



# Extreme Weather Research Overview & Coordination: On-going Research Projects

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U.S. Environmental Protection Agency (EPA)  
Office of Research and Development (ORD)

Chesapeake Bay Program (CBP):  
Joint Meeting of the Urban Stormwater Workgroup, Modeling Workgroup, and Climate Resiliency Workgroup  
December 9, 2020

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# National Stormwater Calculator (SWC) Weather Data Updates

## Existing historical weather data:

- EPA's Better Assessment Science Integrating Point and Nonpoint Sources (BASINS) historical weather data. Goes back ~30 years (2006-2009)

## Existing extreme weather and climate change data:

- Extreme weather data from Climate Resilience Evaluation and Awareness Tool (CREAT) 2.0 from 2013
  - Near and far time period: 2020 – 2049 and 2045 – 2074
  - Three future scenarios: hot/dry, medium, and warm/wet
  - 24-hour annual maximum daily rainfall: 5-, 10-, 15-, 30-, 50-, and 100-year storm events
  - NRCS (SCS) 1986 storm type rainfall distribution method

\*SWC web app website: <https://swcweb.epa.gov/stormwatercalculator/>



# SWC Web App: Existing Historical Weather and Climate Change Modules

EPA National Stormwater Calculator

NEW SAVE OPEN

**Precipitation/Evaporation**

Directions

Rain Gage:

ANNAPOLIS POLICE BRKS

Weather Station:

ANNAPOLIS POLICE BRKS

**Rainfall and Evaporation Information:**

Record Start Date: 1970/01/01  
Record End Date: 2005/12/31  
Annual Rainfall: 46

[Download rainfall/evaporation data](#)

[Help](#)

EPA National Stormwater Calculator

NEW SAVE OPEN RESOURCES CONTACT

**Climate Change**

Directions

Helpful Resources

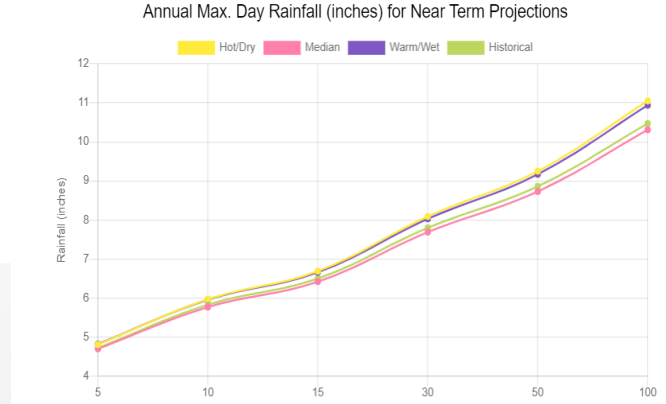
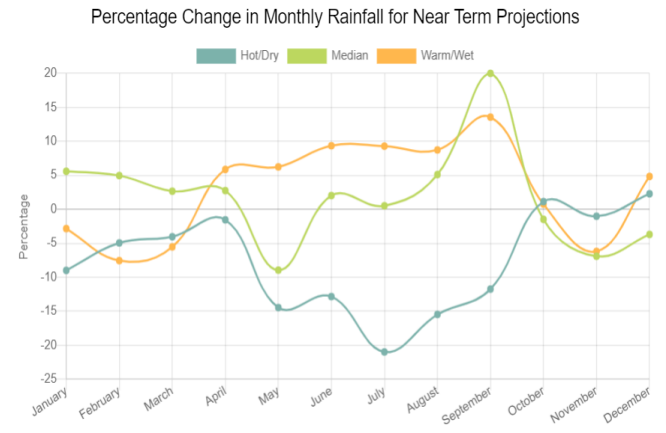
Select a future climate change scenario to apply:

- No Change
- Hot/Dry
- Median Change
- Warm/Wet

Select the time period to which the climate change scenario applies:

- Near Term (2020 - 2049)
- Far Term (2045 - 2074)

[Print Charts to PDF File](#)





# National Stormwater Calculator (SWC) Weather Data Updates

## Updating of historical weather data:

- Updating historical weather data using NOAA's Integrated Surface Database (ISD) and Hourly Precipitation Dataset
  - Focusing on station data from principal airports and National Weather Service (NWS) Cooperative Observer Program (COOP) – 3,800 stations
    - COOP stations: 1,860
    - ISD stations: 1,970
  - Stations with at least 10 years of continuous data, going from 1990 – 2019
  - ISD and COOP data to be appended to BASINS historical weather data where the stations IDs match
  - Automatically updated annually as new data recorded in ISD and COOP HPD

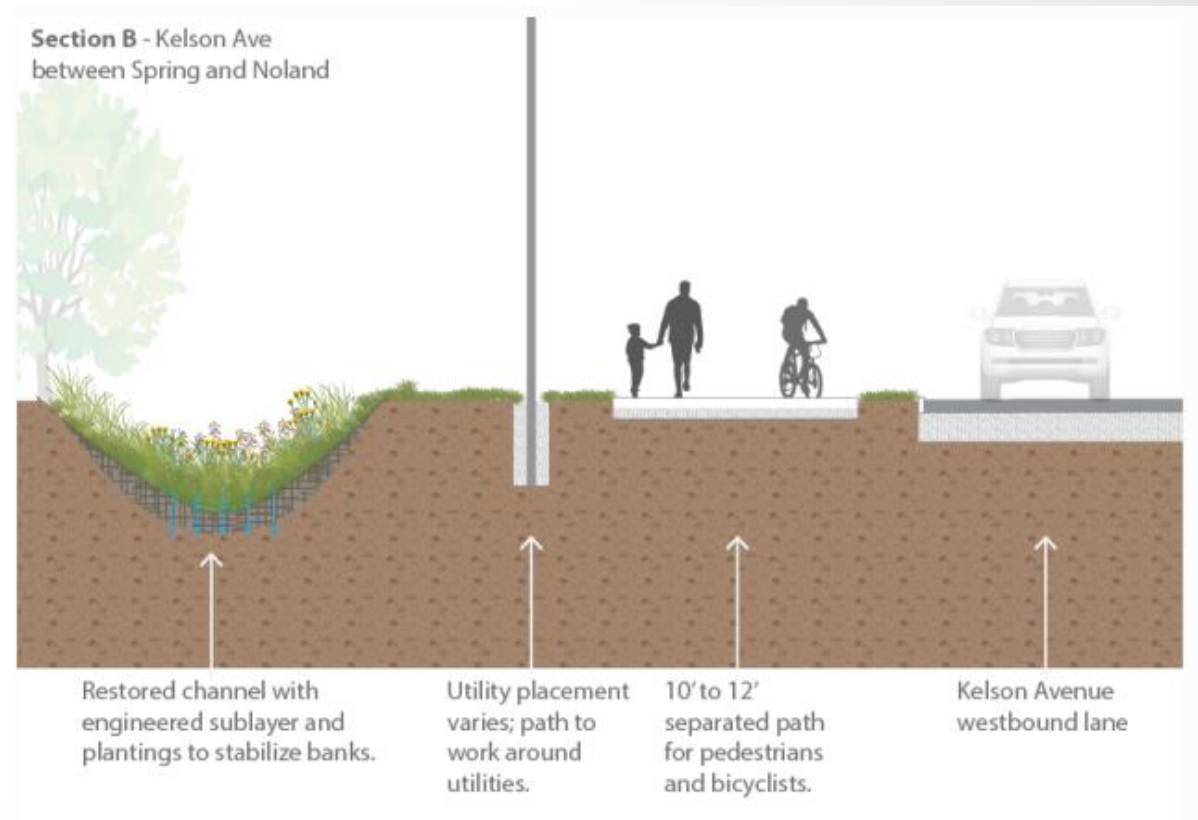
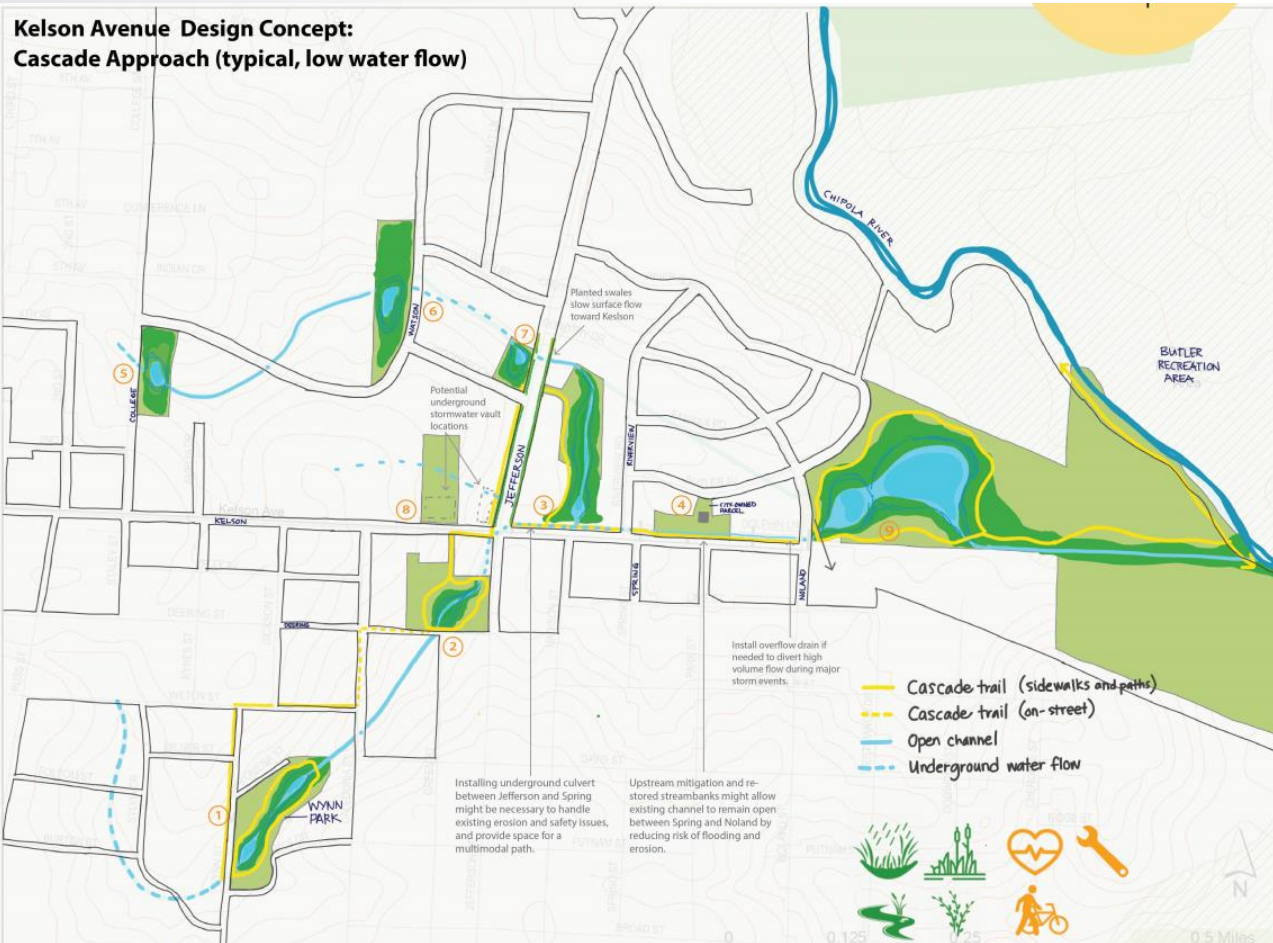
## Updating of extreme weather data:

- Climate change scenarios from EPA's CREAT 3.0 (based on CMIP 5)
- NRCS 2019 National Engineering Handbook Rainfall Distribution Methodology using NOAA Atlas 14
- Use of NOAA Atlas 14 Generalized Extreme Value (GEV) curves where available
- Compare extreme weather and climate change runoff results with other on-going ORD extreme weather projects



# SWC Extreme Weather Application: FEMA and EPA Recovery and Resiliency Partnership Project: City of Marianna, FL (2020)

EPA Region 4 provided green infrastructure design assistance to the City of Marianna, FL in 2020 as part of recovery efforts from Hurricane Michael. The SWC was used for estimating the stormwater runoff reduction and estimated costs of LID controls along a proposed greenway trail.

