

# Extreme Rainfall in NYC from Ida: Insights from a High-Resolution Forecast Model



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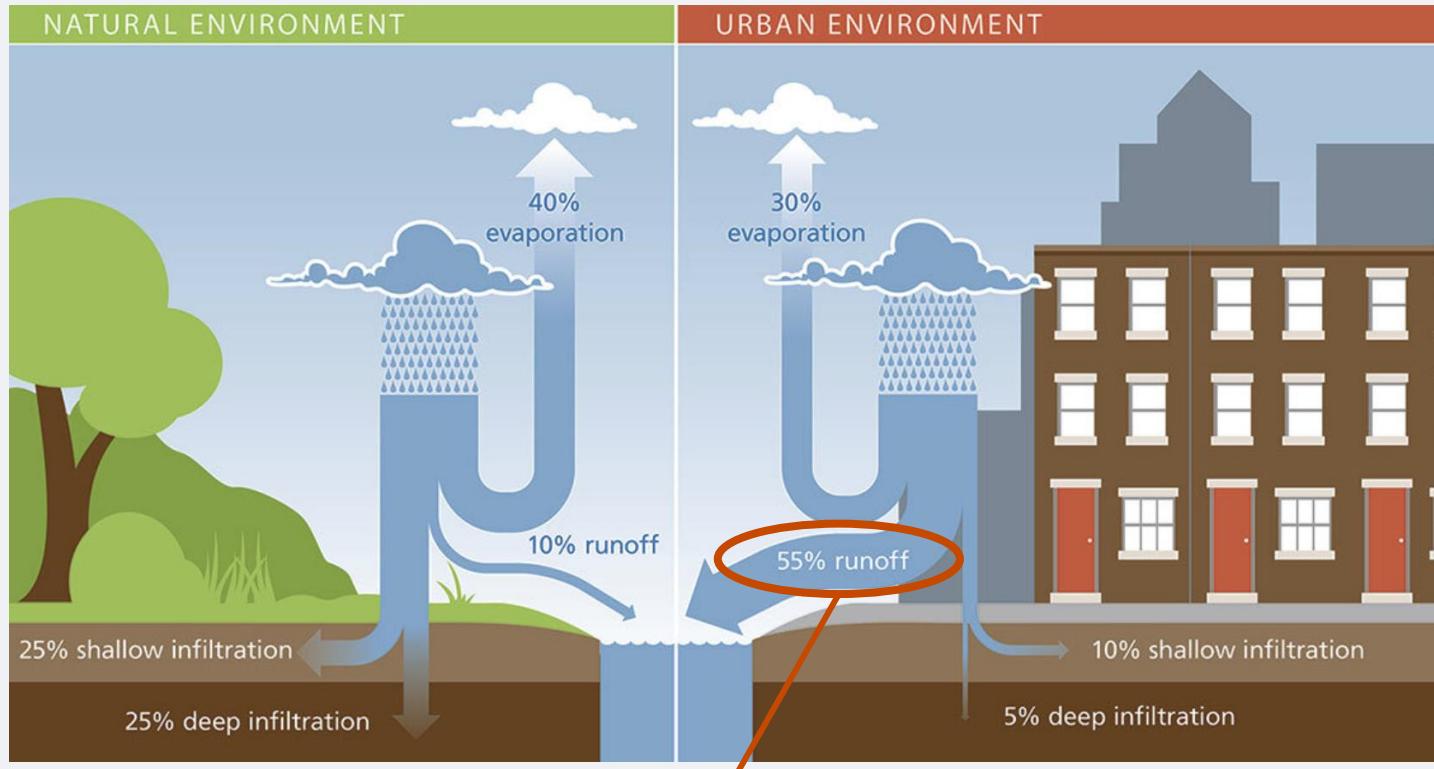
The City College  
of New York



# Background & Motivation

# Urban Areas are Susceptible to Flash Flooding

- Permeable soil
- Vegetation to intercept and absorb water
- Only ~10% runoff

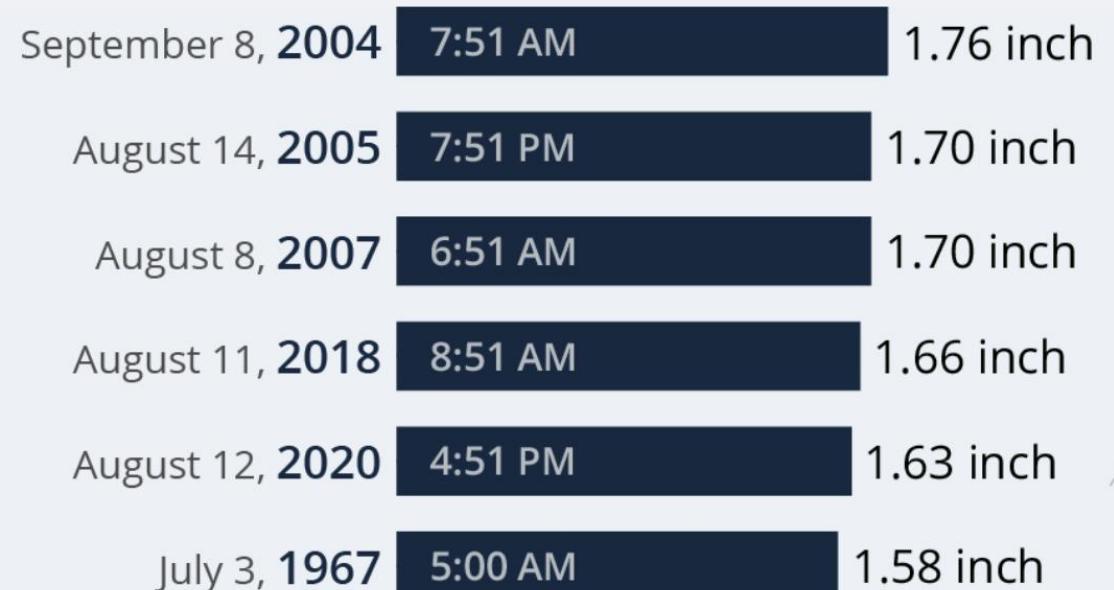


- Impervious concrete & asphalt
- Little vegetation
- **55% runoff on average**

# Hourly Rainfall in NYC

- **Hourly rainfall** → main metric for flash flood risk
- NYC infrastructure designed for a **5-year storm** (20% chance any year)
- **Threshold: 1.75 inches per hour**

Central Park ASOS Hourly Rainfall Amounts



Source: Iowa Environmental Mesonet

*Statistica*

Note: These values are from only one weather station and regular hourly METARs

# 2021 was already anomalously wet

- 2021: NYC's **second wettest summer on record**
- 21 August 2021: Hurricane Henri brings 1.94"/hr



Central Park ASOS Hourly Rainfall Amounts



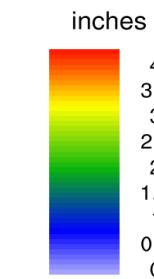
Source: Iowa Environmental Mesonet

*Statistica*

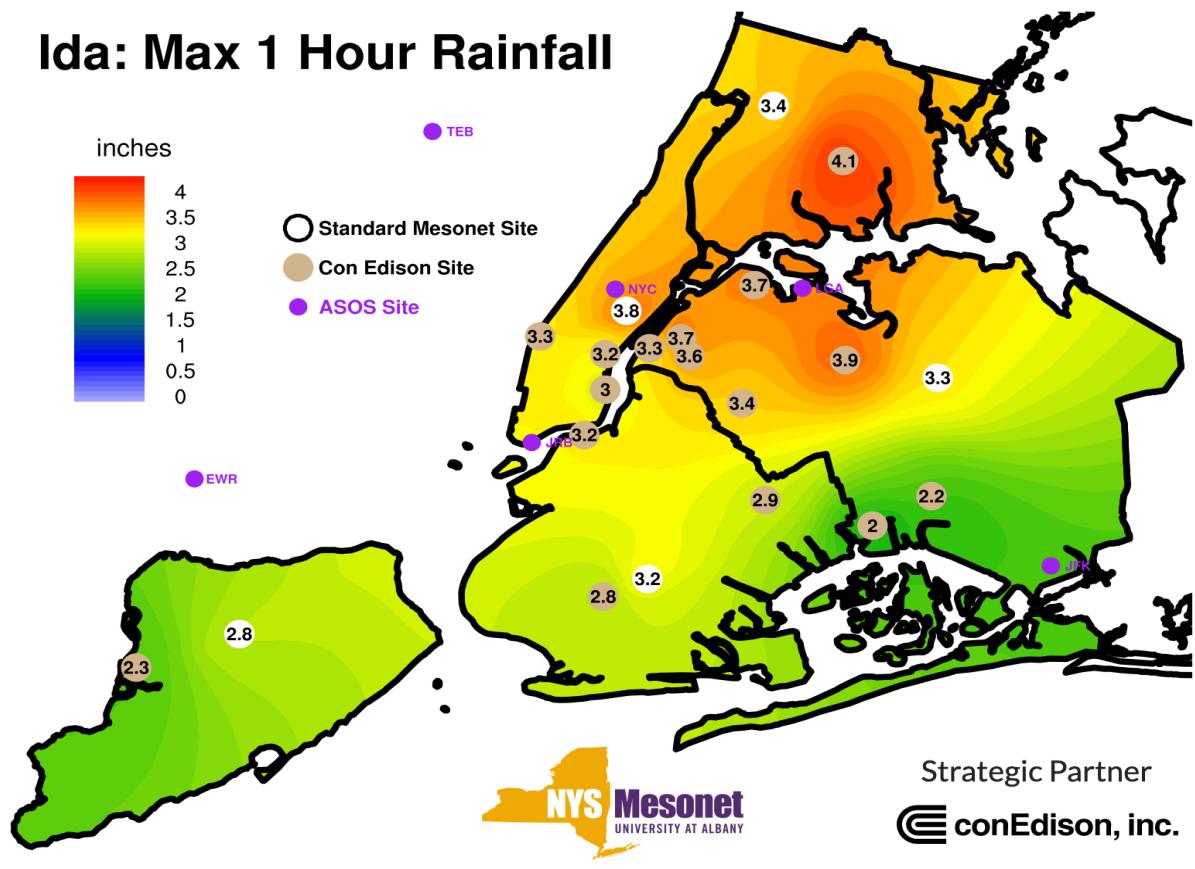
# Ida shatters previous records

NYC Micronet Rainfall Measurements

## Ida: Max 1 Hour Rainfall



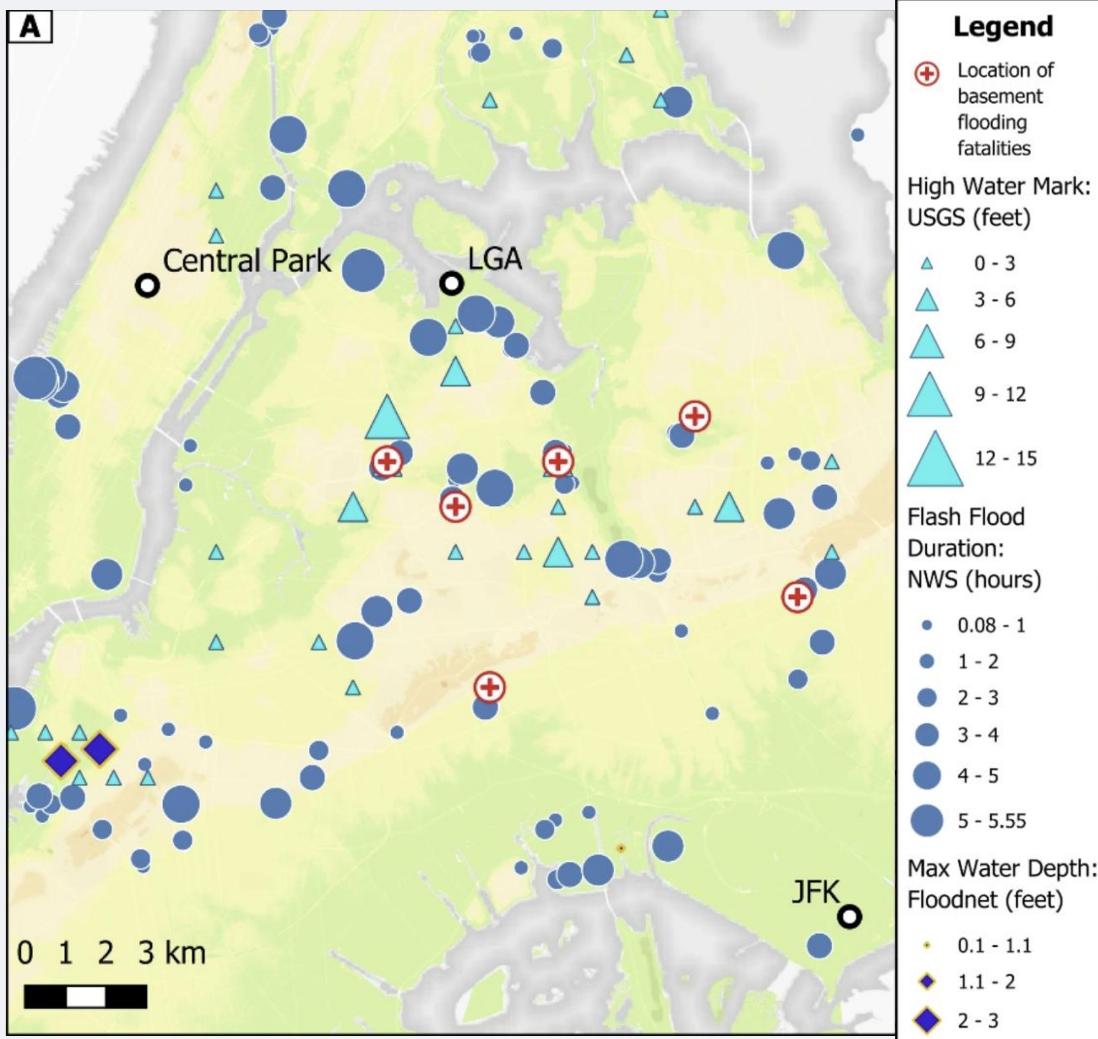
- Standard Mesonet Site
- Con Edison Site
- ASOS Site



