for any damages resulting from the authorized or unauthorized use of the information

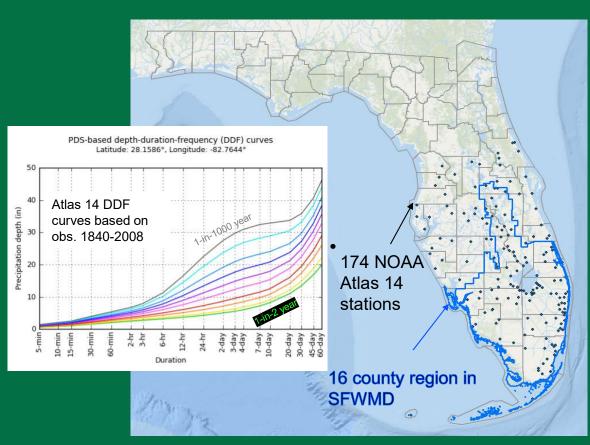
## Outline

- Project Description and Scope
- Brief Review of Datasets and Methods
  - Change Factors
  - Downscaled Datasets
  - Technical Approach
  - Model Culling
- Results
  - Historical Bias
  - Model Culling
  - Summarizing Change Factors



### Project Description and Scope

- Ensemble of downscaled climate models used to determine median change factors of future (projected) rainfall depths at NOAA Atlas 14 stations located in SFWMD. To be used in evaluating flood protection level of service in SFWMD region.
  - 40-year hist. period: 1966-2005
  - 40-year future period: 2050-2089
  - Durations of rainfall event: 1, 3, and 7 days
  - Return periods (ave. # of years between events of the given rainfall depth): 5, 10, 25, 50, 100\*\*, and 200\*\* years



**USGS** \*\*Low confidence in long return periods

#### Change Factor

 Ratio of future (f) rainfall depth (D in inches) from model to historical (h) rainfall depth from model for a given duration (d in days) and return period (T in years).

 $CF = D_f(d,T) / D_h(d,T)$ 

Data from model grid cells for two 40-year periods

Future DDF = 
$$D_{Atlas14}(d,T) * CF$$

Atlas 14 DDF curves use obs. 1840-2008



#### **Downscaled Climate Datasets**

Statistical methods differ in observational datasets used; how many analog days\*\* are considered, averaged, weighted; how biascorrection is done; how tail (extreme values) are bias-corrected; whether actual values or anomalies are used; whether trends in GCM data are preserved, etc.

\*\*Analog days are days in an observational dataset that best match coarse-scale meteorological field(s) for the day that is being downscaled.

Dynamically-downscaled datasets (25-50 km) are "physically-based" in terms of solving equations of hydrodynamics and thermodynamics; however, convection is parameterized based on empirical equations. <u>To actually simulate convection in</u> <u>momentum equation requires model resolution</u> < <u>2-km.</u> Sea-breeze, lake-breeze, coastal curvature enhanced convection not adequately captured at coarse RCM resolutions.  Both methods depend on regional climate, tropical storms, and remote
 teleconnection patterns that drive precipitation in south Florida being adequately captured in the parent GCM.





#### Downscaled Climate Datasets

- LOCA: Statistical. Localized Constructed Analogues product by University of California at San Diego.
  - 1/16<sup>th</sup> degree, ~6 km
  - 30 historical runs, 30 RCP4.5, 30 RCP8.5
- MACA: Statistical. Multivariate Adapative Constructed Analogs.
  - Livneh training data: 1/16<sup>th</sup> degree, ~6 km
  - gridMET training data: 1/24<sup>th</sup> degree, ~3 km
  - 20 GCMs for each training dataset in historical period, RCP4.5 and RCP8.5.
- CORDEX: Dynamical. North American
  Coordinated Regional Downscaling Experiment
  - 0.22 to 0.44 degree, ~25 to ~50 km (parameterized convection)
  - bias corrected by DayMet and gridMET
  - 54 historical runs, 14 RCP4.5, 54 RCP8.5.