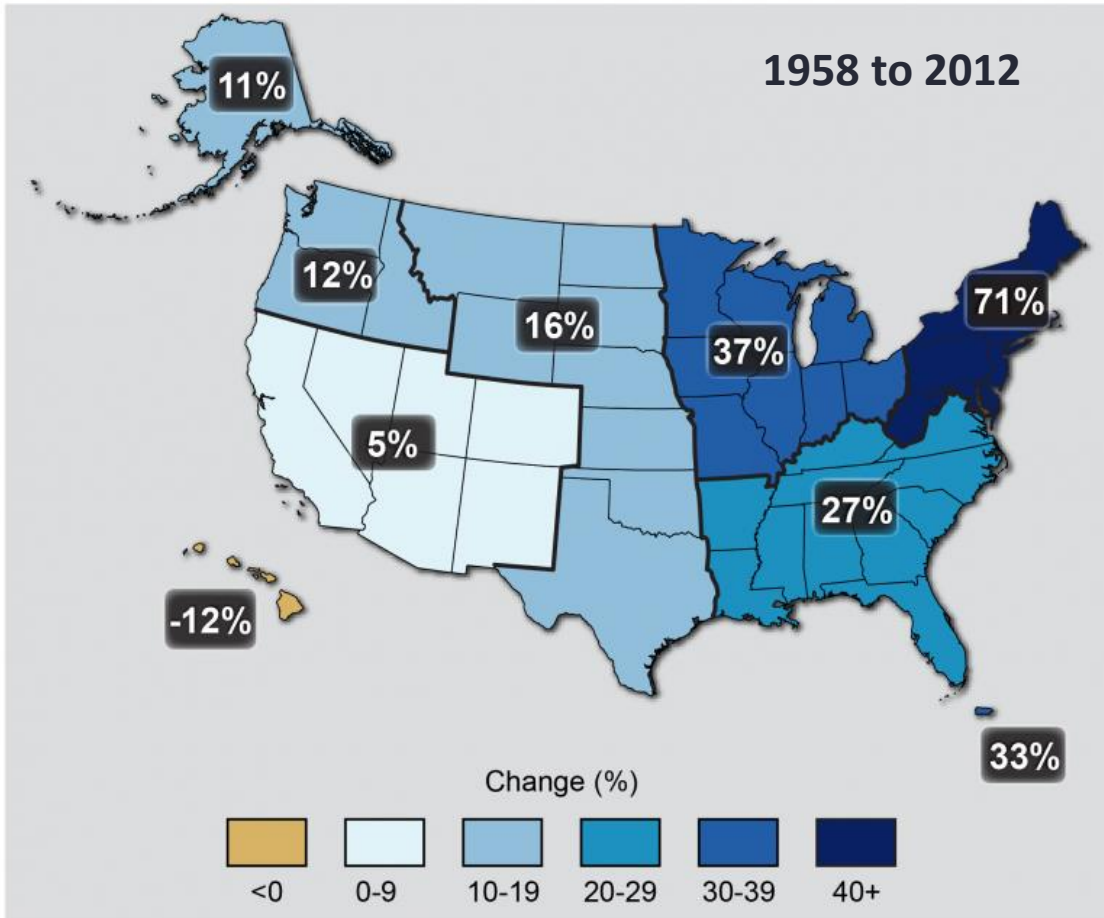


\*Project website: <http://r2p2.skeo.com/marianna/>

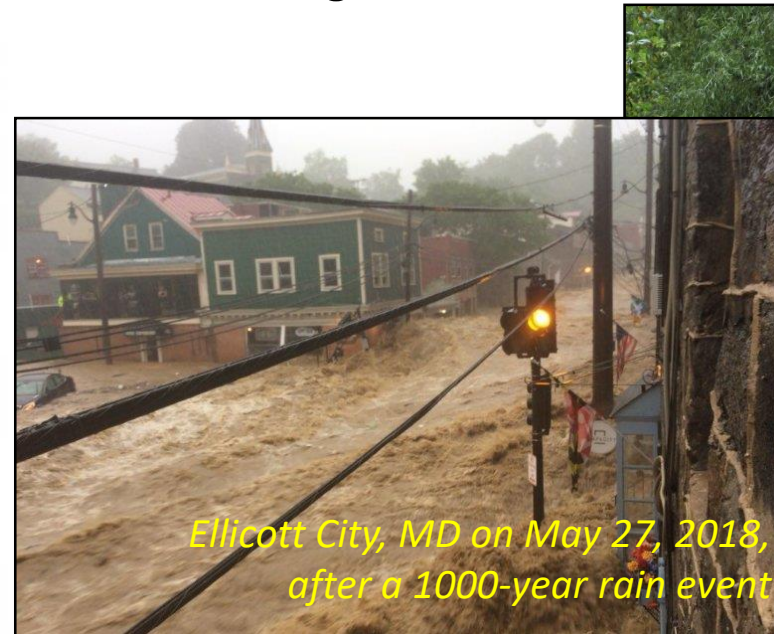


# Observed Extreme Precipitation

Observed Change in Very Heavy Precipitation



- Warming temperatures increase the capacity of the atmosphere to hold water vapor, leading to increased precipitation extremes.
- Implications for existing stormwater infrastructure, roads, levees, bridges, culverts ...



*Ellicott City, MD on May 27, 2018, after a 1000-year rain event*

*Photo credit: Twitter user Max Robinson/@DieRobinsonDie*



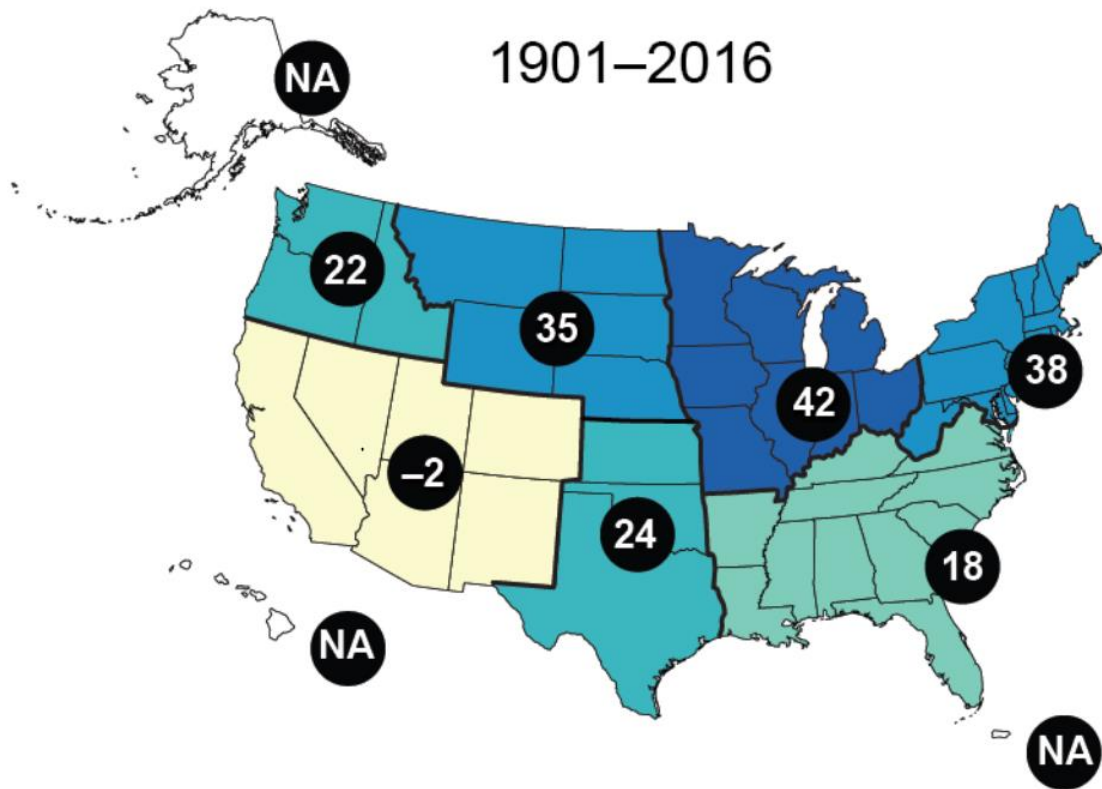
*Santee Experimental Forest, SC  
Hurricane Joaquin  
October 2015*

*Photo credit: Rick Wrenn*



# Observed Extreme Precipitation

## Change in the amount of precipitation falling in the heaviest 1% of events



- Warming temperatures increase the capacity of the atmosphere to hold water vapor, leading to increased precipitation extremes.
- Implications for existing stormwater infrastructure, roads, levees, bridges, culverts ...