Central Park ASOS Hourly Rainfall Amounts



# Impact: Flash Flooding

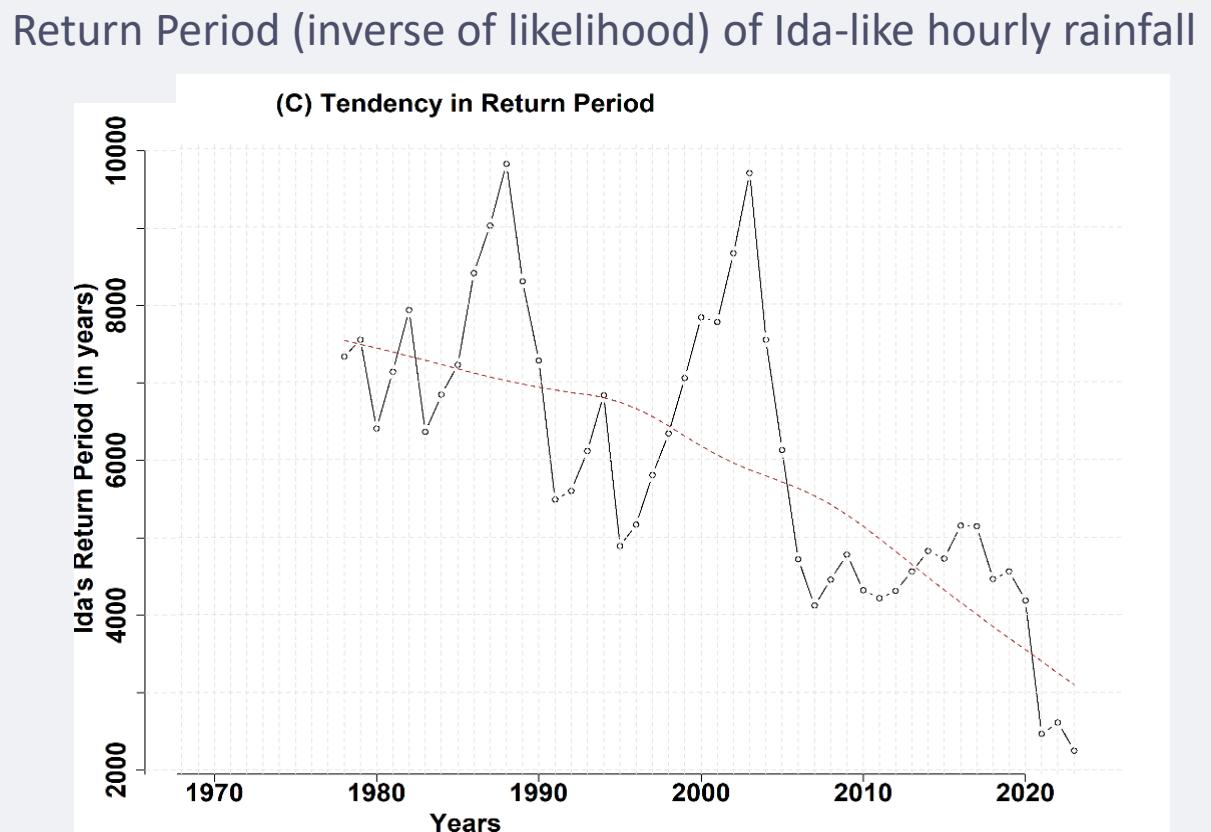


Adapted from Fig. 1 of Mossel et al. (2024)

- Hours-long flash flooding
- 13 dead in NYC
  - 11 from basement flooding
- Billions of dollars of damage
- Different from past storms:  
**affected inland neighborhoods instead of coasts**

# Events like Ida have and will become more common

- Mossel et al. (2024) found that the return period of Ida-like events **has decreased over past 4 decades**
- Trend likely to continue and accelerate with **climate change**
- **Despite increasing likelihood, Ida-like events are super rare**



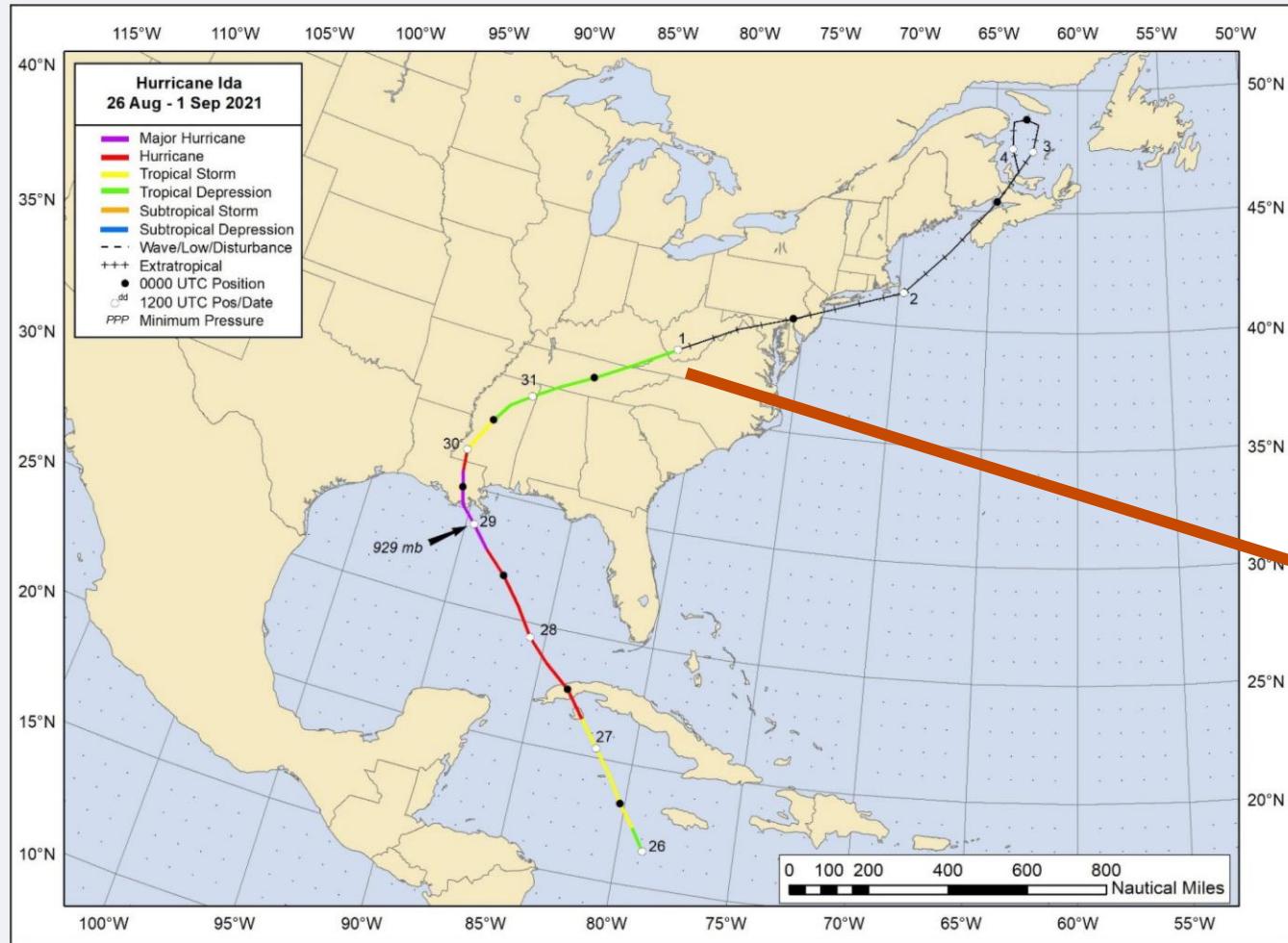
Adapted from Fig. 2 of Mossel et al. (2024)

# Meteorological History

What exactly happened?