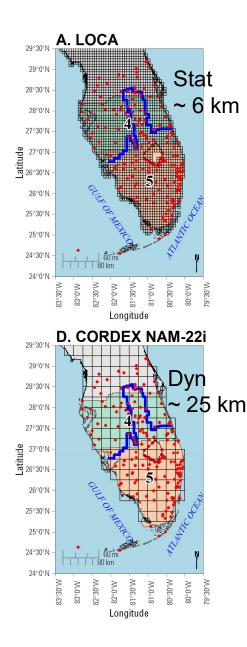


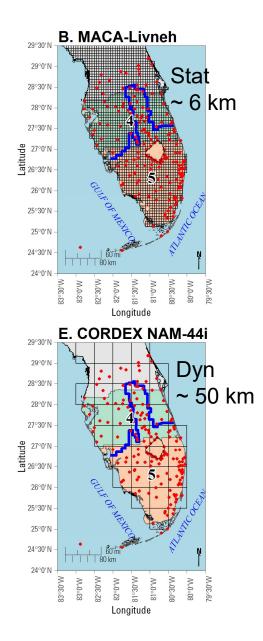
JupiterWRF: Hybrid

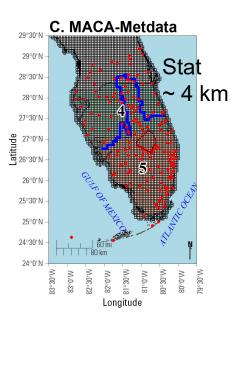
- Dynamical model: Weather Research and Forecasting (WRF) model developed by Jupiter Intelligence.
  - 4 km resolution.
- Hybrid approach: intensity scaling x analog resampling
- Based on CMIP5 and CMIP6 GCM output (7 GCM/RCP\_SSP combinations).
- Model datasets developed by Jupiter Intelligence, coordinated by FIU.
- Only extreme events of 1-day duration can be evaluated.

RCP4.5: Medium-low emission scenario with 4.5  $W/m^2$  increase in radiative forcing by 2100 with respect to pre-industrial era RCP8.5: High emission scenario with 8.5  $W/m^2$  increase in radiative forcing by 2100 with respect to pre-industrial era

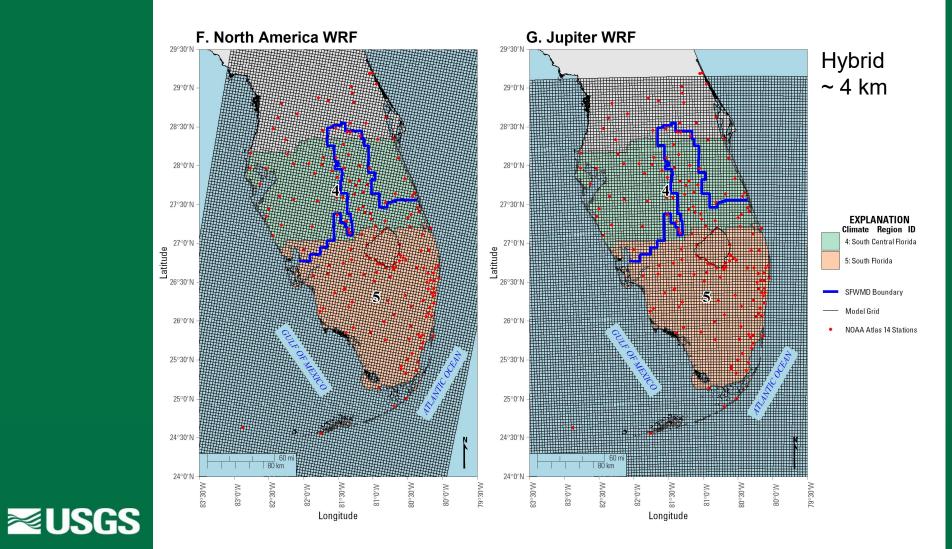












### Technical Approach

- Peak over threshold (POT): Model exceedances above a sufficiently high threshold. Uses more data in the tail of the distribution than classical annual maxima approach.
- Grid cell values represent areal averages. Fitting DDF curves at each model grid cell at its native resolution and applying areal reduction factors for conversion to station values
  - Allows comparison of station-scale DDF curves fit for historical period against NOAA Atlas 14 PDS-based official DDF curves.
  - Assumption constancy of area reduction factors from current to future. The ARF would cancel out in the computation of change factors.