*Ellicott City, MD on May 27, 2018, after a 1000-year rain event*

*Photo credit: Twitter user Max Robinson/@DieRobinsonDie*

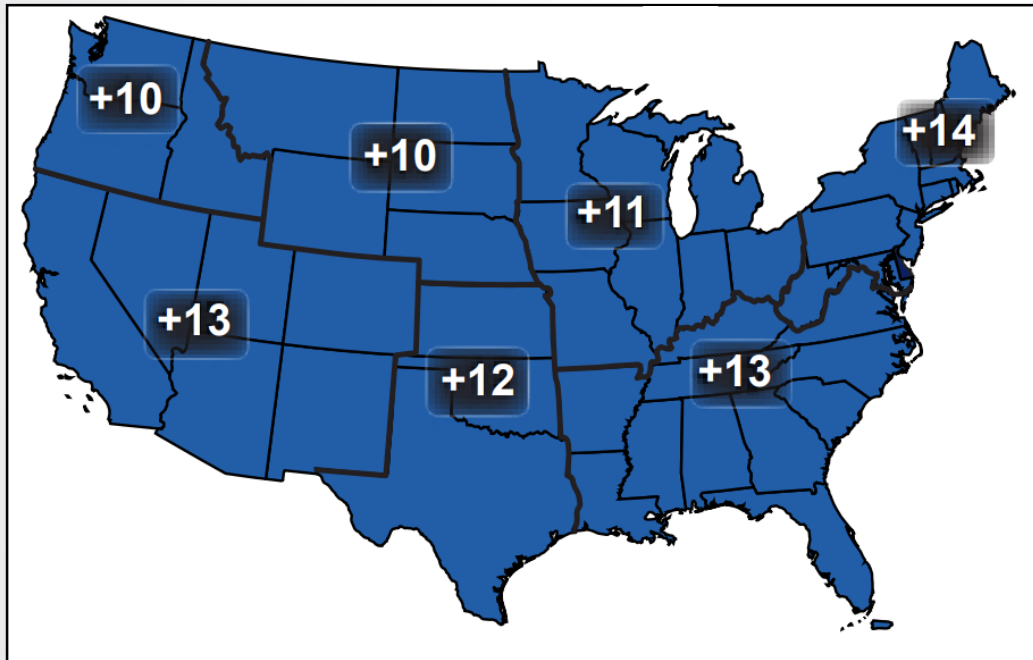


*Santee Experimental Forest, SC  
Hurricane Joaquin  
October 2015*

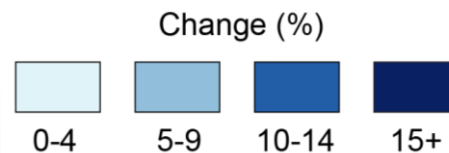
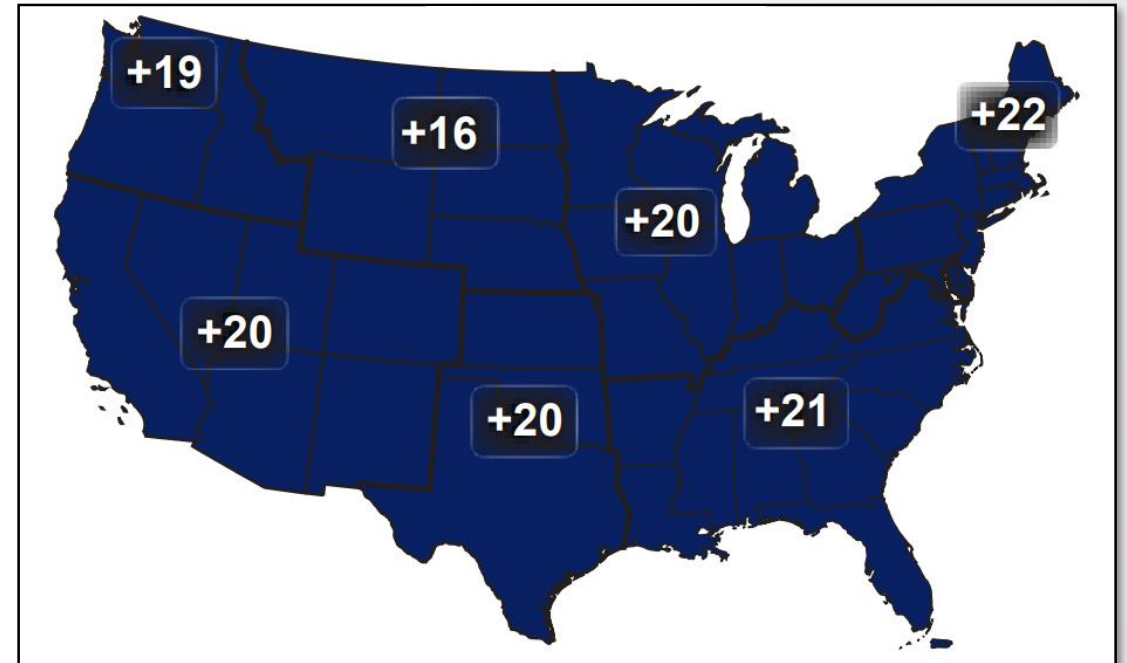
*Photo credit: Rick Wrenn*

## Projected Change in Daily, 20-year Extreme Precipitation by the end of century

Moderate increase in GHG emissions (RCP 4.5)



Higher increase in GHG emissions (RCP 8.5)







# Global and Regional Climate Models

GCM

DOWNSCALING

Statistical -> LOCA (5 km) and MACA (7 km)

Dynamical -> WRF (36 km)

Pros	Cons
<ul style="list-style-type: none"><li>• Based on standard and accepted statistical procedures</li><li>• Cheap and computationally efficient</li><li>• Easily transferable between regions</li><li>• Many scenario ensembles available</li></ul>	<ul style="list-style-type: none"><li>• <b>Stationarity assumption</b></li><li>• Require long and reliable observed historical data series for calibration</li><li>• Dependent upon choice of predictors</li><li>• Bias corrected</li></ul>

Pros	Cons
<ul style="list-style-type: none"><li>• Produces responses based on physically consistent processes</li><li>• Based on the information from GCMs resolves atmospheric processes on a smaller scales</li><li>• <b>Non-stationarity in the predictor-predictand relationship</b></li></ul>	<ul style="list-style-type: none"><li>• Computationally intensive</li><li>• Limited number of scenario ensemble members</li><li>• No bias correction</li></ul>

Both are dependent on GCM boundary forcing and affected by biases in underlying GCM e.g., cold biases, underrepresentation of ocean temps

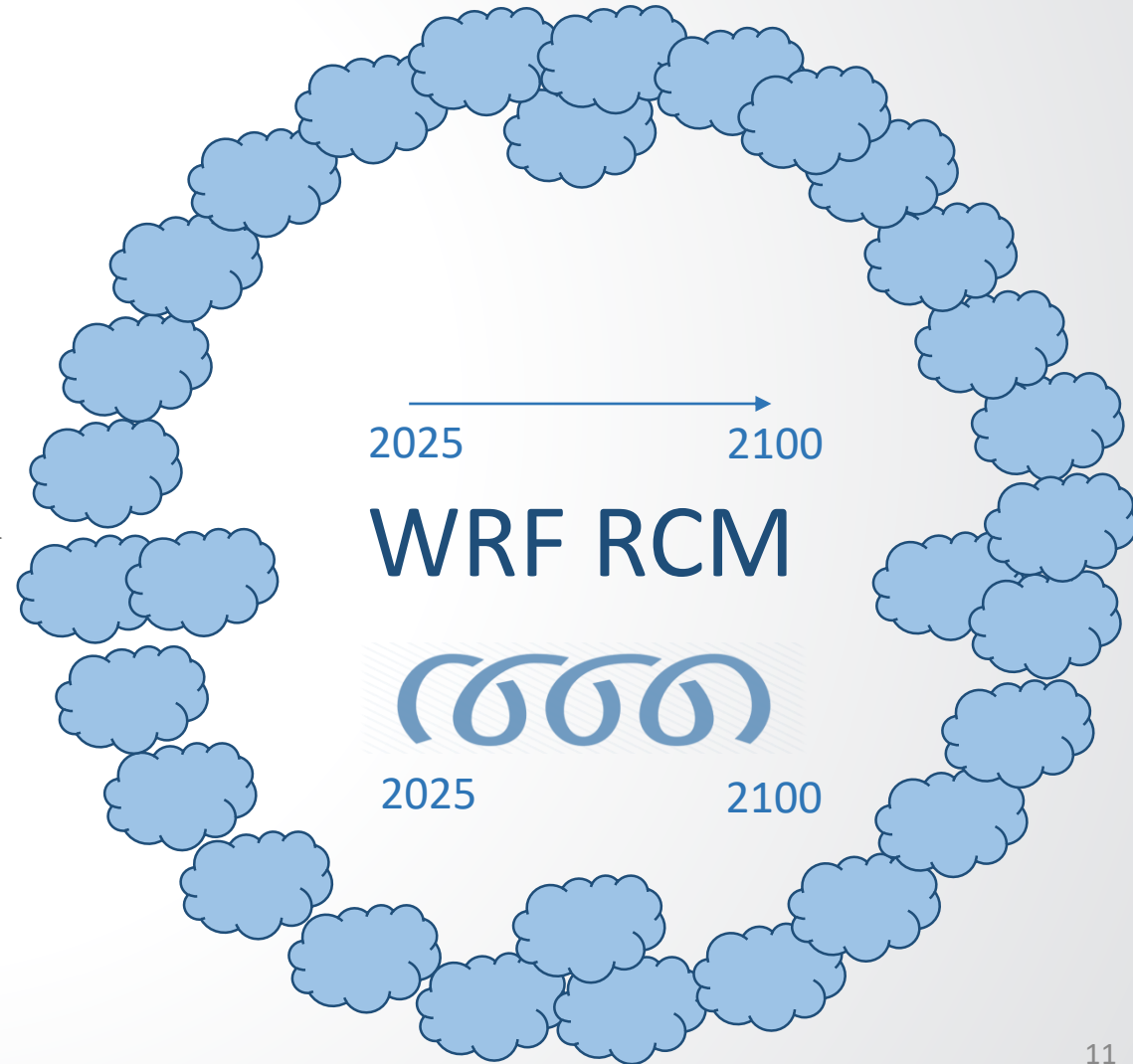
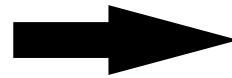
Table adapted from Brekke et al. (2009)

Weather:  
Short-term  
state of the  
atmosphere

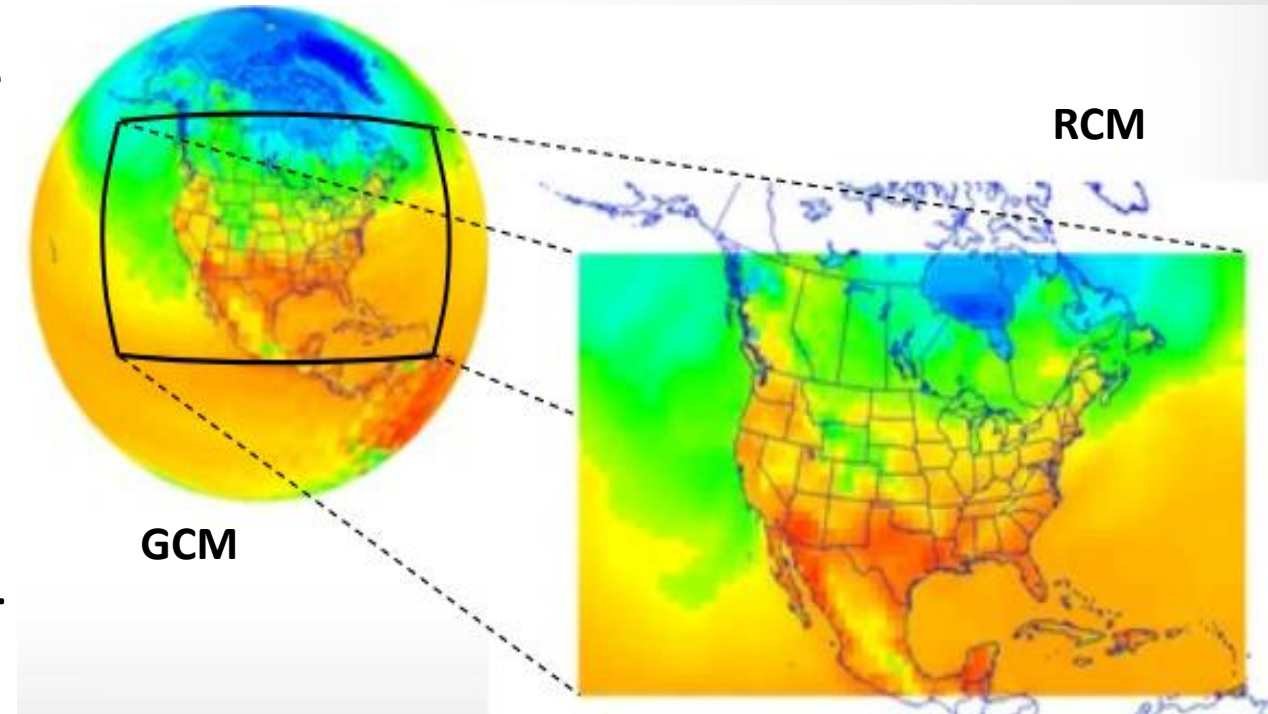


Climate:  
Long term pattern of  
weather  
(~30 years)

GCM

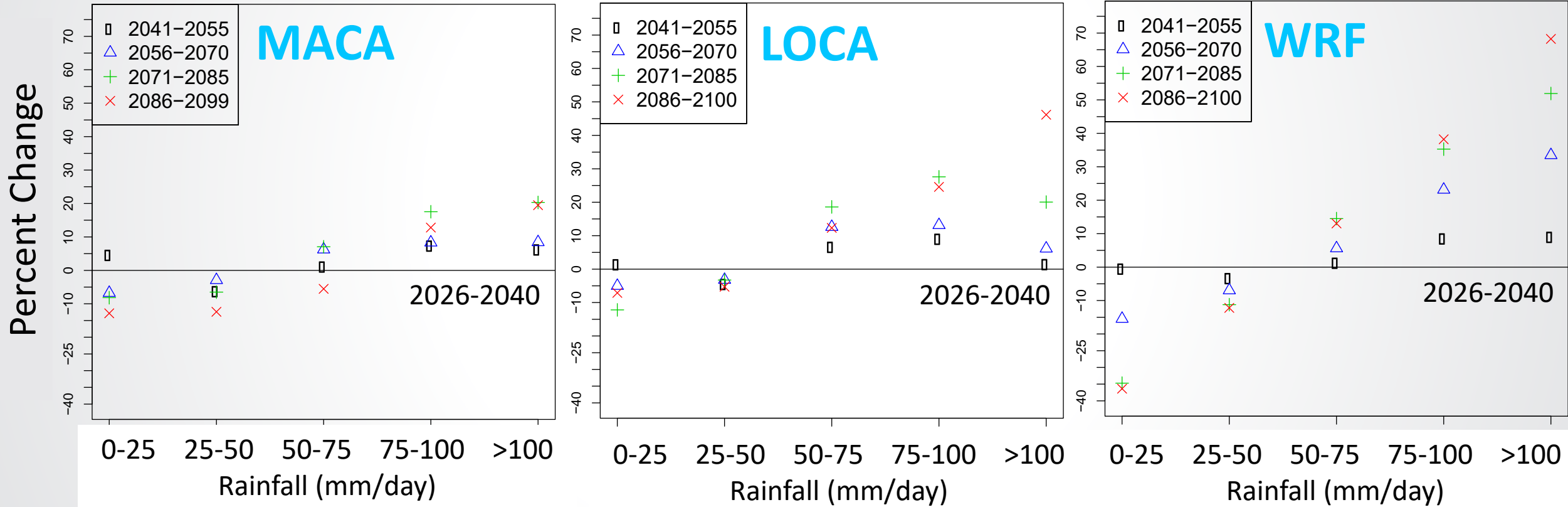


- Simulations use 2 GCMs (out of 21-model ensemble from **CMIP5**)
  - **CCSM4 (CESM)** at grid spacing of  $\sim 1^\circ$
  - **GFDL (CM3)** at grid spacing of  $\sim 2.5^\circ$
- Two scenarios used for greenhouse gas emissions (RCP 4.5 and RCP 8.5) to create experiments
- GCMs **dynamically** downscaled using **WRF** to a **36-km** grid over the CONUS. No bias correction. Hourly data aggregated to daily.
- Data for CONUS for 76-year future period (**2025-2100**)
- GCMs **statistically** downscaled by two methods: **LOCA** to **5-km** and **MACA** to **7-km** grid. Bias corrected. Daily data.