# Hurricane Ida Track & Transition

## Hurricane Ida Dated Track & Classifications

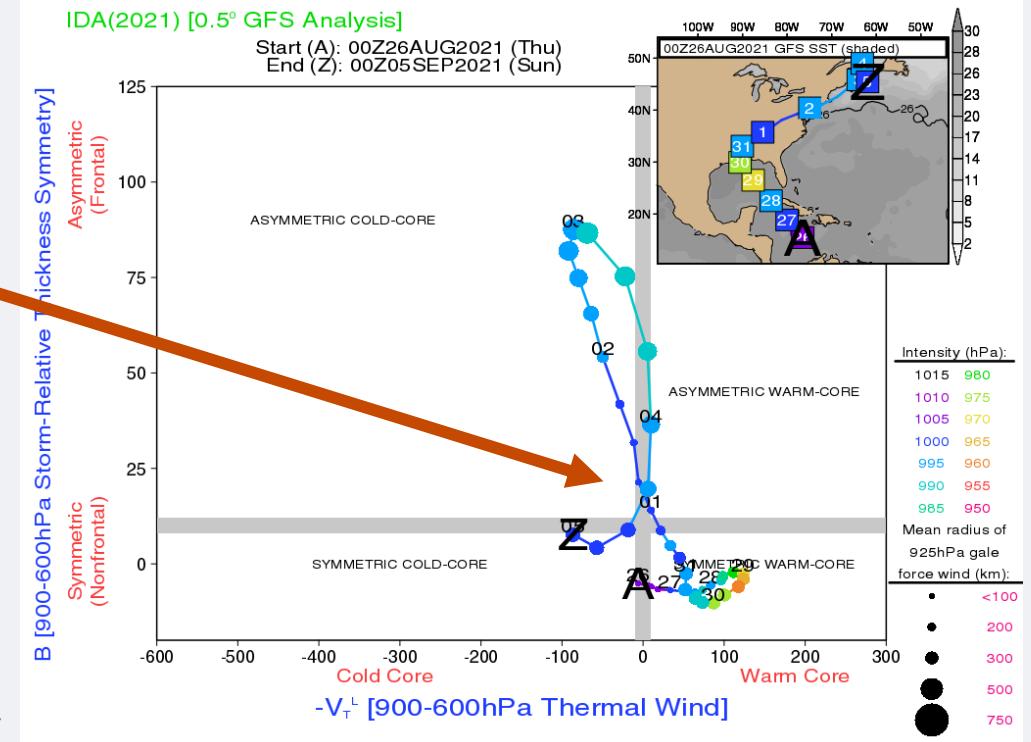


**Formed: Aug 26, 2021**

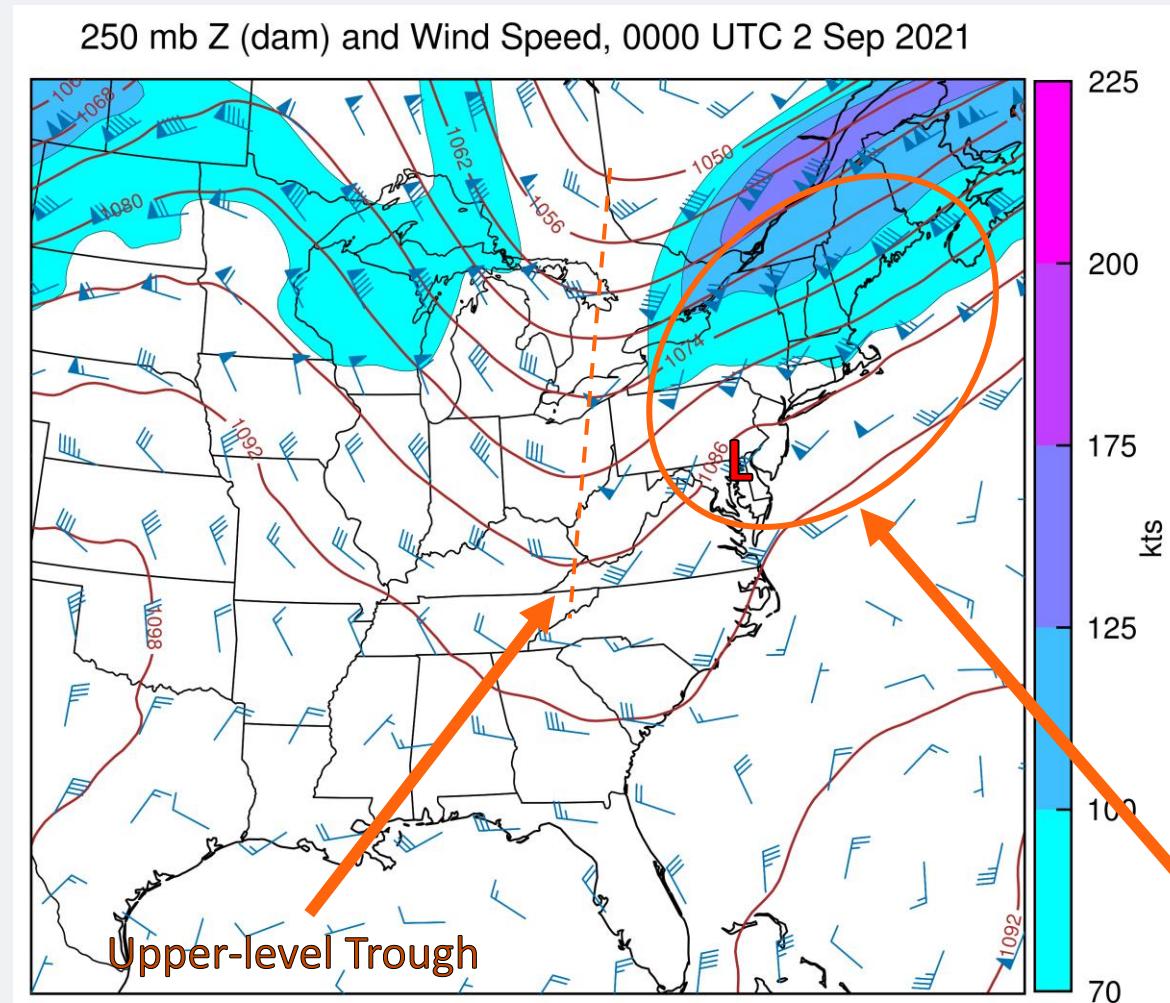
ET Transition: Sep 1, 2021

Dissipated: Sep 4, 2021

# Hurricane Ida Phase Space Diagram



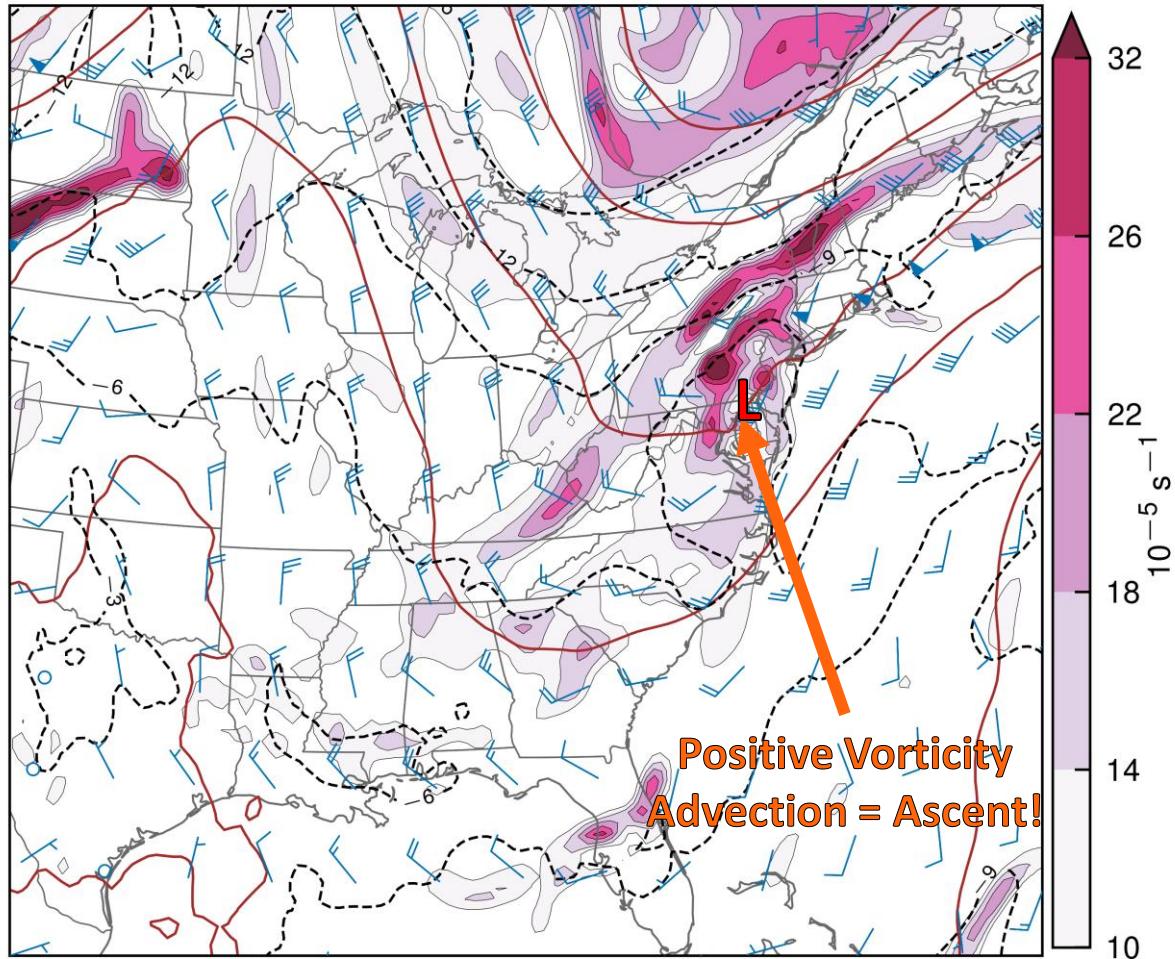
# Upper-level support



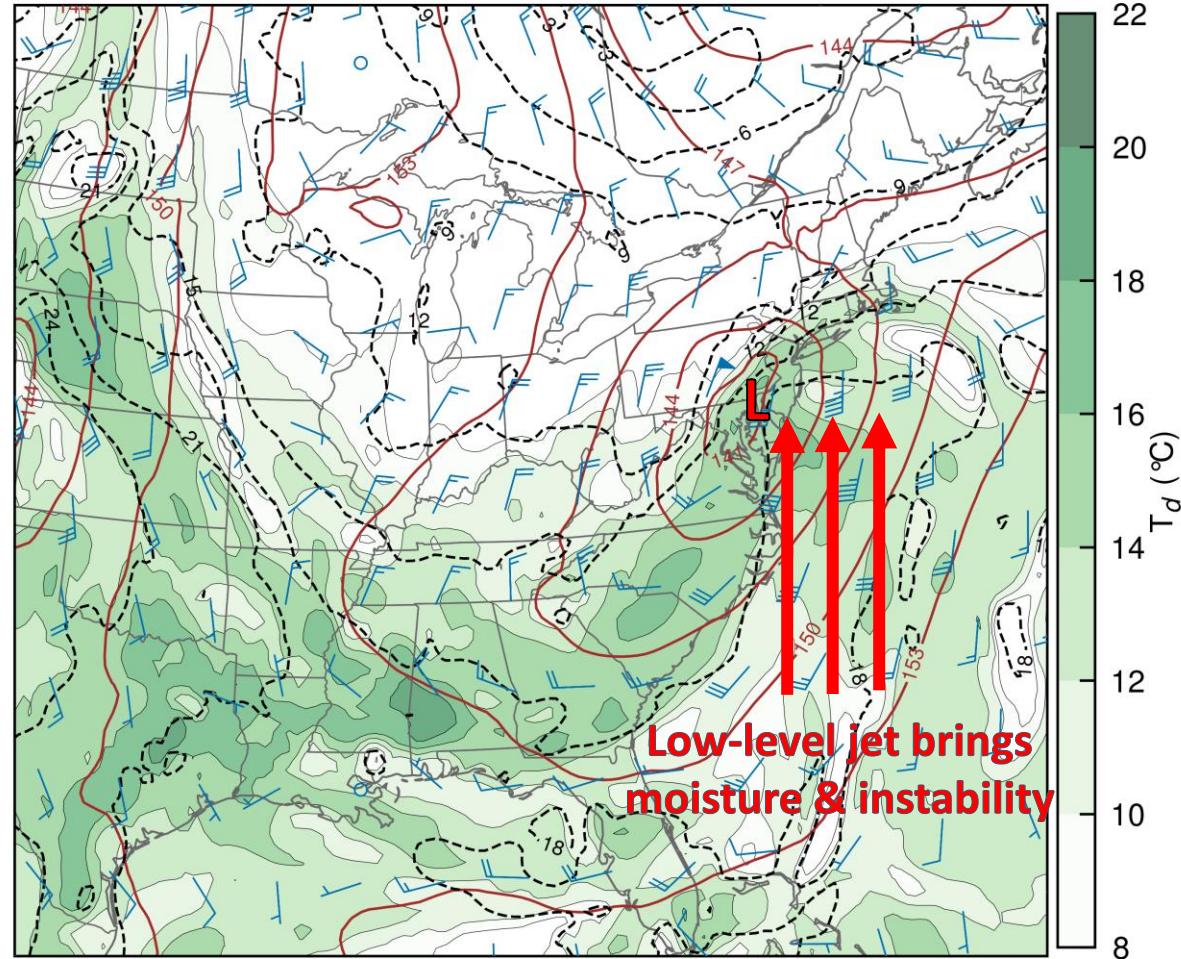
- Ohio Valley trough @ 250 mb
- Right-entrance region of jet streak = UL Divergence = Ascent = **Precipitation!**
- UL divergence also supports a **low-level jet** to bring energy and moisture

# Mid-level Support

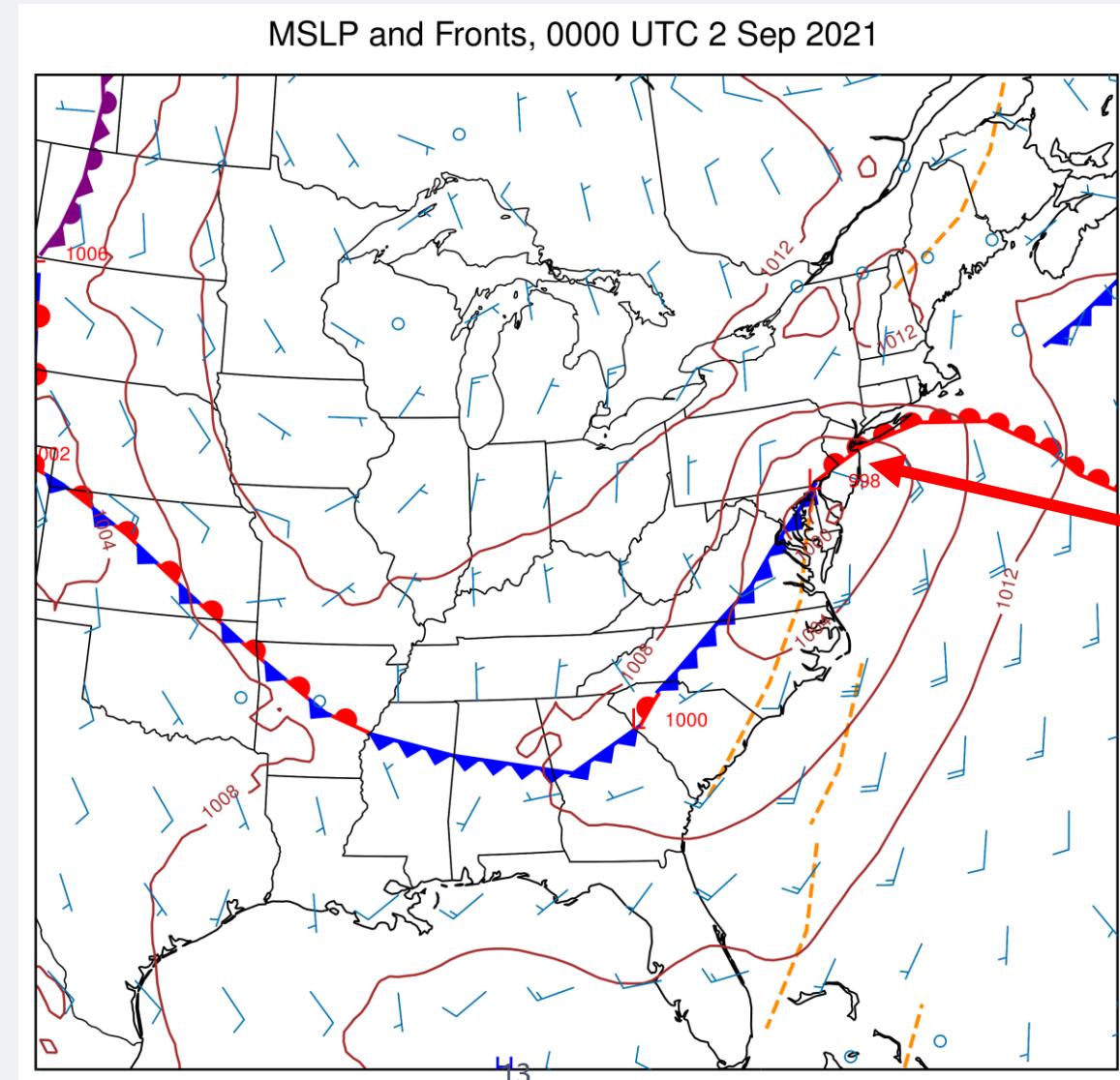
500 mb Z (dam) and Abs Vort, 0000 UTC 2 Sep 2021