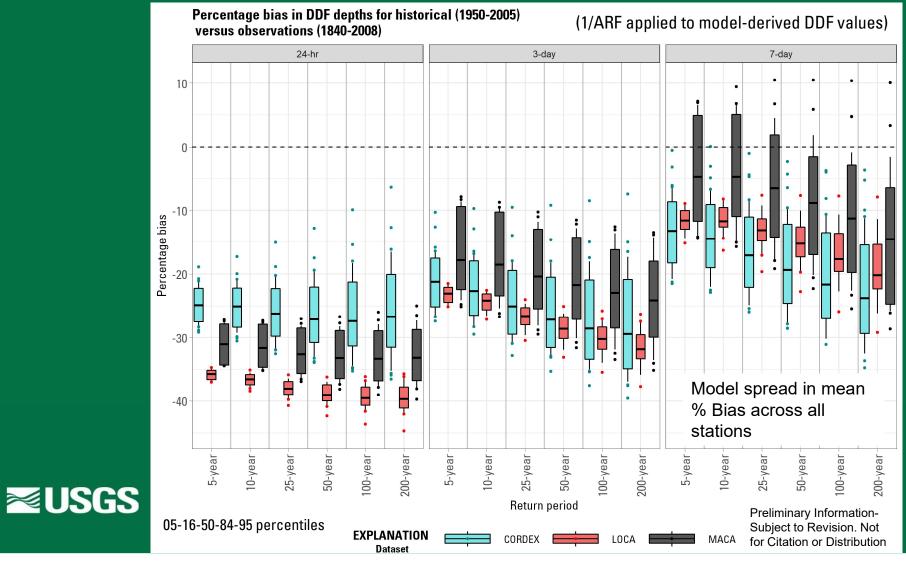


### Summarizing Change Factors

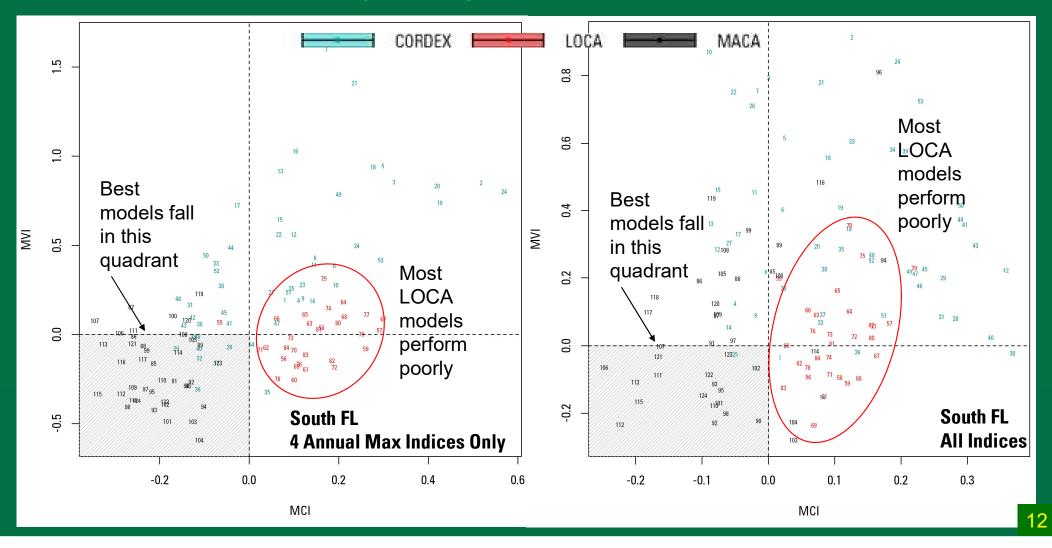
- Median and model spread in change factors at 174 NOAA Atlas 14 stations in south and central Florida for durations and return periods of interest.
- Model selection criteria performance metrics based on precipitation climate extremes indices from Expert Team on Climate Change Detection Indices (ETCCDI).
  - Two observational datasets: PRISM and SFWMD Super-grid for the period 1981-2005.
  - 4 indices for annual maxima of various durations (1, 3, 5, 7 days) used in model culling for central FL and south FL regions. 11 additional indices used to inform overall dataset performance.
  - Evaluated based on how well the models reproduce the observed climatology and interannual variability of climate extreme indices.