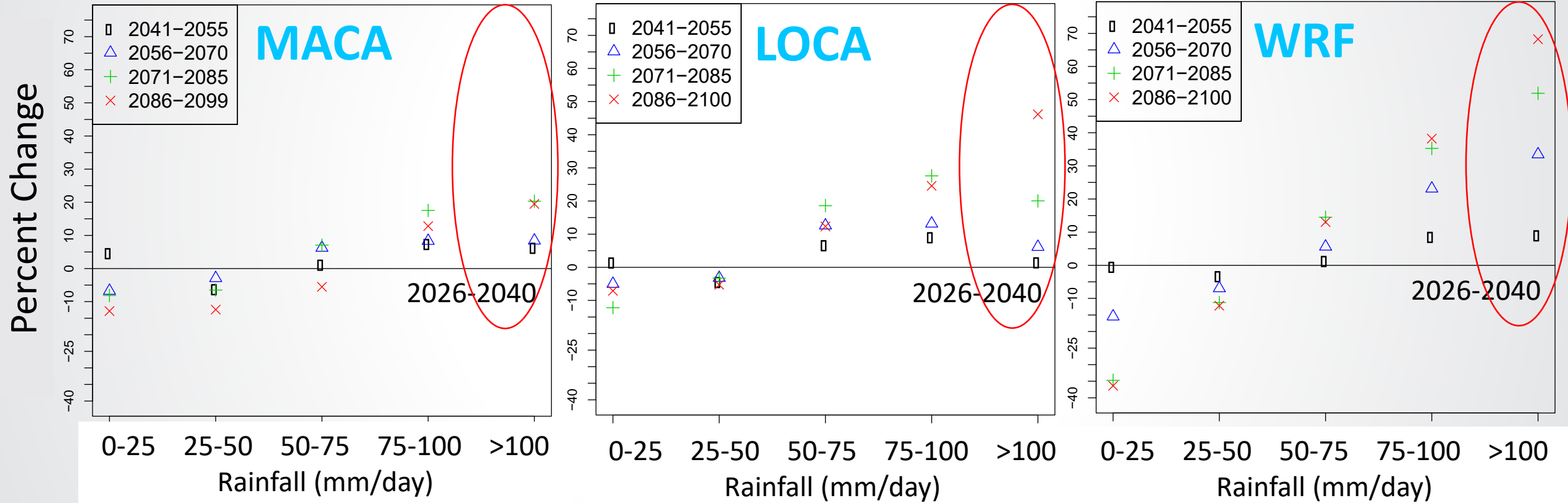


# Evolution of the projected number of rainfall events across the CONUS (relative to 2026-2040)





# Evolution of the projected number of rainfall events across the CONUS (relative to 2026-2040)

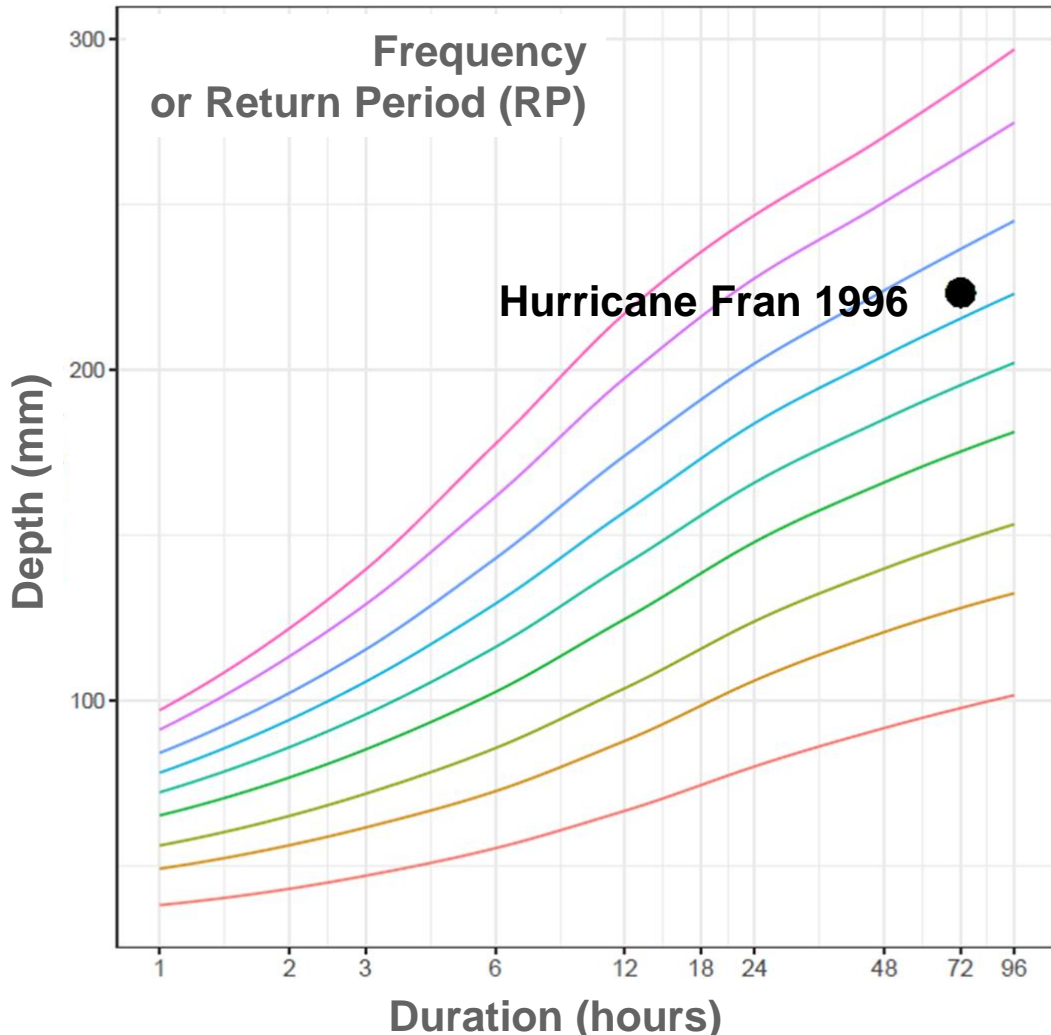




# Application

A “design storm” approach to look at changes in hurricane-level future extreme precipitation using dynamically downscaled data

**PIDF curves for Raleigh, NC (Atlas14)**



- Precipitation frequency estimates are probabilities of the occurrence of extremes events of particular intensity at particular duration.
- 30 years of precipitation data
- Stationarity assumption
- Annual Maximum Series (AMS)
- Fitting probability functions (GEV)
- Calculating probability distributions

**Frequency or Return Period**

- |           |             |              |
|-----------|-------------|--------------|
| — 2 years | — 25 years  | — 200 years  |
| — 5 years | — 50 years  | — 500 years  |
| — 10 year | — 100 years | — 1000 years |

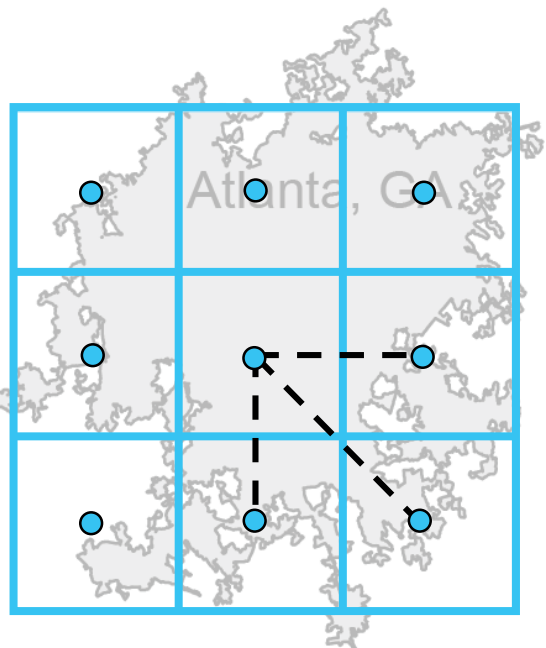


## OBSERVED DATA

- Extensive analysis of NOAA Atlas14 methodology and NCEI datasets
- Reproduced NOAA Atlas14 methodology and adapted it to gridded/modeled data

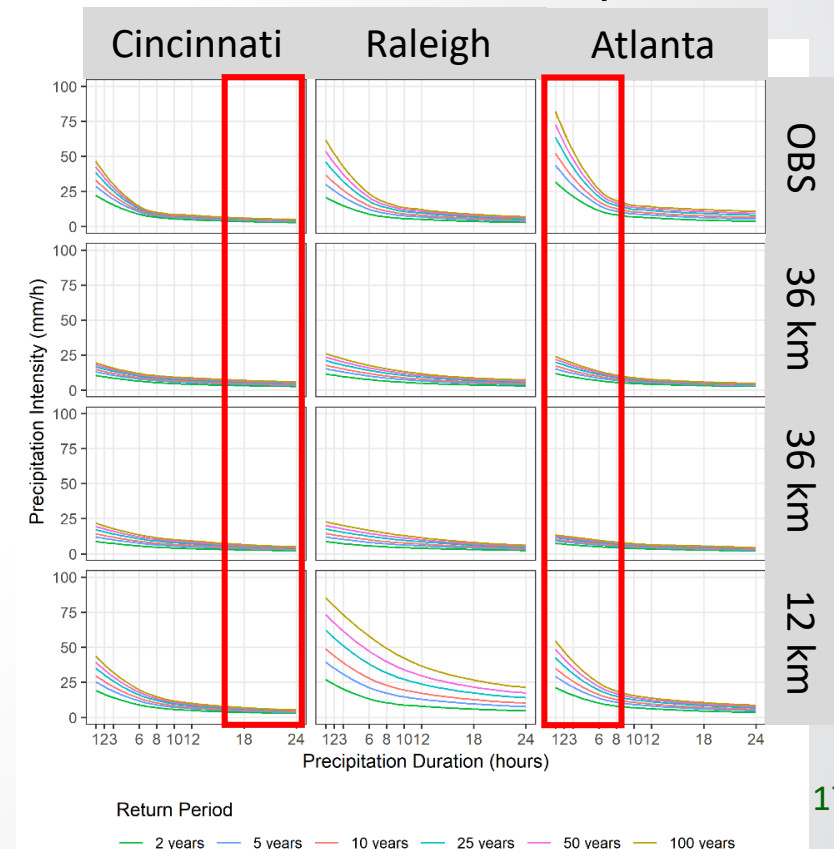
## MODEL RESOLUTION

- 36-km grid spacing is not sufficient to reproduce sub daily data but can be used for daily extreme precipitation.
- 12-km grid spacing was able to resolve sub-daily information.