850 mb Z (dam) and T ( $^{\circ}\text{C}$ ), 0000 UTC 2 Sep 2021



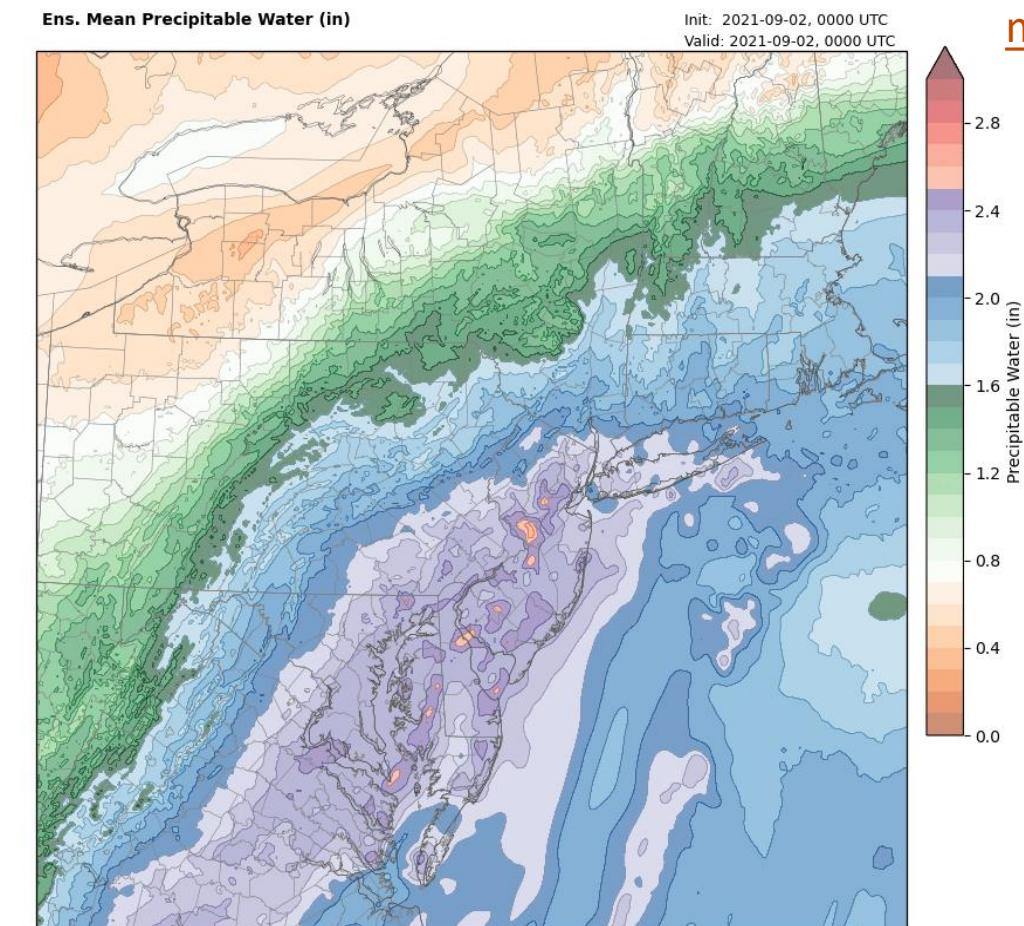
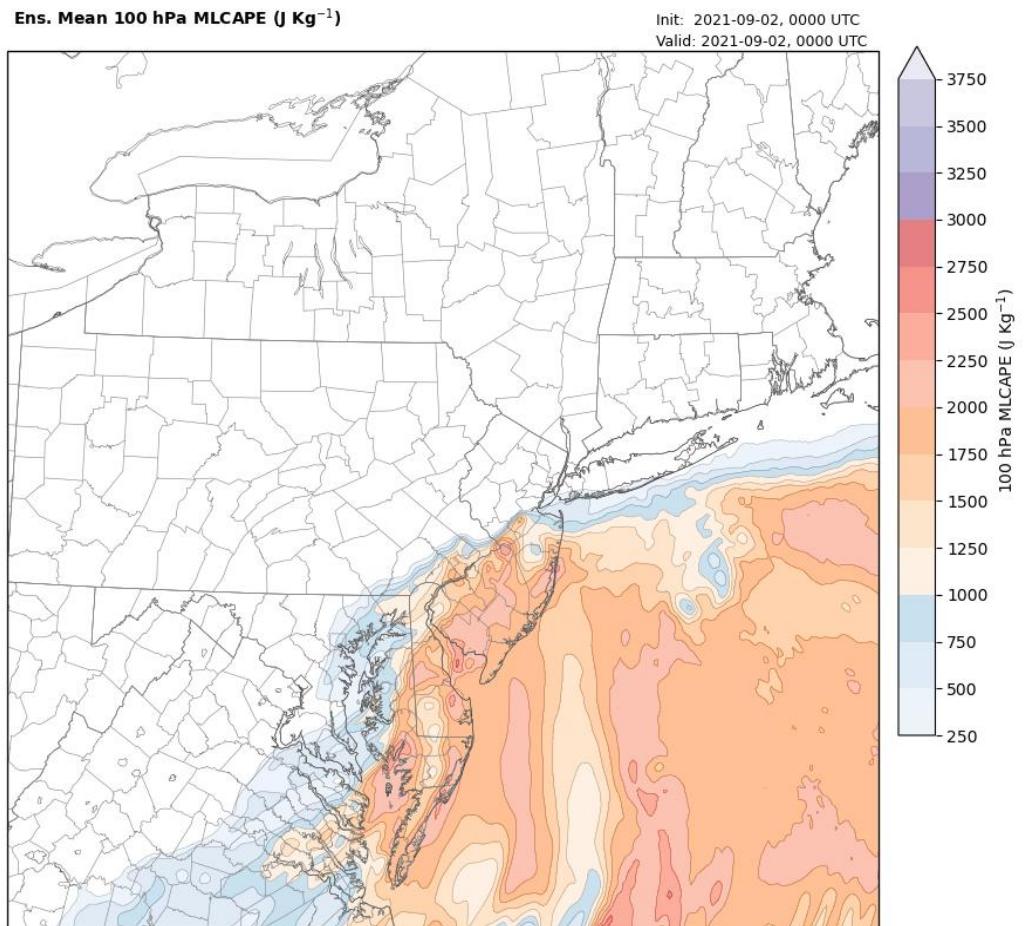
# Surface Analysis



# Mixed-Layer CAPE & Precipitable Water

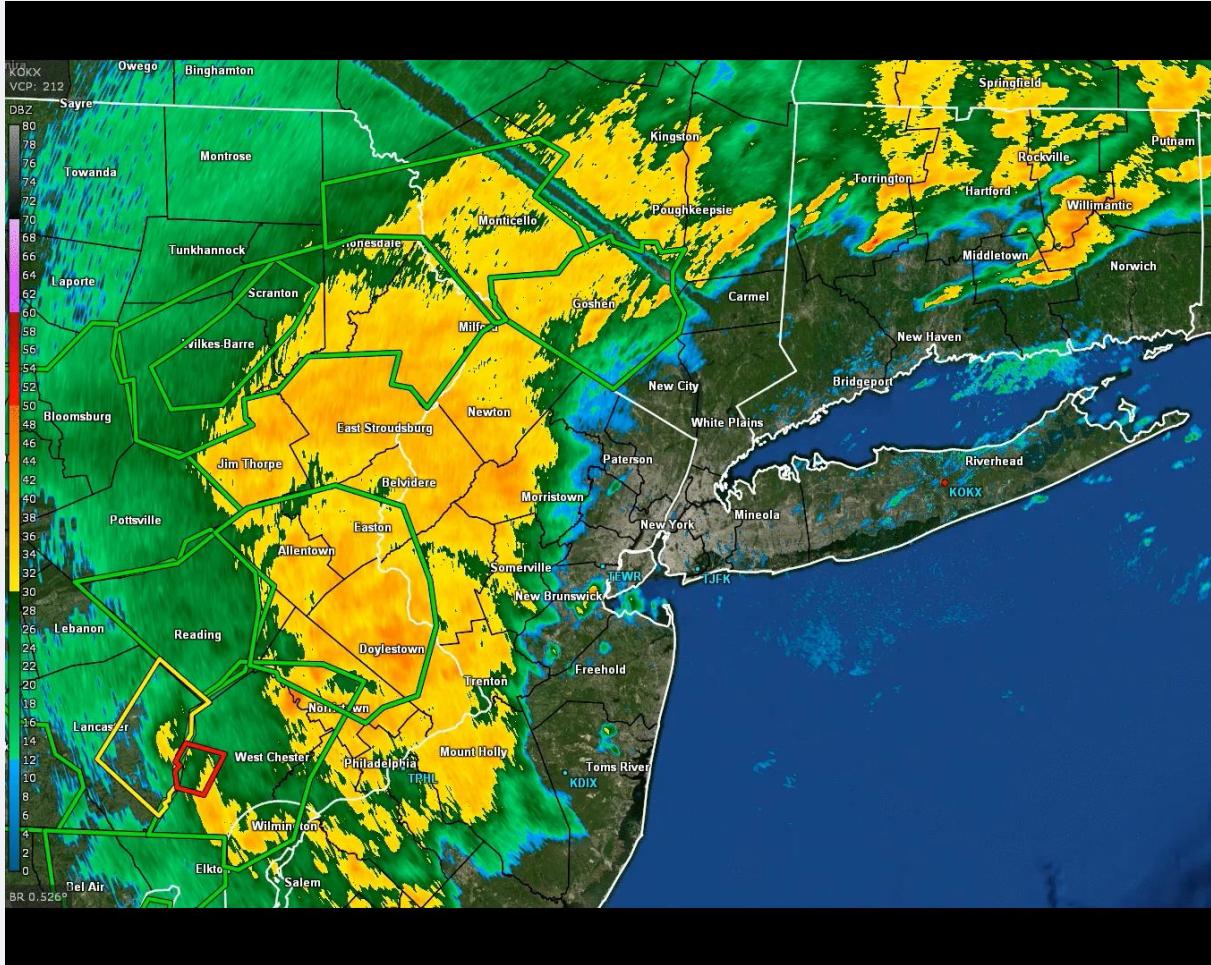
Precipitable water is a measure of moisture

Mixed-Layer CAPE is a measure of instability



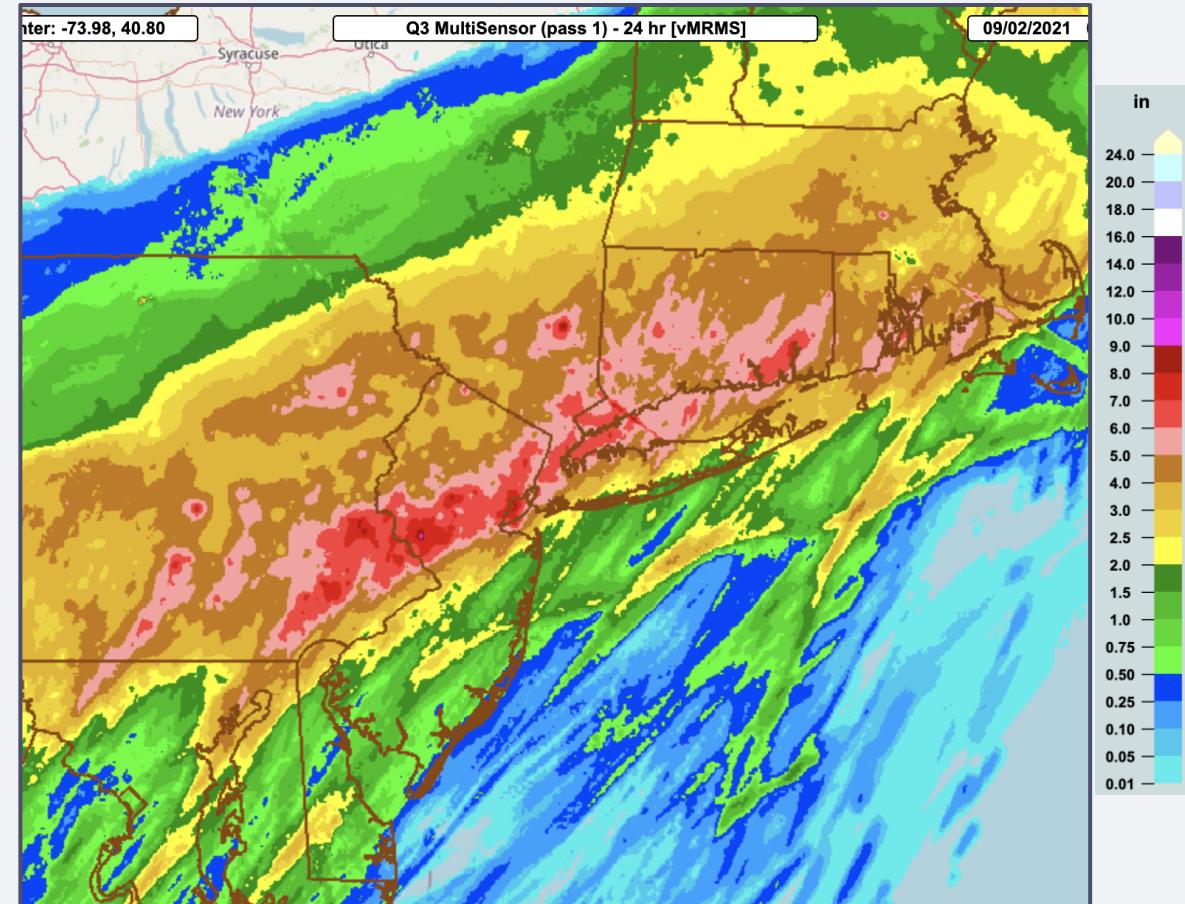
# Radar Loop and 24 hr Rainfall

KOKX Radar & NWS Watches/Warnings



Valid 2030 UTC 1 Sep 2021–0310 UTC 2 Sep 2021

MRMS Q3 Multisensor 24-hour Rainfall



Period Ending 0900 UTC 2 Sep 2021

# Summary & what's next

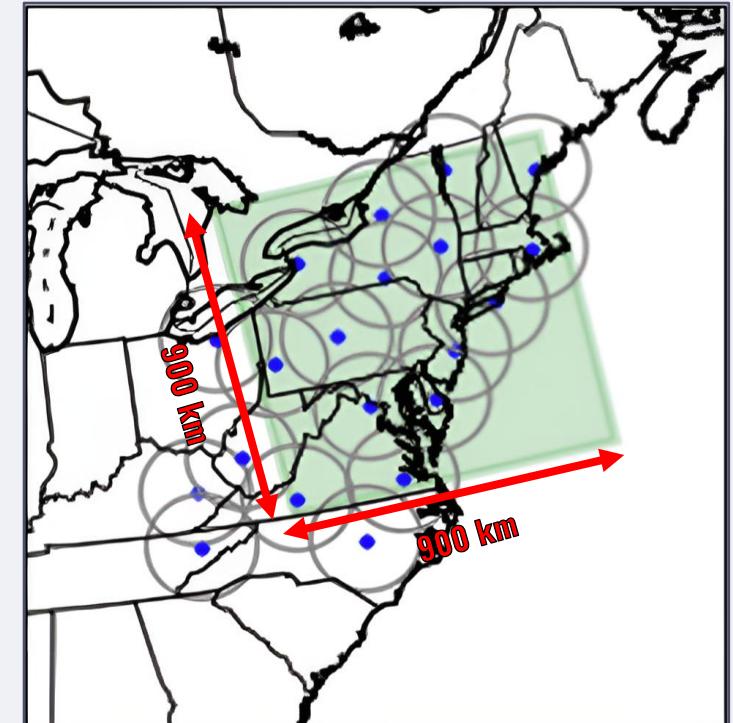
- The confluence of many factors across spatial scales made Ida so destructive
  - Trough and jet streak placement
  - Positive vorticity advection
  - Low-level jet
  - Slow-moving front
  - Ample moisture and instability
  - Training supercells
- **Which of these offers the most predictability for Ida and future extreme rainfall events?**