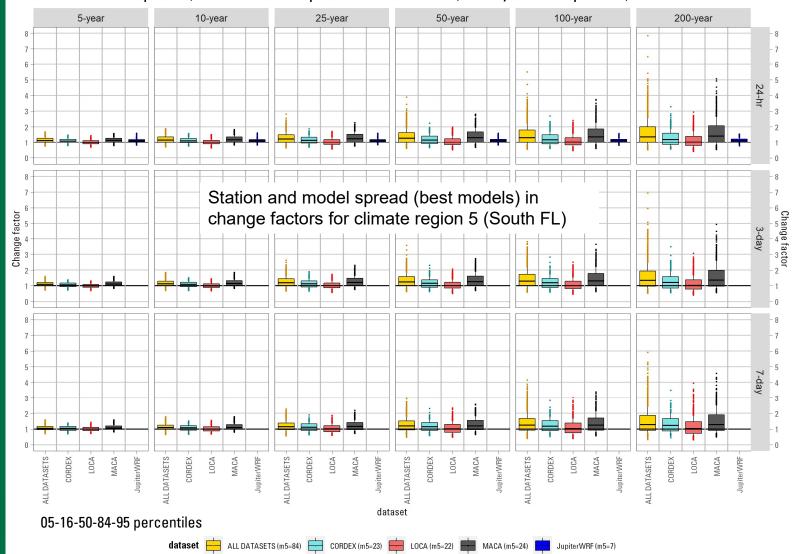


#### Bias Results



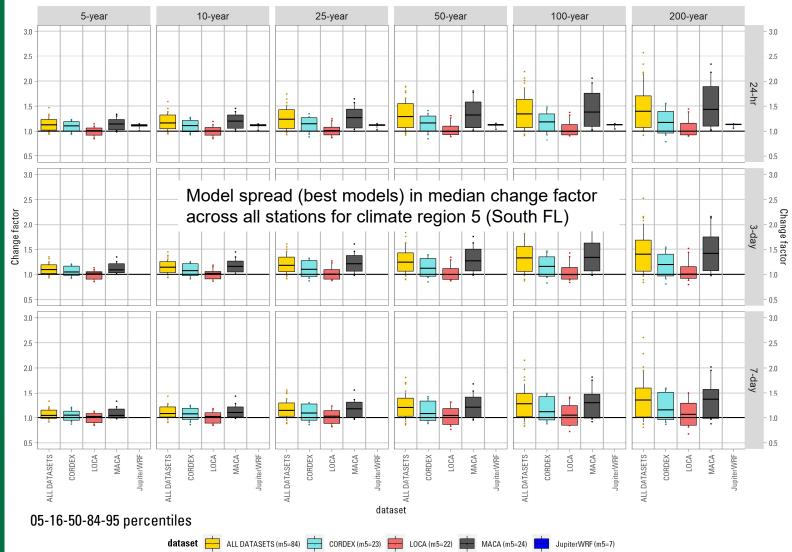
#### "Best" Models





#### Change factors across datasets for all stations in Climate Region 5 (best models, all RCPs) Box includes 16-50-84th percentile, whiskers are from 5-95th percentile. Points show outliers (values beyond 5th & 95th percentiles).

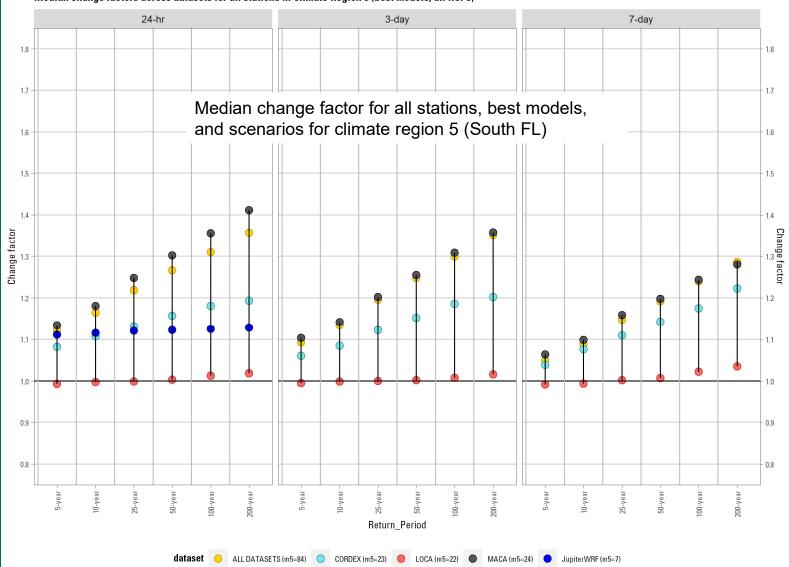




#### Median station change factors across models in Climate Region 5 (best models, all RCPs) Box includes 16-50-84th percentile, whiskers are from 5-95th percentile. Points show outliers (values beyond 5th & 95th percentiles).



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



#### Median change factors across datasets for all stations in Climate Region 5 (best models, all RCPs)

**USGS** 

# Summary

- Uncertain CFs with no consensus across datasets (except most agree with CFs > 1; that is, an increase in precipitation extremes in the future). Selection of CFs should be based on risk-based approach. Median CFs of 1-1.6 may be adequate for low-risk situations, but CF values of 2-3 or even higher may be desirable in designing critical infrastructure.
- Highest and lowest CFs obtained from statistical downscaling datasets (MACA and LOCA, respectively).
- Lowest CFs from LOCA yet LOCA performs the worst in terms of reproducing historical climate extreme indices. Low daily CFs obtained from Jupiter WRF hybrid dataset. Intermediate CFs from CORDEX.
- To reduce some of the uncertainties in downscaling, a high-resolution regional climate model that can simulate convection in momentum equation (< 2-km) should be developed for the state of Florida.