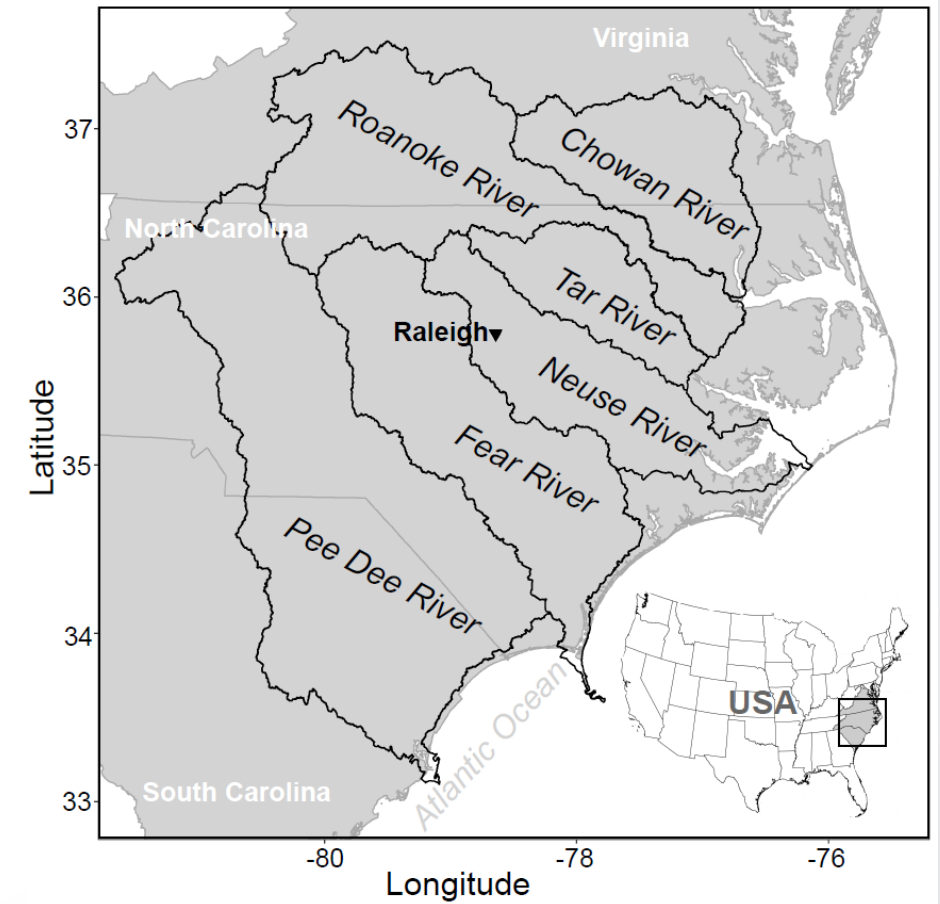


## METHODOLOGY

Best results with data aggregated using the Inverse Distance Weighting (IDW) method (RFA and other methods tested)



- Processed PRISM 4-km gridded observational data for three hurricanes:
  - **Floyd 1999**
  - **Matthew 2016**
  - **Florence 2018**
- Re-gridded (aggregated) the data to 36-km WRF domain.
- Subset for the Eastern North Carolina (ENC) watersheds.



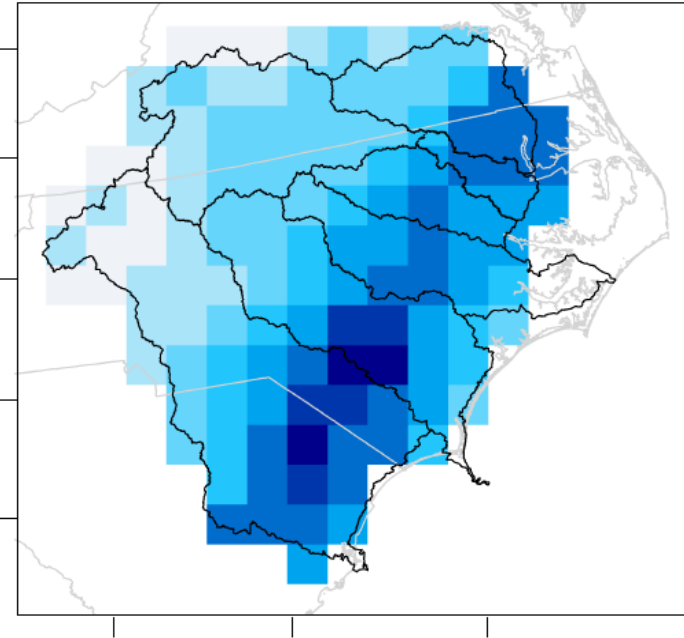
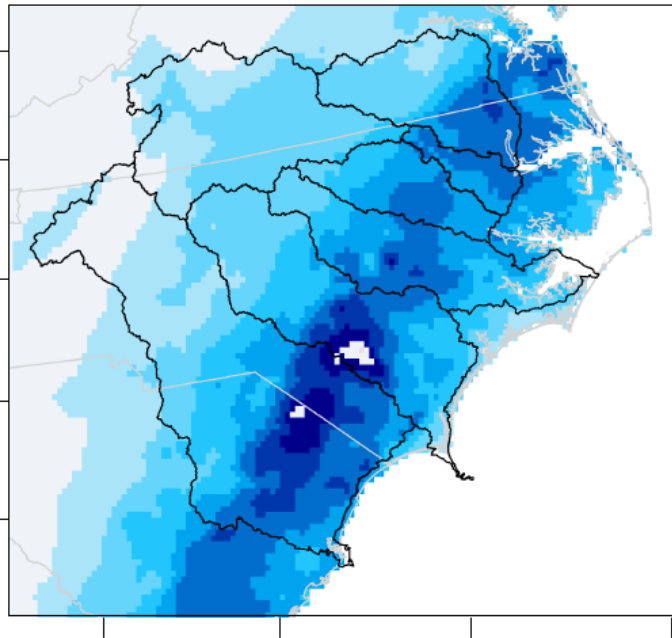
*Jalowska et al., in review*



# Observational Data PRISM

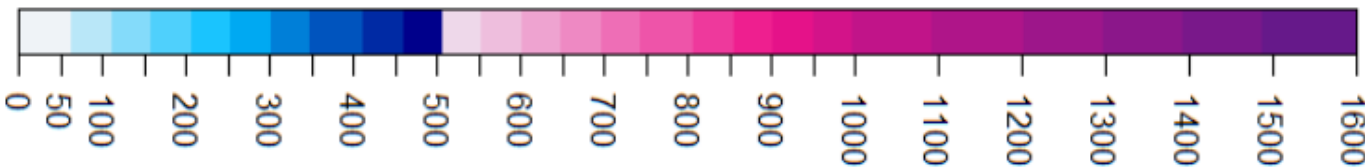
Hurricane Matthew (2016)  
Total 72h rainfall from PRISM  
at 4 km

Hurricane Matthew (2016)  
Total 72h rainfall from PRISM  
re-gridded to 36 km



Jalowska et al., in review

Total rainfall (mm)



- Downloaded 10-1000 years RP's for every 36-km grid-cell center from NOAA Atlas 14
- Assigned RP to each grid-cell based on the hurricane precipitation

NOAA ATLAS 14 POINT PRECIPITATION FREQUENCY ESTIMATES: NC

Data description  
Data type: Precipitation depth Units: Metric Time series type: Annual maximum

Select location  
1) Manually:  
a) By location (decimal degrees, use ° for S and W): Latitude Longitude Submit  
b) By station (list of NC stations): Select station  
c) By address Search

2) Use map (if ESR interactive map is not loading, try adding the host: https://js.arcgis.com/ to the browser, or contact us at hdsr\_question@noaa.gov)

Map Terrain  
a) Select location: Move crosshair or double click  
b) Click on station icon: Show stations on map

Location information:  
Name: Pinebluff, North Carolina, USA  
Latitude: 35.667°  
Longitude: -78.533°  
Elevation: 11.19 m

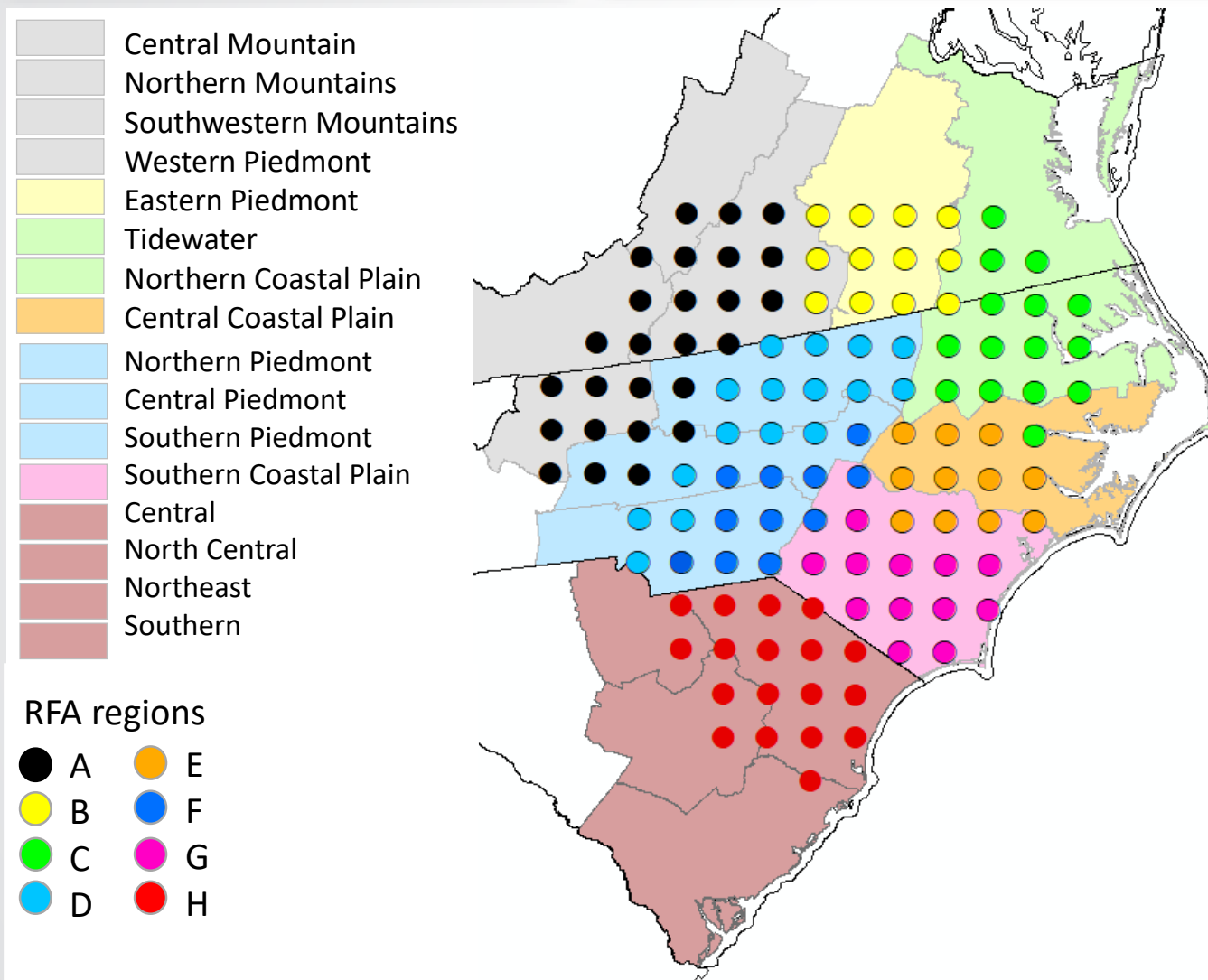
POINT PRECIPITATION FREQUENCY (PF) ESTIMATES  
With 90% CONFIDENCE INTERVALS AND SUPPLEMENTARY INFORMATION  
NOAA Atlas 14, Volume 2, Version 3