# **Results with the Warn-on-Forecast System**

# Model: Warn-on-Forecast System (WoFS)

- 18-member ensemble analysis and forecast system
  - 3 km horizontal resolution; 51 vertical levels
  - Based on WRF-ARW
- Movable domain!
- Data assimilation every 15 minutes
  - Radar (WSR-88D)
  - Satellite (GOES)
  - Surface observations
- 6-hour-long simulations initialized hourly 1700–0300 UTC
- Physics and initial conditions (EnKF) differ between members

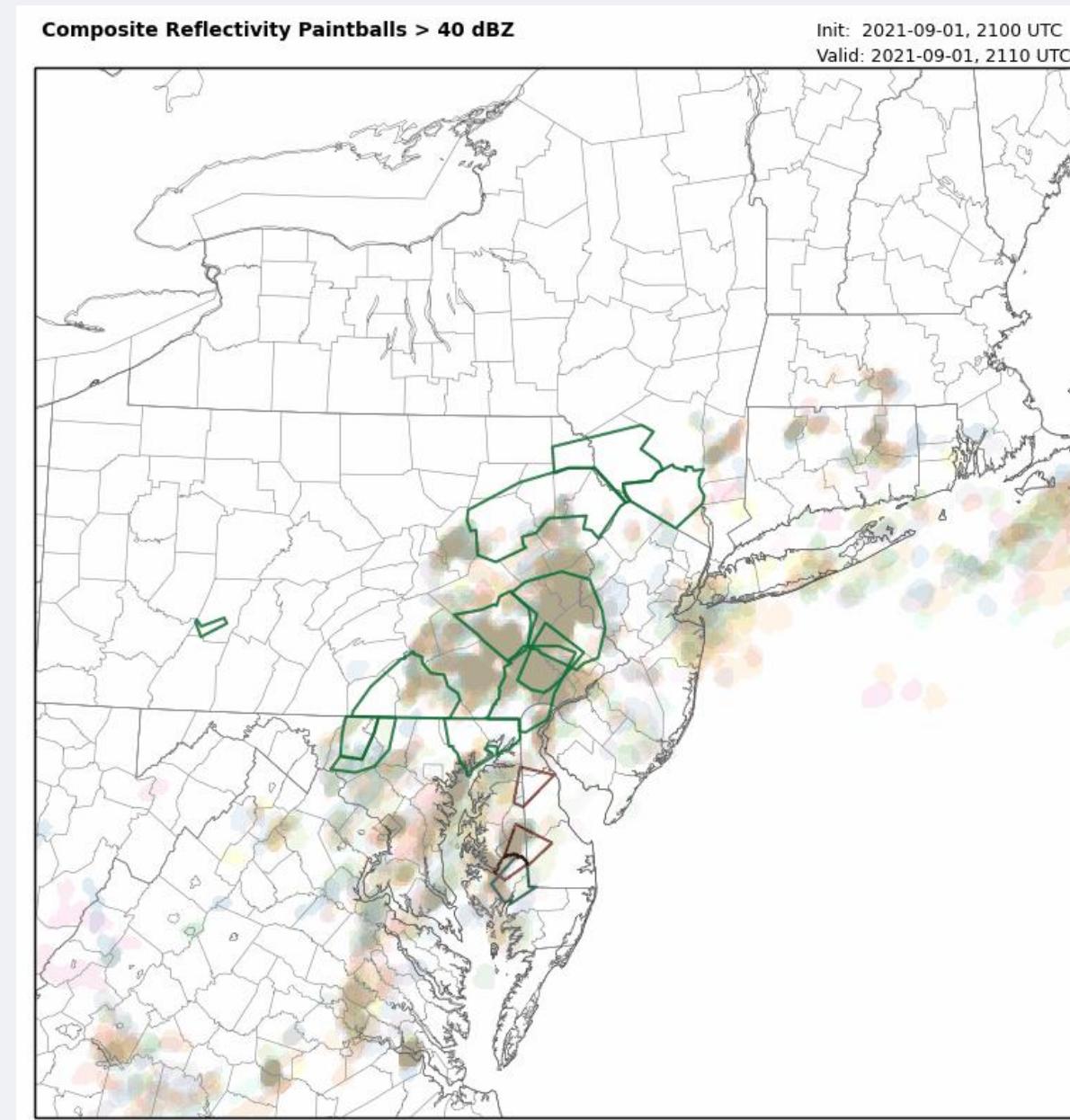
WoFS Grid on 1 Sep 2021



Radar locations (♦) with 150 km range rings

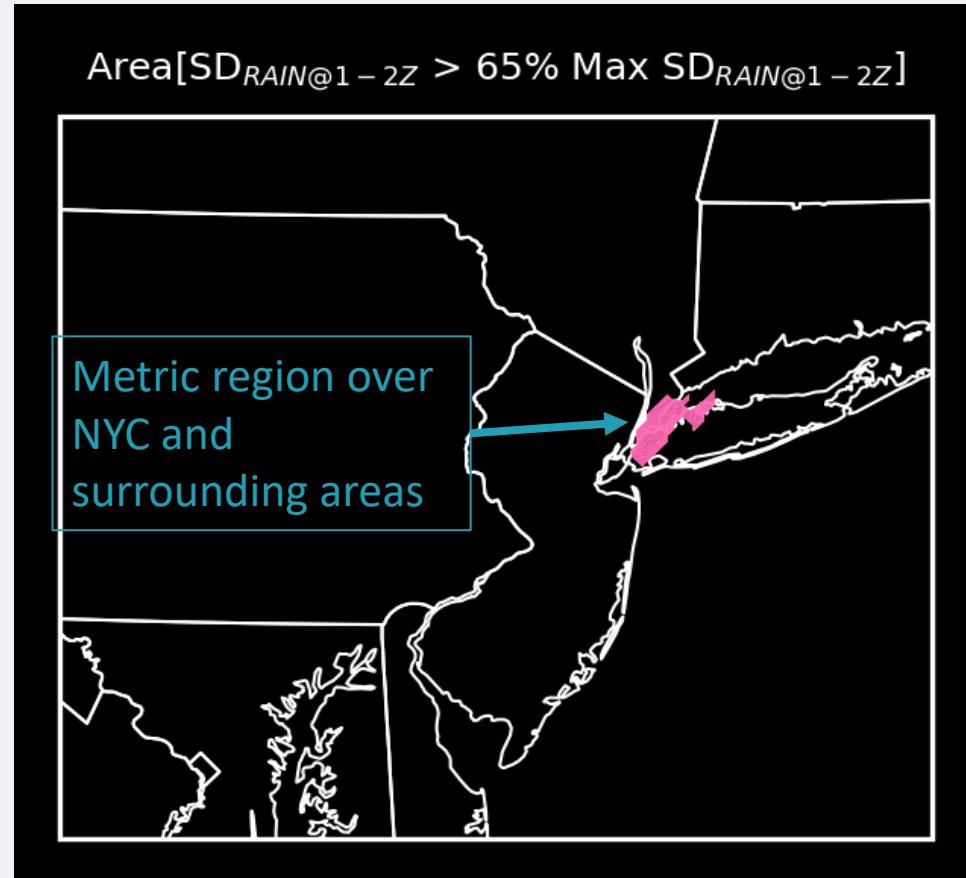
# 2100 Initialization of WoFS

Paintball plot  
of reflectivity  
values > 40  
dBZ



Green = flash  
flood warning  
Red = Tornado  
warning  
Blue = Severe  
Thunderstorm  
Warning

# Region and rainfall metric



“NYC” = region where 72-member ensemble SD of hourly rainfall  $\geq 65\%$  of domain-maximum SD

**WoFS Output:** Hourly initializations from 2000–2300 UTC 1 Sep 2021

- 4 initializations  $\times$  18 members = 72-member ensemble

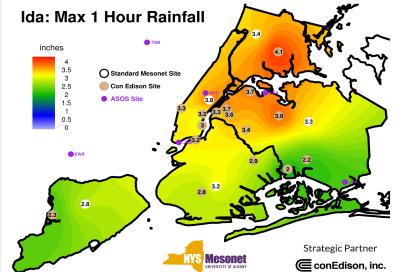
**Rainfall metric:** 0100–0200\* UTC rainfall in NYC = **RAIN<sub>1-2, NYC</sub>**

- Corresponds to observed maximum rainfall rates in Manhattan

\*Results are relatively insensitive to hour chosen

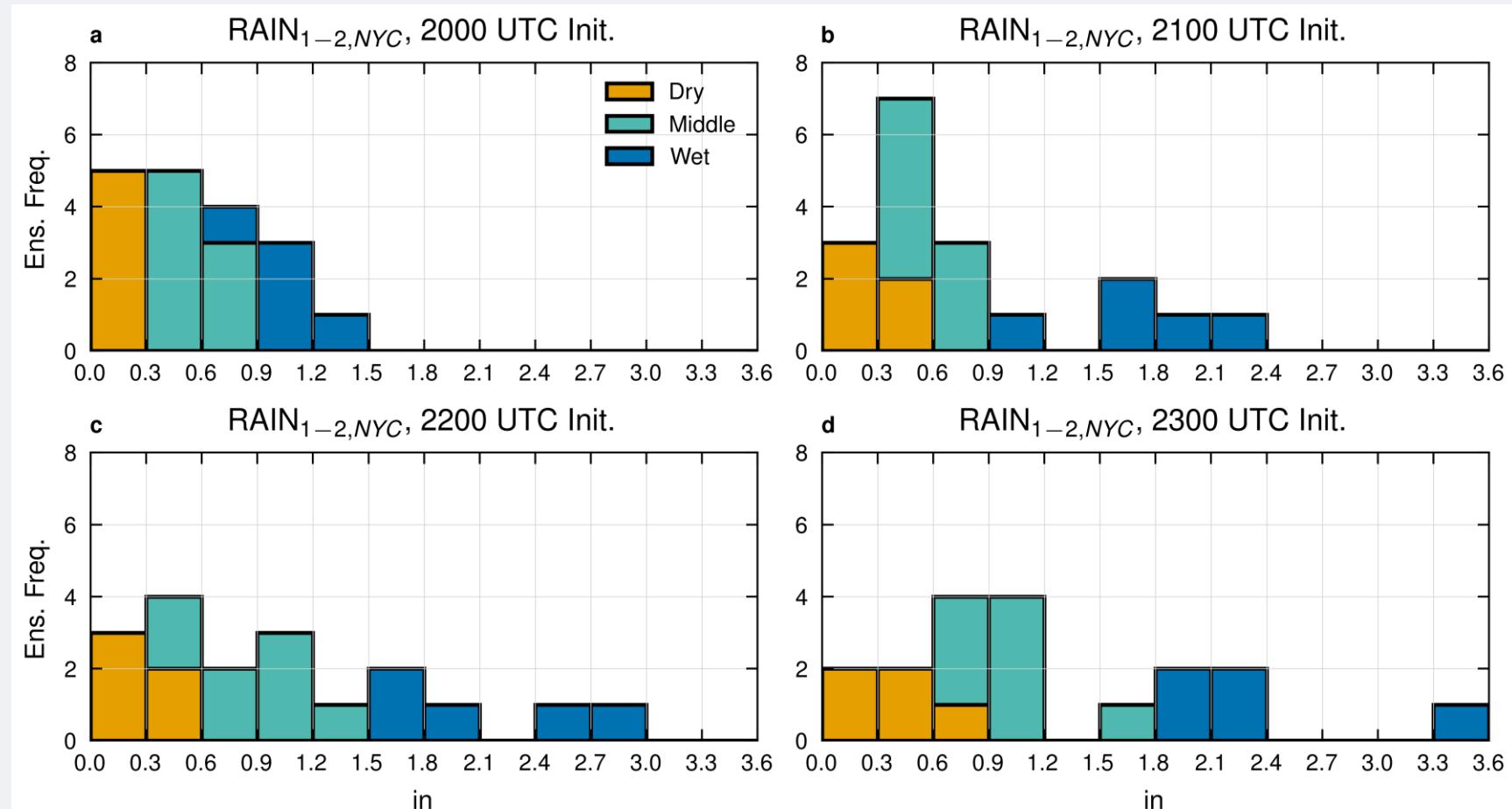
# Huge ensemble spread in NYC rainfall!

Few ensemble members come close to matching observations!