PF tabular PF graphical Supplementary information Print page

Duration	Annual recurrence probability: years															
	1/2	1/3	1/5	1/10	1/25	1/50	1/100	1/200	1/500	1/1000	1/2000	1/5000	1/10000	1/20000	1/50000	
5mm	12	15	22	33	50	77	110	150	210	290	390	520	690	910	1180	1550
10mm	18	23	33	48	70	100	140	190	260	340	450	590	770	1000	1300	1700
15mm	24	30	42	60	85	120	160	220	290	380	500	650	850	1100	1400	1800
30mm	34	42	56	78	105	140	180	240	310	400	520	680	880	1150	1450	1850
60mm	42	51	67	92	120	160	210	270	350	450	580	750	960	1250	1600	2000

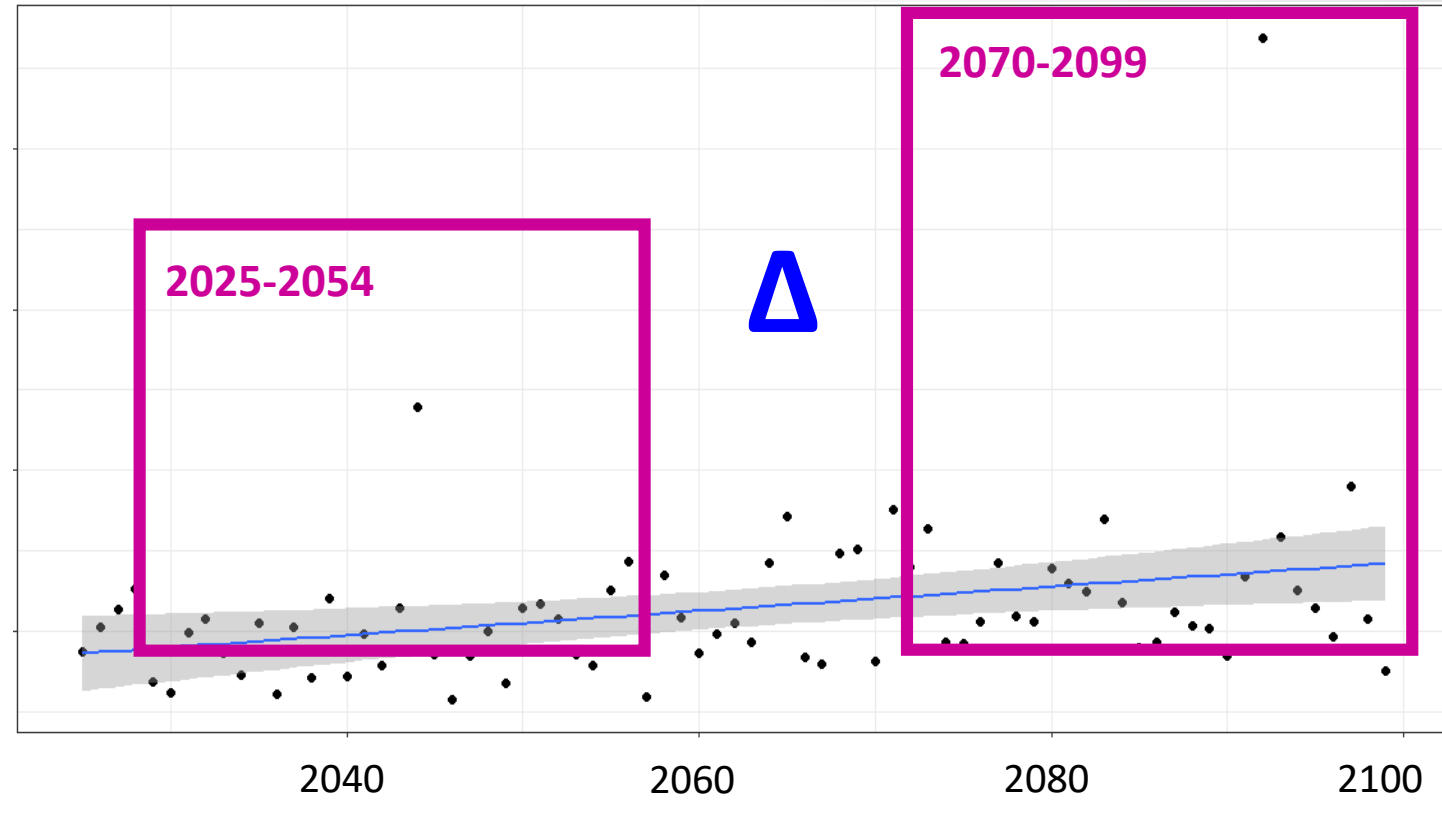
[https://hdsc.nws.noaa.gov/hdsc/pfds/pfds\\_map\\_cont.html](https://hdsc.nws.noaa.gov/hdsc/pfds/pfds_map_cont.html)

Start of the pink scale indicates the 1000-year rainfall or more (as of current NOAA Atlas 14)



- Constructed Annual Maximum Series for 2025-2100 for every grid cell
- Regional Frequency Analysis (RFA) allows us to develop the IDF curves for the regions with the same probability distribution, confirmed by tests
- Defined 8-10 regions for RFA based on the US climate divisions: Mountains, Northern, Central and Southern Piedmont and 4 Coastal Plain regions, using each grid cell as a “station”

- Using the RFA approach we developed PIDF for three realizations for two 30-year periods: **2025-2054** and **2070-2099**
- Calculated relative change ( $\Delta$ ) in PIDF curves between these two periods for two durations:
  - **24-hour**
  - **72-hour**
- Applied developed  $\Delta$  to precipitation totals of three hurricanes by the frequency of each grid cell (Design Rainfall Approach- DRA)

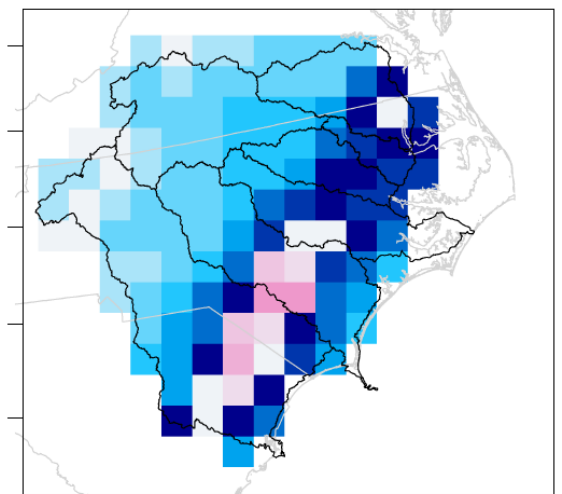
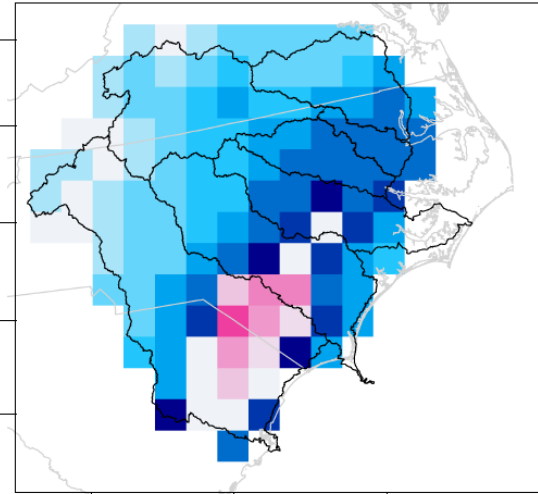
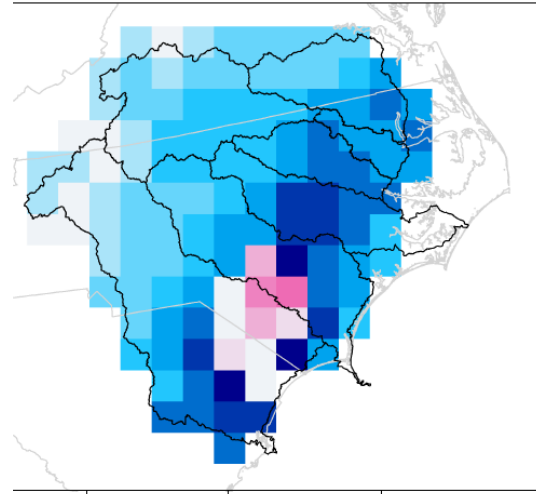
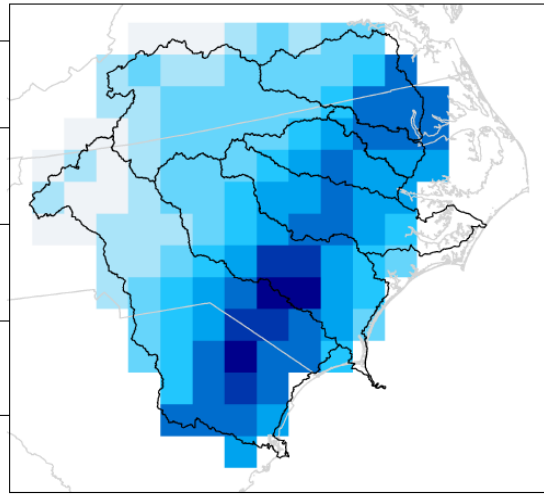


**OBSERVED (PRISM)**  
**Total 72h rainfall**  
 max = 375 mm  
 total = 26,876 Mt of water

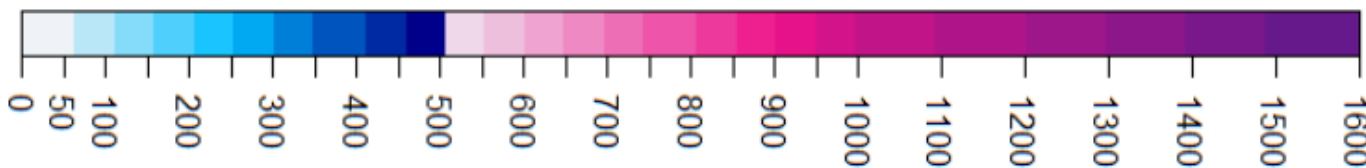
**CESM RCP 4.5**  
**Total 72h rainfall**  
 max = 714 mm  
 total = 32,891 Mt of water

**CESM RCP 8.5**  
**Total 72h rainfall**  
 max = 819 mm  
 total = 36,257 Mt of water

**CM3 RCP 8.5**  
**Total 72h rainfall**  
 max = 633 mm  
 total = 36,127 Mt of water



Total rainfall (mm)