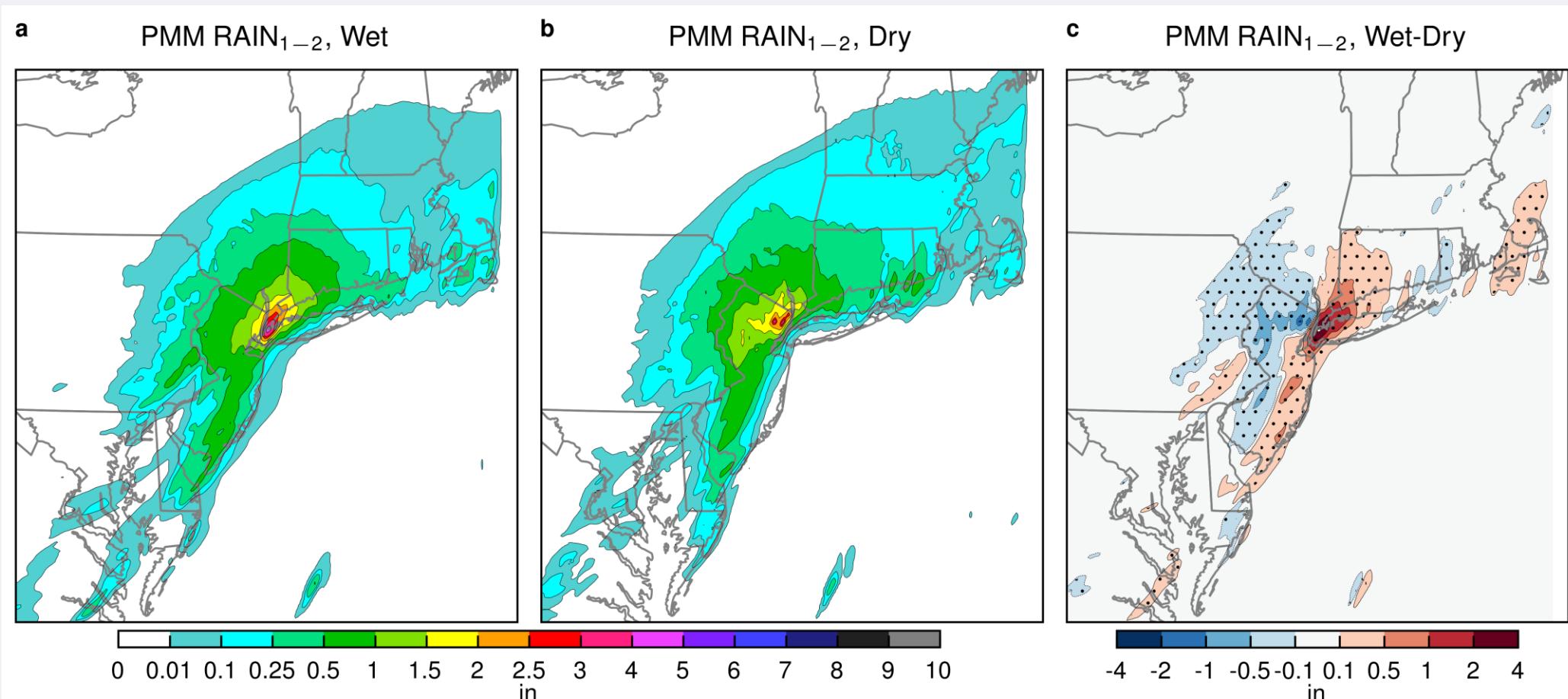
**Approach #1: Compare the 5 wettest and 5 driest ensemble members from each initialization → Wet & Dry Subsets (n=20)**

## Histograms of the 0100-0200 UTC NYC Rainfall in each WoFS initialization



# Dry subset has less rain, shifted NW

Probability Matched Mean\* of 0100–0200 UTC Rainfall

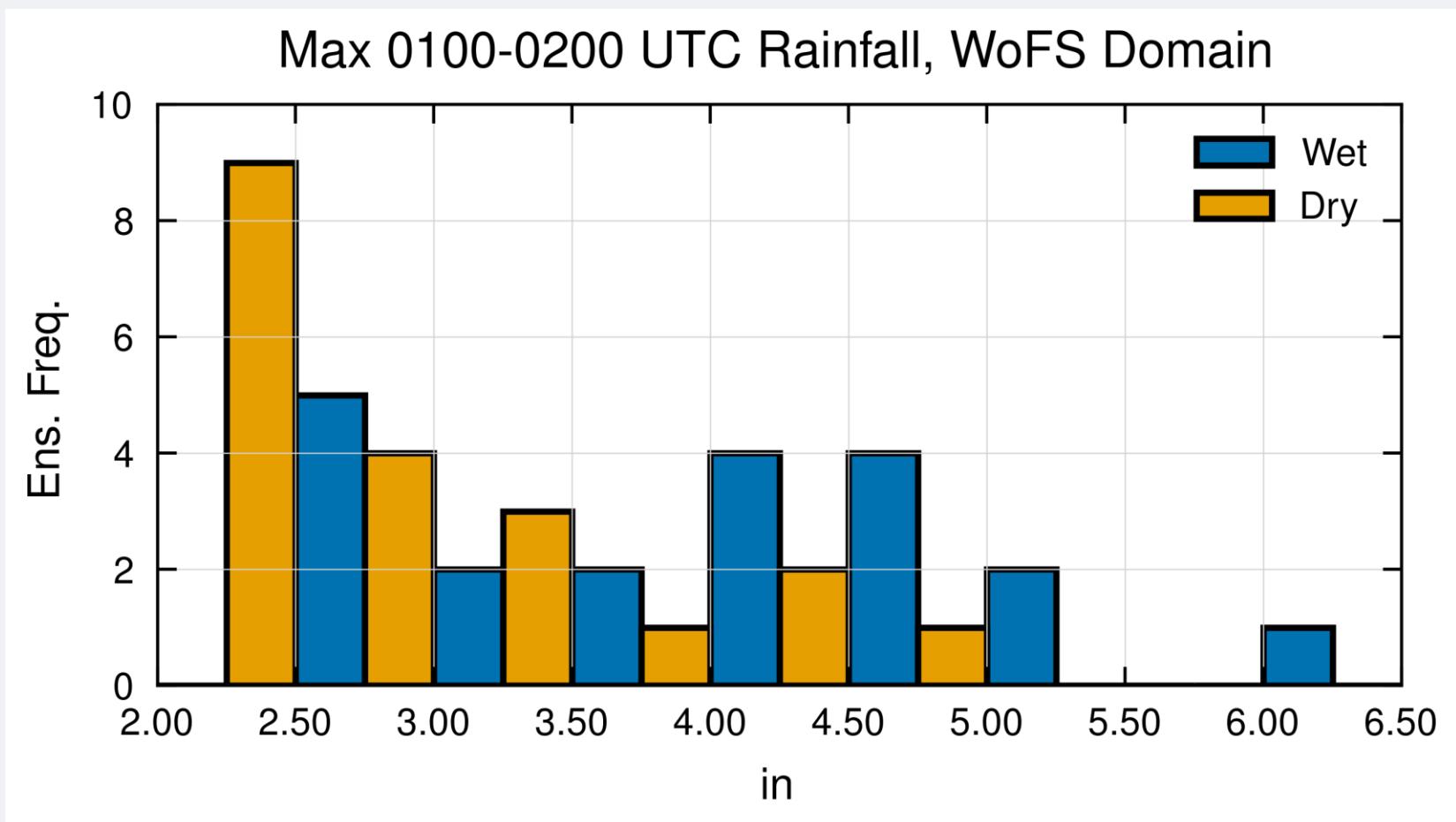


\*Probability  
matched  
mean →  
similar to  
ensemble  
mean, but  
retains  
extreme  
values

Dotted = statically  
significant

# Wet subset has higher maximum rainfall over entire WoFS domain

Histogram of the Max 0100-0200 UTC Rainfall over all ensemble members and WoFS domain



# Ensemble Sensitivity Analysis

Ensemble Sensitivity Analysis = a linear regression of a forecast metric against the atmosphere at an earlier time

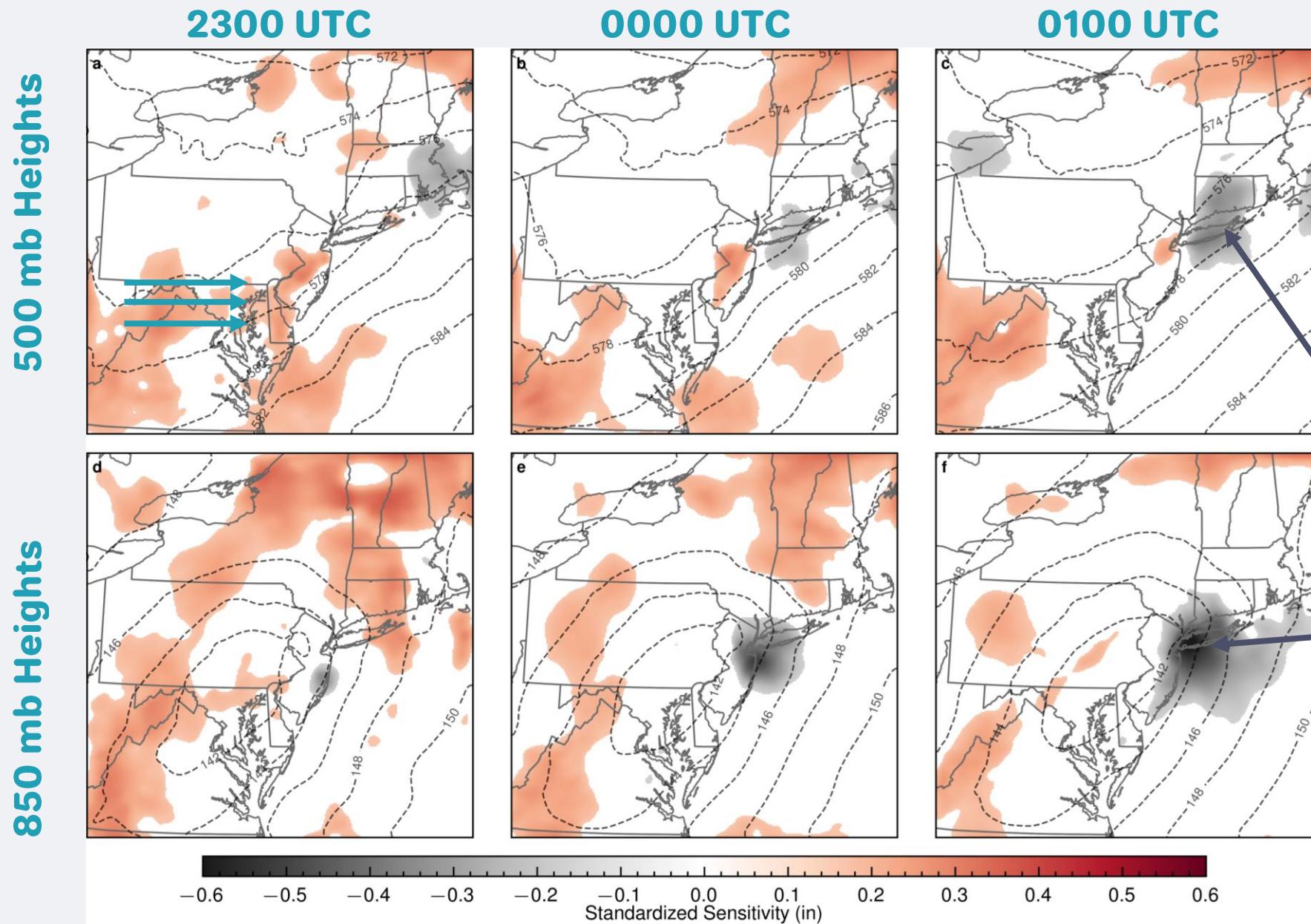
$$\frac{\partial \text{RAIN}_{1-2, \text{NYC}}}{\partial x} = \frac{\text{cov}(\partial \text{RAIN}_{1-2, \text{NYC}}, x)}{\text{var}(x)} \times \text{SD}(x)$$

↓                      ↓                      ↓  
Sensitivity          Slope of linear regression      StDev of predictor

Here,  $\partial \text{RAIN}_{1-2, \text{NYC}}$  (0100-0200 UTC NYC Rainfall) is our **forecast metric**, while  $x$  is our **predictor** (geopotential heights, winds, etc.)

Resulting values = expected change in  $\partial \text{RAIN}_{1-2, \text{NYC}}$  given a one-standard deviation change in  $x$

# Sensitivity of RAIN<sub>1-2, NYC</sub> to Geopotential Heights



Higher heights to N and SW impact **steering flow direction and speed of supercells!**

Likely a sign of **existing convection, not synoptic scale circulation differences!**

I only show values that are statistically significantly different from 0!