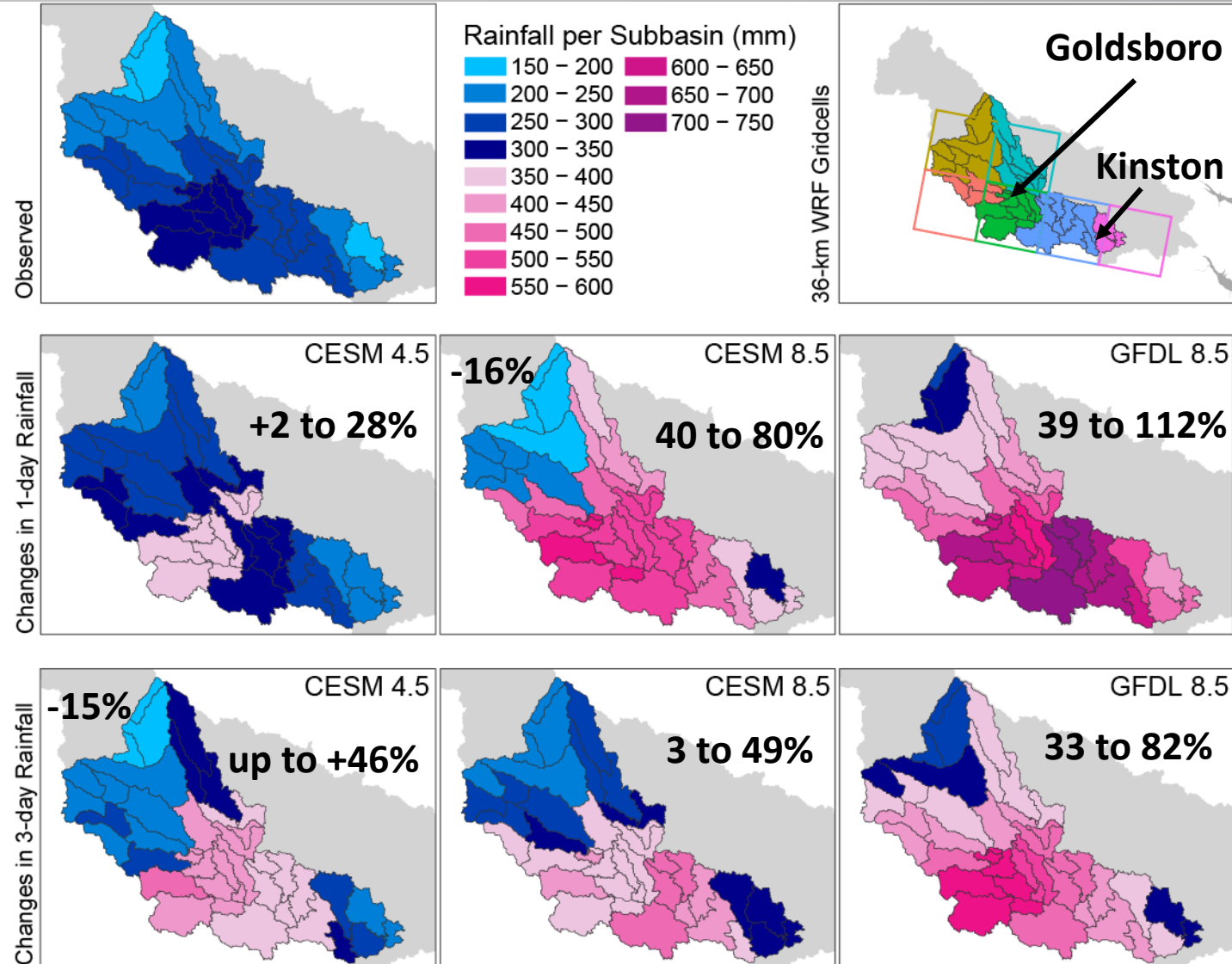
	CESM RCP 4.5	CESM RCP 8.5	CM3 RCP 8.5
Total	22%	42%	34%
Max	90%	118%	69%



# Matthew “2100” Induced Runoff

- Used generated rainfall data to produce runoff and stream flow from Matthew “2100” in the Neuse River Basin.
- Used observed rainfall per subbasin from station data instead of PRISM.
- CESM 4.5 scenario shows higher increases in 3-day rainfall than in 1-day rainfall. Both 8.5 scenarios show smaller changes in the 3-day rainfall than in the 1-day rainfall

*Jalowska et al., in prep*



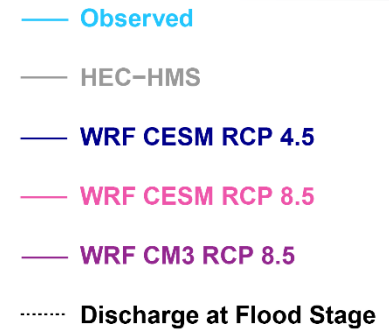
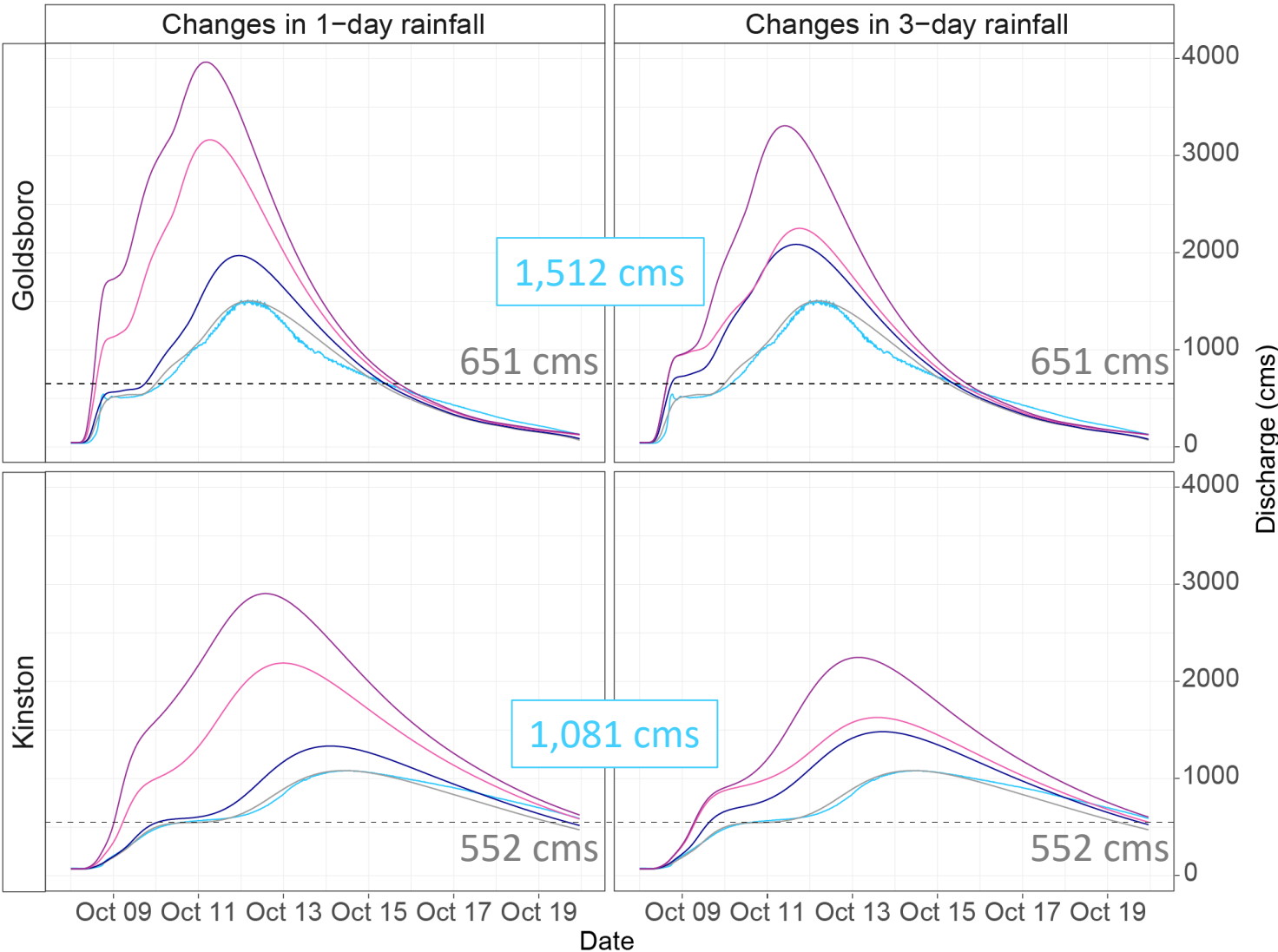


# Matthew "2100" Induced Runoff

- The modeled Matthew 2016 peak discharge (HEC-HMS) was within 1% of observed for the both gaging stations.

**FSD- Flood Stage Discharge**

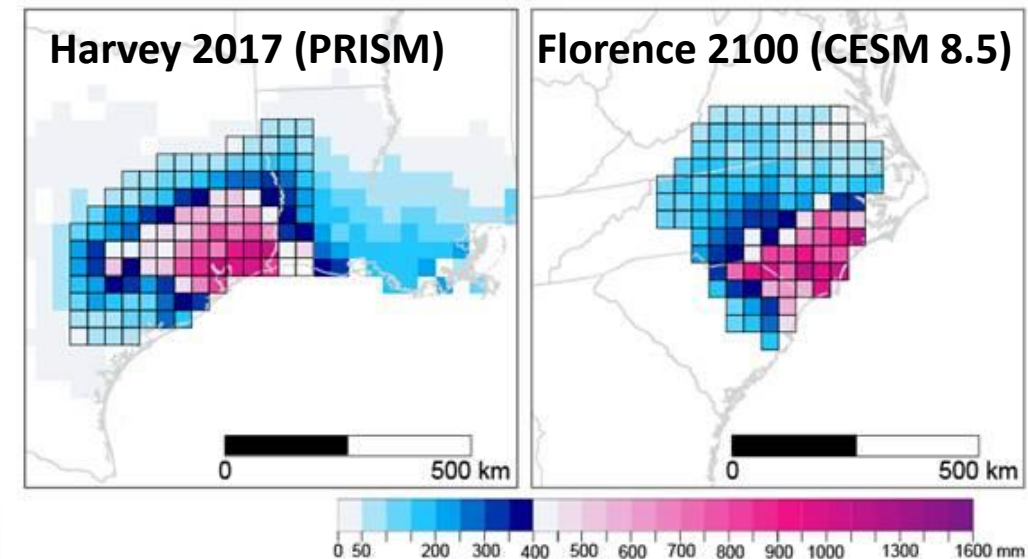
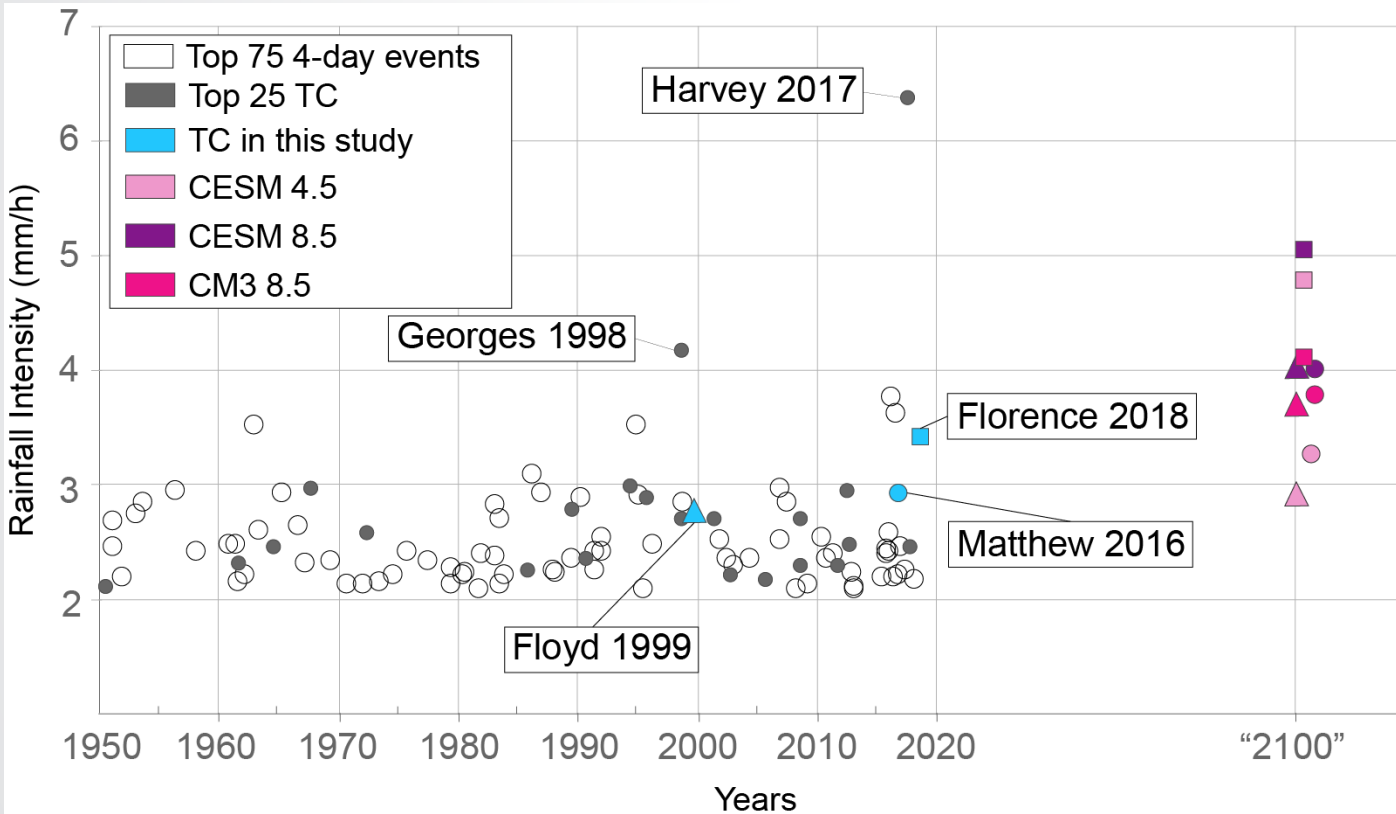
	Floyd (1991)	Matthew (2016)	CESM 4.5	CESM 8.5	CM3 8.5
		-> 25 yr			
Goldsboro	x 1.7	x 2.3 +0.6	x 3	x 5 (1d) x 3.5 (3d)	x 6 (1d) x 5 (3d)
Kinston		x 2	x 2.5	x 4 (1d) x 34 (3d)	x 5 (1d) x 4 (3d)





- The increases in precipitation using the DRA are realistic in the light of other, historical events, and overall suggest underestimation in our downscaled data (makes sense).