# Sensitivity of RAIN<sub>1-2, NYC</sub> to 850 mb Winds

2300 UTC

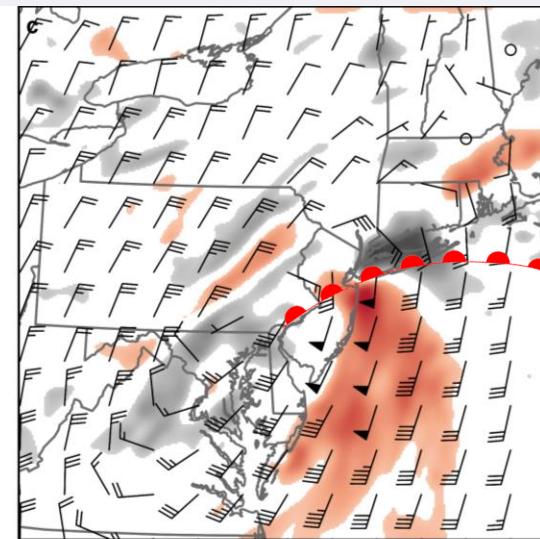
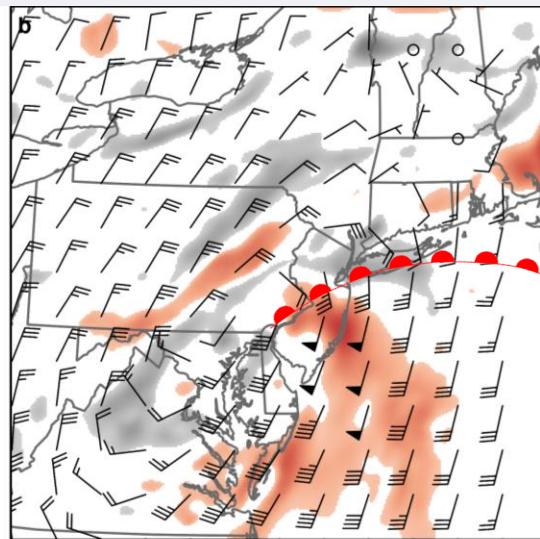
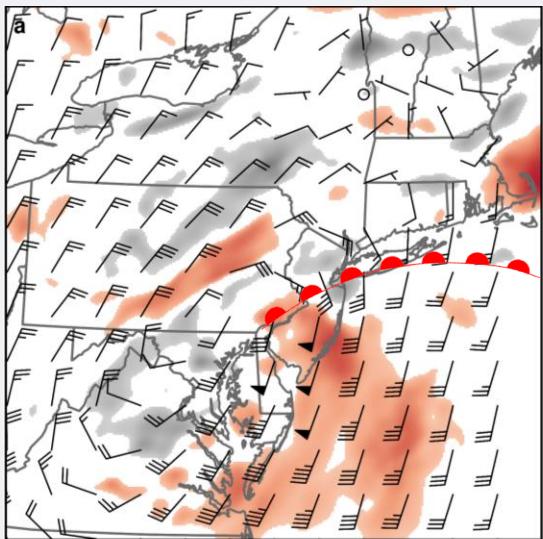
0000 UTC

0100 UTC

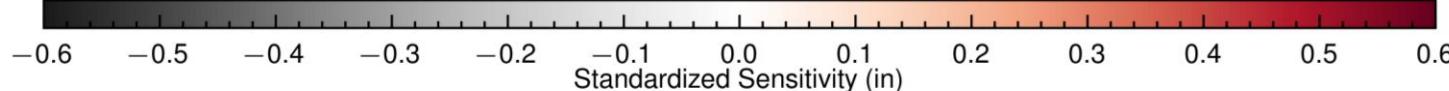
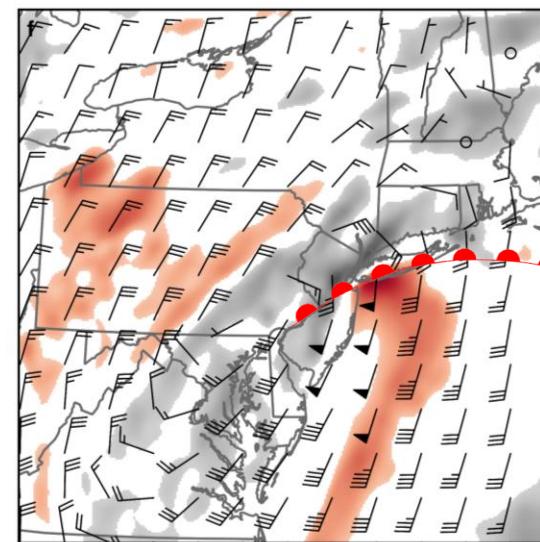
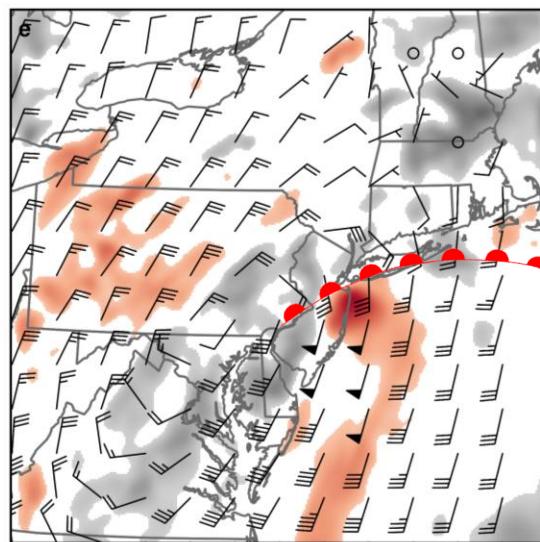
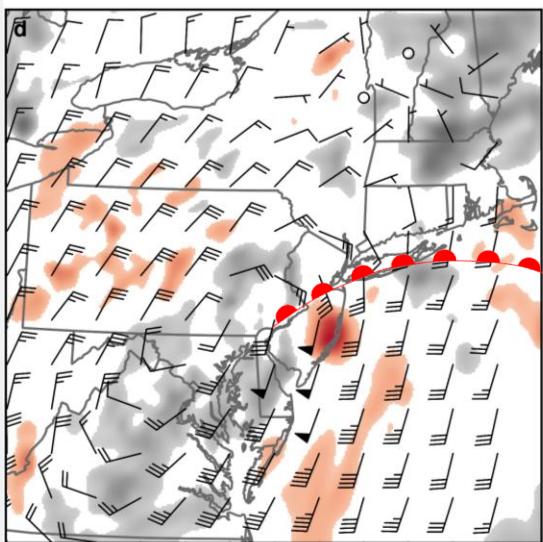
Cyclonic shear↑ =  
frontogenesis↑ =  
Ascent↑ = Rain↑

Stronger low-level  
jet = more  
moisture and  
instability

850 mb Zonal Wind

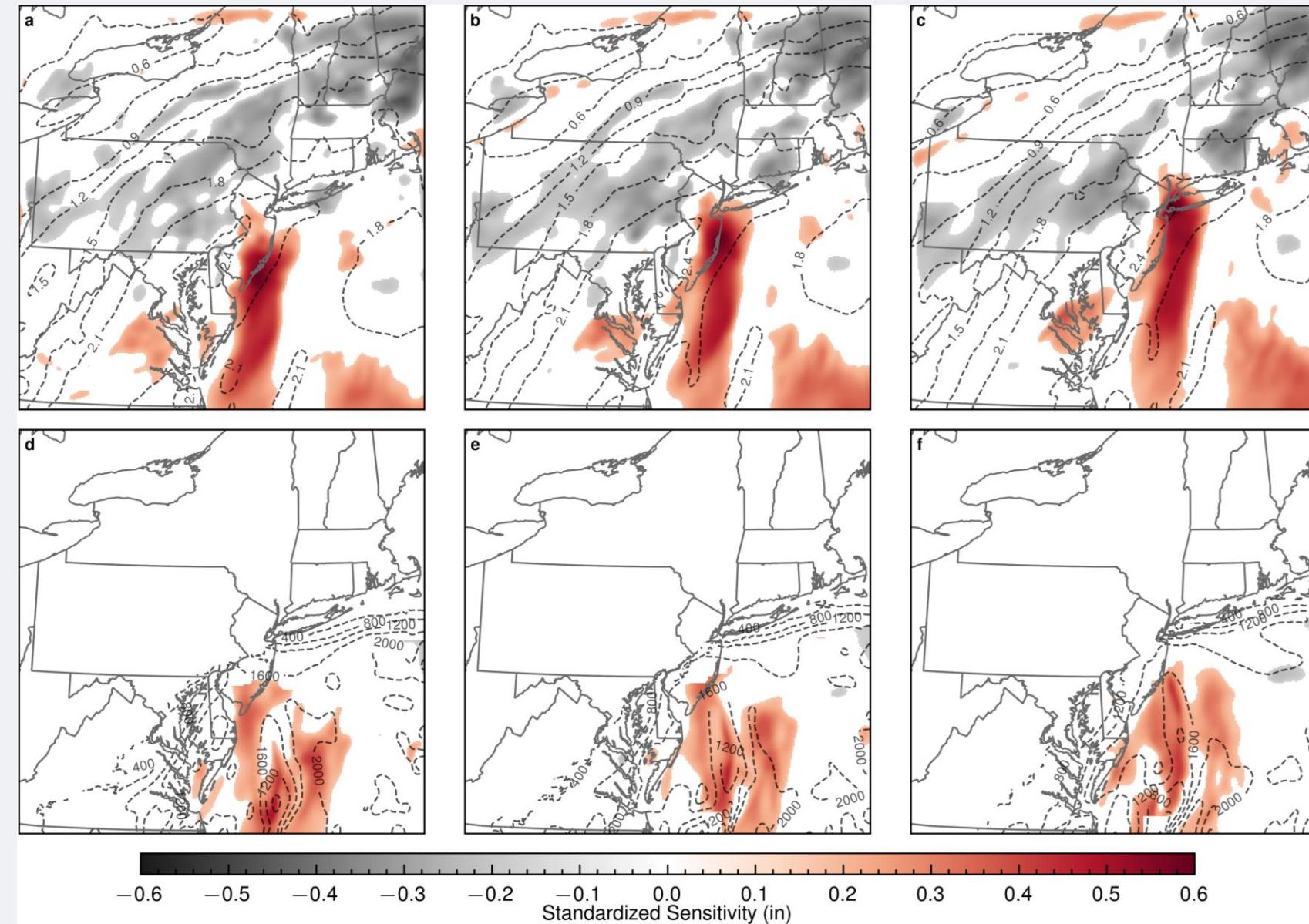


850 mb Meridional Wind



# Sensitivity of RAIN<sub>1-2, NYC</sub> to PW and CAPE

2300 UTC      0000 UTC      0100 UTC



Precipitable water  
to the south  $\uparrow$  =  
RAIN<sub>1-2, NYC</sub>  $\uparrow$

Mixed-Layer  
CAPE  $\uparrow$  = More fuel  
for convection

CAPE: Convective Available  
Potential Energy (Instability)

**Let's look at smaller  
(meso, frontal) scale  
features**

Back to wet vs. dry!

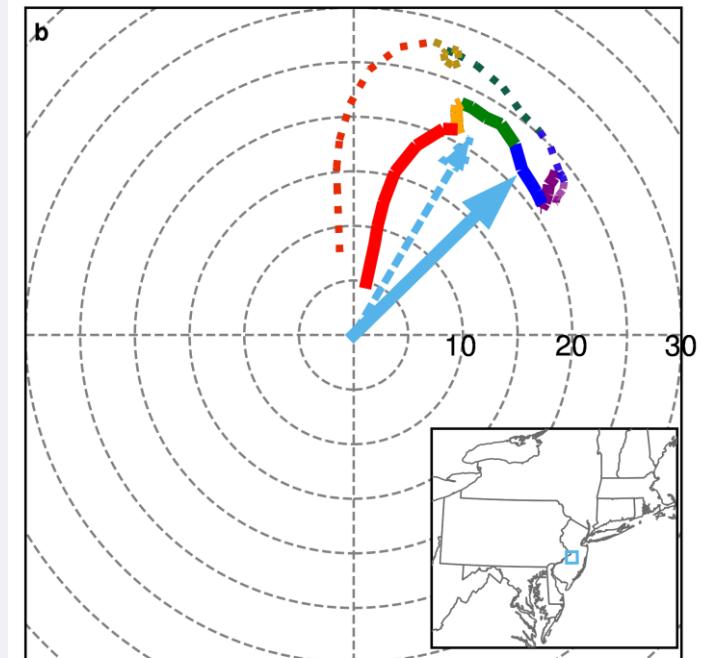
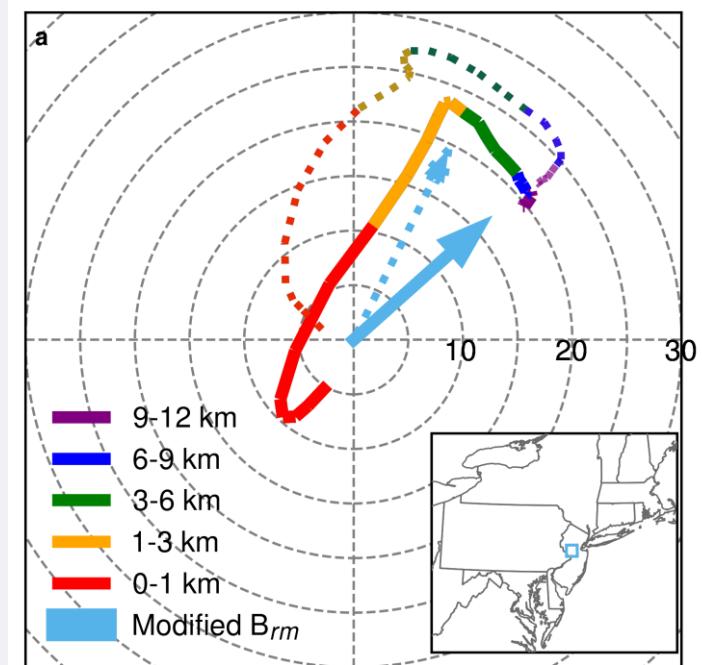
# Hodographs

- Hodographs show how **wind** changes with **height**
- Wet subset has **slower, more eastward moving supercells**
- Our first clue that the **front location** is different in wet vs. dry!

Polar coordinates:  
 $\theta$  = Wind direction  
 $r$  = Wind Speed

Blue arrows are an estimate of storm motion

Dashed = Dry Subset Mean  
Solid = Wet Subset Mean



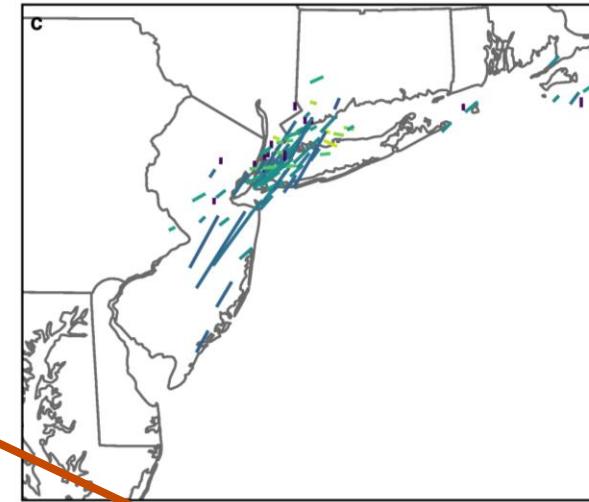
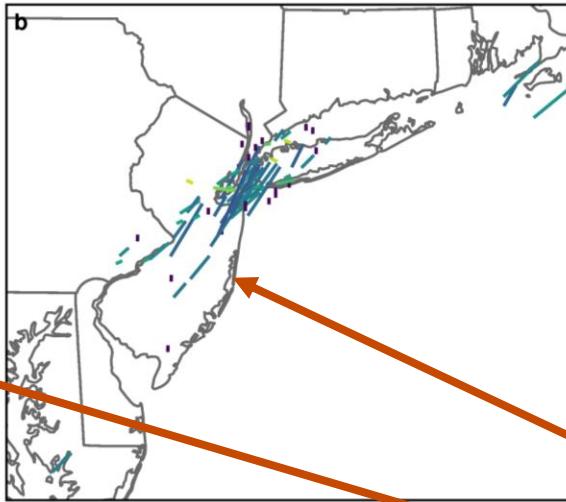
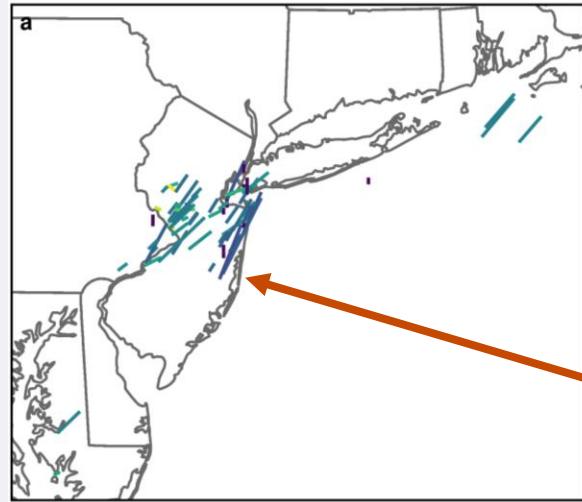
# Wet & Dry UH > 20m<sup>2</sup>s<sup>-2</sup> with Trajectory Angles

2300 UTC

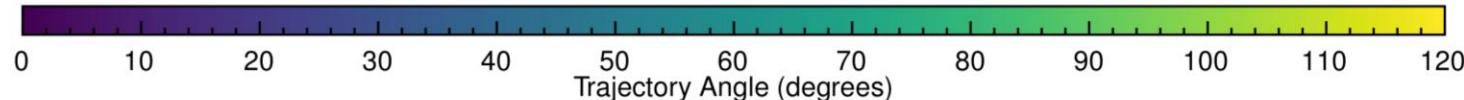
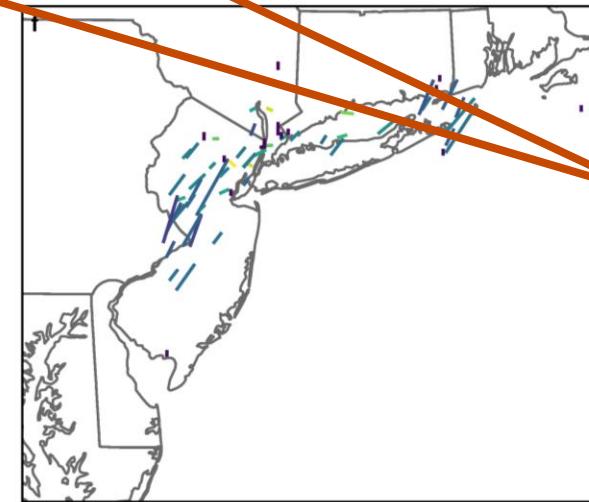
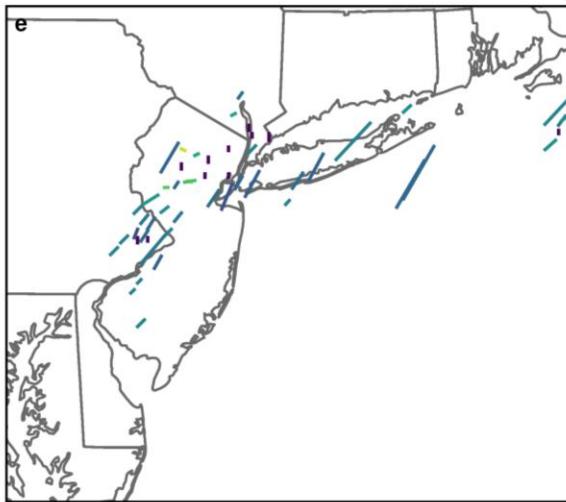
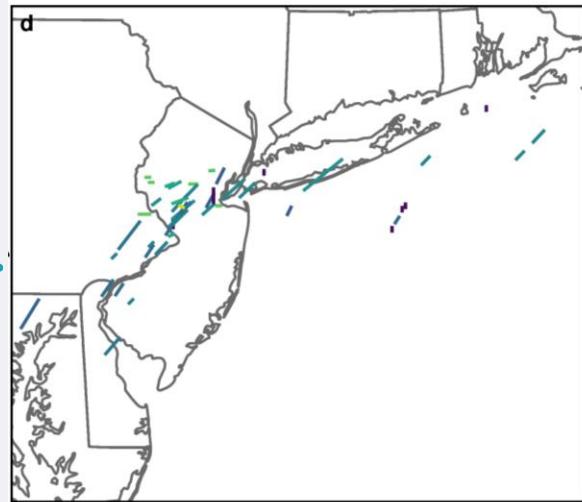
0000 UTC

0100 UTC

Wet



Dry



- Wet subset has greater number and trajectory angle of supercells
- Dry subset missing Jersey Shore storms

UH = Updraft helicity, a measure of “spininess” to track supercells!

# Front Finding Algorithm

- Created a new front-finding algorithm for WoFS using the **thermal front parameter (TFP)** (Schemm et al. 2015)

$$TFP = -\nabla|\nabla\theta_{e,850}| \cdot \frac{\nabla\theta_{e,850}}{|\nabla\theta_{e,850}|}$$

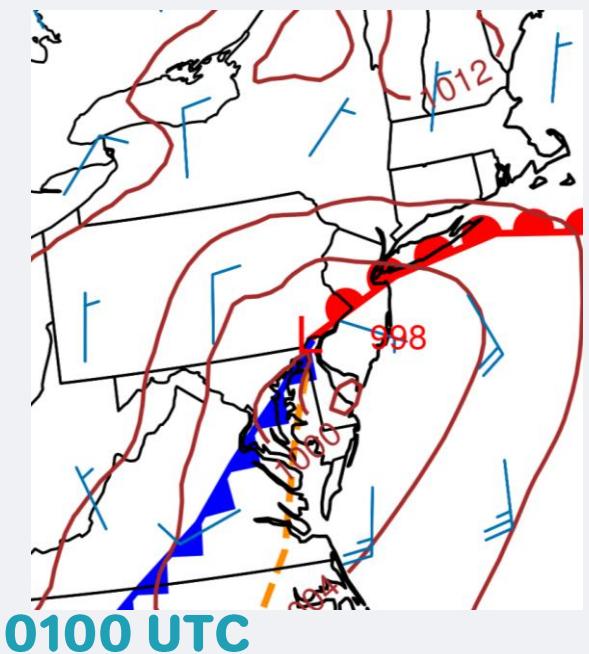
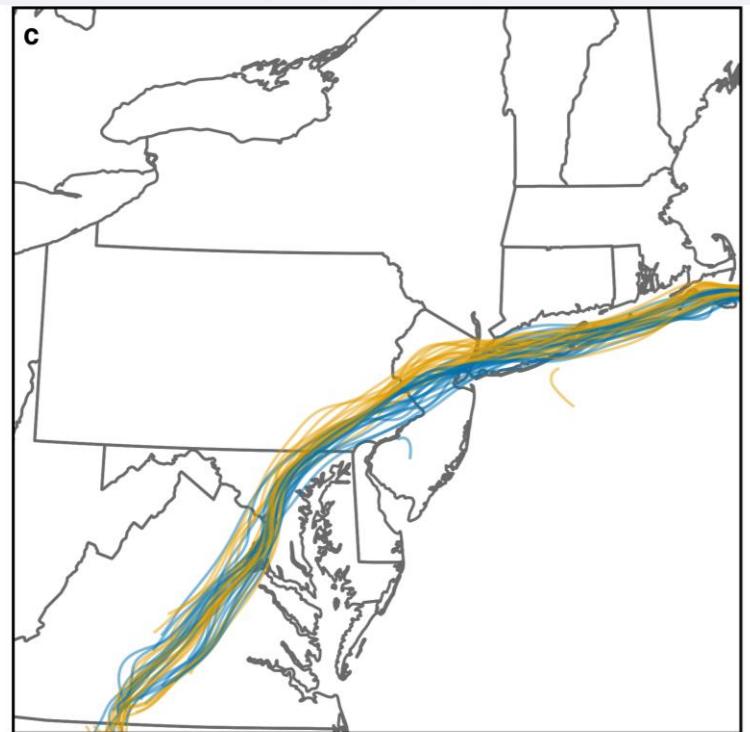
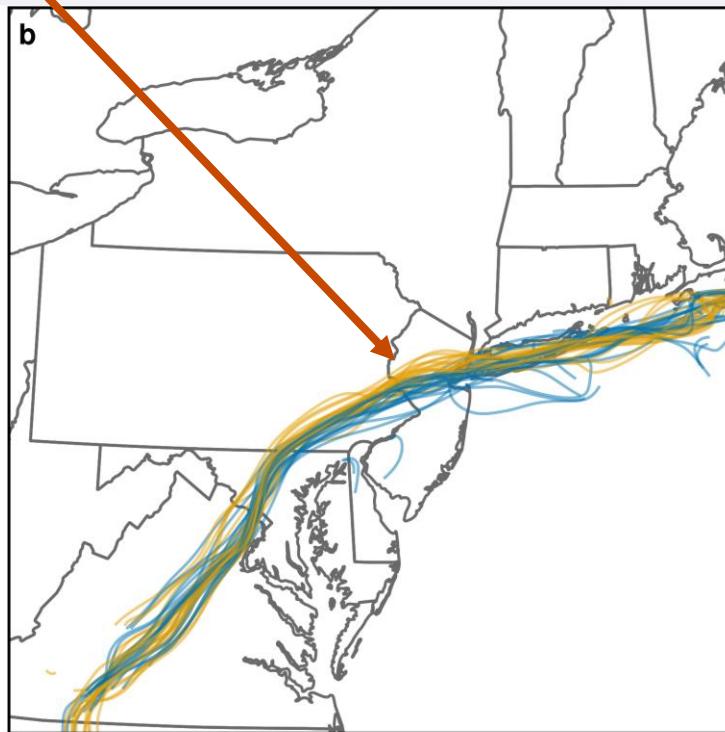
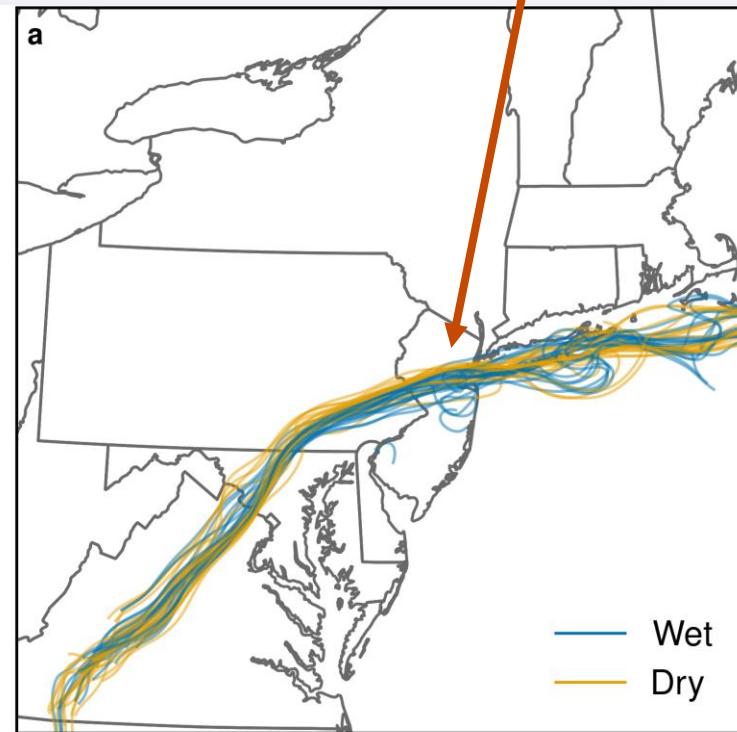
where  $\theta_{e,850}$  is the equivalent potential temperature at 850 hPa

- Fronts occur where  $TFP = 0$  and  $|\nabla\theta_{e,850}|$  is above a certain threshold
- Use **connected component labeling** to find the largest feature, which is the front!

Warm front in wet subset located  
farther south than in dry subset!

2300 UTC

### WoFS Front Locations 0000 UTC



# Take-home Messages

- Hurricane Ida flooding in NYC resulted from a confluence of factors to produce record-breaking hourly rainfall
- It was a **tough forecast**, with huge spread among WoFS members
- The WoFS members that produced the most rain in NYC have:
  - A synoptic setup that favors **slower, more eastward moving supercells**
  - A **stronger low-level jet** to bring moisture and instability
  - A **stronger warm front located farther south in NJ**
- These factors should be considered

# References

- Mossel, C., Hill, S. A., Samal, N. R., Booth, J. F., & Deveneni, N. (2024). Increasing extreme hourly precipitation risk for New York City after Hurricane Ida. *Scientific Reports*, 14(1), 27947. <https://doi.org/10.1038/s41598-024-78704-9>
- Schemm, S., Rudeva, I., & Simmonds, I. (2015). Extratropical fronts in the lower troposphere—global perspectives obtained from two automated methods. *Quarterly Journal of the Royal Meteorological Society*, 141(690), 1686–1698. <https://doi.org/10.1002/qj.2471>