- Total rainfall for Harvey within area corresponding to ENC (indicated in black grid net) was 53,808 Mt, and from Florence 2100 under CESM 8.5 was 47,809 Mt.
- However, we don't know if we could produce Harvey in NC.



Rainfall intensity is based on area of  $2^\circ \times 2^\circ$  (50,000 km<sup>2</sup>) area used in Kunkel & Champion, 2019

- extended North America + Puerto Rico 12 km domain for CESM 8.5
- Extended historical period: 1975-2005 (future: 2025-2100)

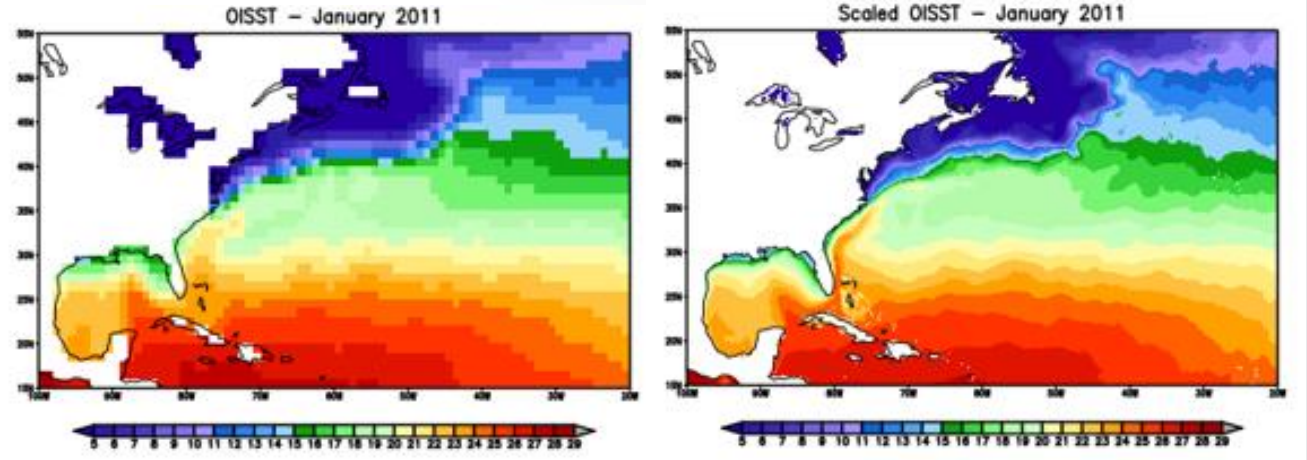
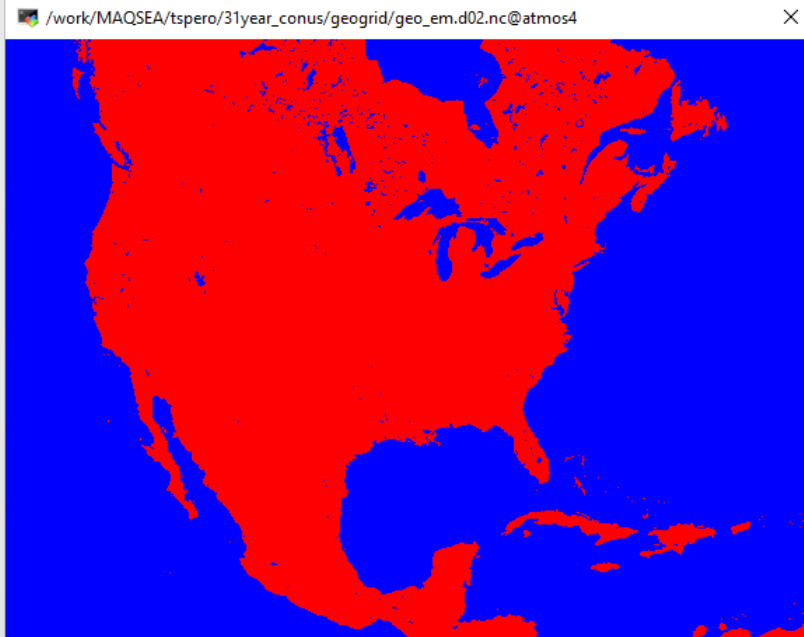


Figure credit Jared Bowden



- Updated sea surface and lakes temperatures
- Updated WRF version
- Updated modeling options and newer science.

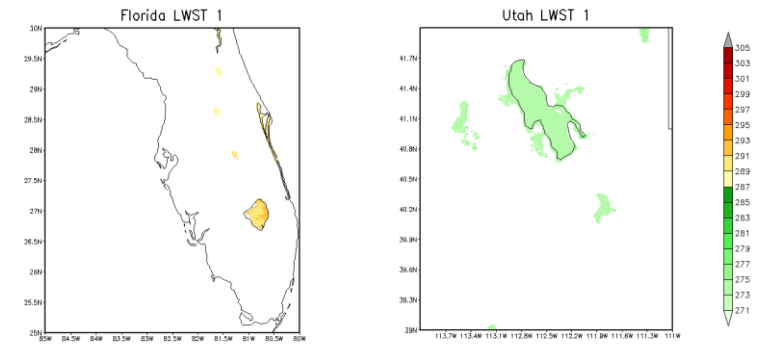


Figure credit Jared Bowden



# Discussion: Q & A

- Shared or common research interests?
- Sharing relevant information for on-going research efforts
- Exploring ways to coordinate research efforts on extreme weather and stormwater BMP design and planning data and design tools?

## Contact Information:

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- The end



# Modeling options

WRF Model Options		NCAR/DOE CESM	GFDL CM3	
Shortwave Radiation	RRTMG	GCM Resolution	0.875 x 1.25	2 x 2.5
Longwave Radiation	RRTMG	WRF Domain	36-km only	108-36-km, two-way
Microphysics	WSM6	WRF Version	3.4.1 + mods to K-F CuPa (Herwehe et al., <i>JGR-A</i> , 2014)	3.6
Convection	Kain-Fritsch with Radiative Feedbacks	Scenarios	RCP4.5 and RCP8.5	RCP8.5
PBL	WSU	Radiative Forcing	Standard	Following the RCPs
Land-Surface Model	Noah	Lake Temperatures	Imported from CLM (Spero et al., <i>J.</i>	Modeled with FLake (Mallard et al., <i>JGR</i> - <sup>29</sup>
Nudging	Spectral Nudging toward GCM			



# Representative Concentration Pathways

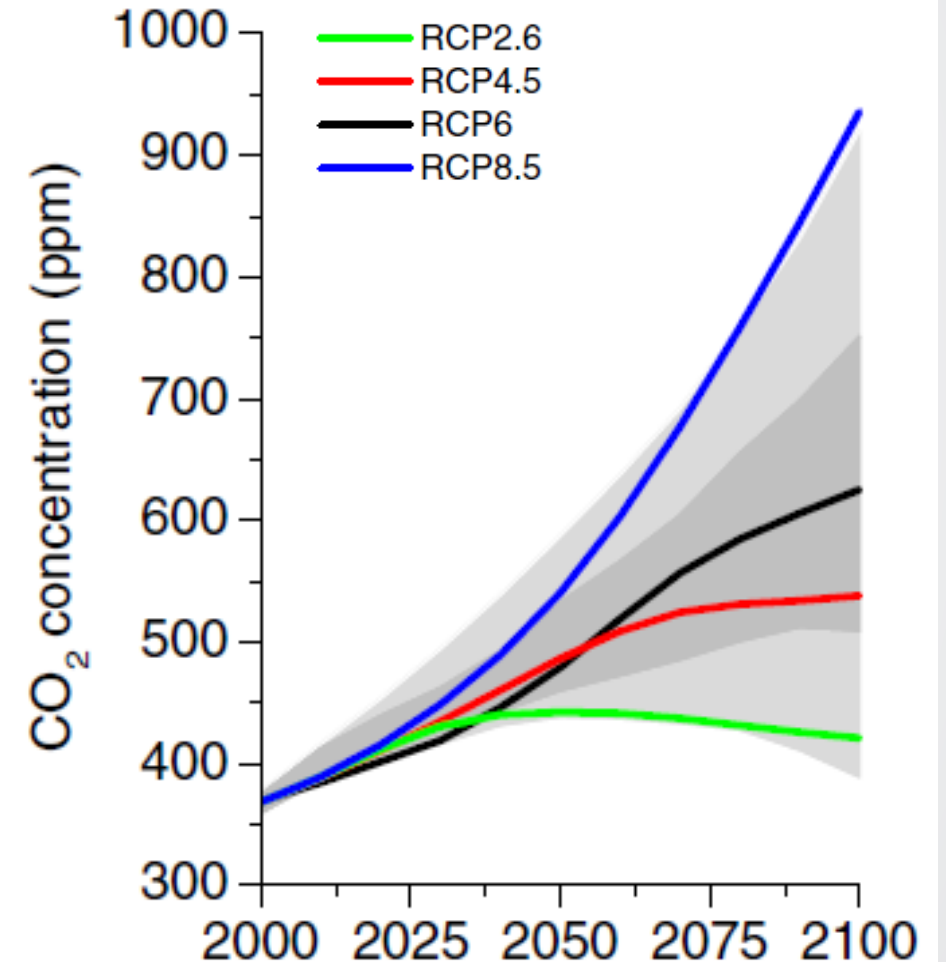
Data for three realizations:  
**CESM 4.5, CESM 8.5 & CM3 8.5**

## CESM RCP4.5

- Moderate global emission reduction.
- Assumes some mitigation strategies and technologies.
- CO<sub>2</sub> concentration rising until 2040.
- CO<sub>2</sub> concentration reaches 540 ppm by 2100.

## CESM RCP8.5 & CM3 RCP8.5

- High GHG emission pathway scenario.
- No global emissions reduction.
- Reaching 940 ppm in 2100.



- Comparison with PGW from our paper (PGW was used on other events)

“The average increases of 24% (20%) in 3-day (4-day) duration under CM3\_8.5 are comparable to those from ensemble averages from pseudo–global warming (PGW) experiments: 24%<sup>32</sup> and 23%<sup>33</sup>. However, the average change in CESM\_8.5 of 39% (36%) in 3-day (4-day) duration surpasses the PGW average.”

*32. Gutmann, E. D., et al. Changes in Hurricanes from a 13-Yr Convection-Permitting Pseudo–Global Warming Simulation. J. Climate, 31, 3643–3657, (2018)*

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