### NOAA Global Systems Laboratory



# Earth System Prediction: Observation Use and Analysis Development for Improved Forecasts

Curtis Alexander
Chief, Assimilation and Verification Innovation Division





### Regional Data Assimilation and NWP

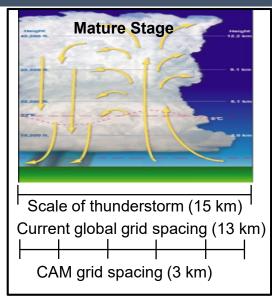


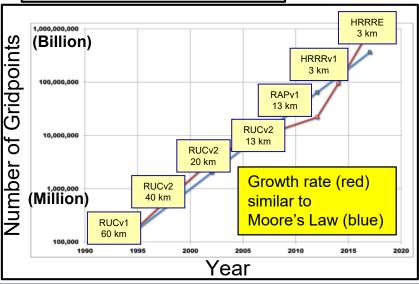
#### Why Convection Allowing Models (CAMs)?

- Captures bulk properties of many hazardous convective weather systems (i.e. rotating updrafts)
- Permits more accurate forecasts of weather conditions in which such hazardous storms may occur
- Employed most often in regional or limited area modeling framework

#### Why Limited Area Models (LAMs)?

- Reduces the compute expense and complications associated with a global CAM
- Enables rapid data assimilation/cycling with lower lowlatency forecasts than global models
- Emphasizes shorter-range (hours/days) prediction



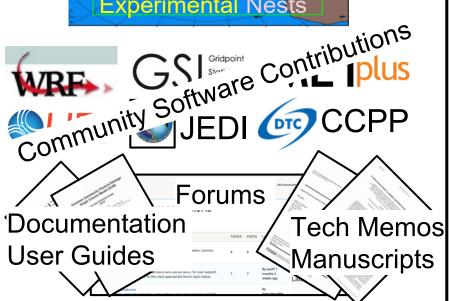


### Regional Modeling Excellence in GSL



#### **Quality of Research**





#### **Relevance to Communities**

Aviation/Transportation Weather Hazar Tactical and Strategic Planning (0-8 hrs FAA, Airlines, Aviators, NCAR, MIT/LL,

Severe Convective Weather
Warn on Forecast (0-2 hrs)
Severe Weather Watches/Discuss
Severe Convective Outlooks (1-2
SPC, NSSL, NWS

Hydrology and Quantitative
Flash Flood Watches/Meso
National Water Model Forci
Heavy Rain/Snow Outlooks
WPC, OWP, GLERL, PSL,

#### Renewable Energy

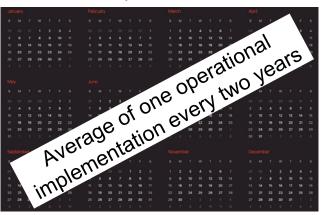
Wind and Solar Power Generation (0-18 Next Day Decision Support (24-48 hrs) Power Authorities, Energy Companies, A

Composition, Air Quality and Health Wildfire Smoke Concentrations (1-2 day IMET, ARL, CSL, NWS

# Organizational Performance

From 2012-2020:

Five RAP operational transitions Four HRRR operational transitions



With each transition:
Added DA/model sophistication

Increasing R2O efficiency

Doubling of R2O capacity every two years "GSL's Law"

### Data Assimilation: GSL Making Forecasts Better

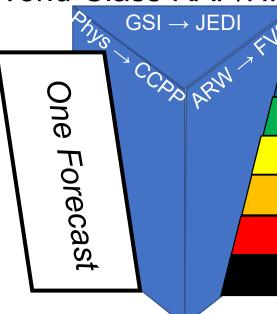


Seamless collaboration across divisions within GSL, CIs, OAR labs, NWS and external organizations

Doubling of DA staff to ~15 in last five years



World-Class RAP/HRRR → Next Generation RRFS (UFS)



Aviation/Transportation(AWRP)

Hydrology (AQPI, JTTI, DSUP)

Renewable Energy (ASRE)

Health/AQ (JPSS, DSUP)

Oceans/Tropics (GOES, JPSS)

Severe Weather (WoF, VSE)

T20 (UFS R20, JTTI, DSUP)

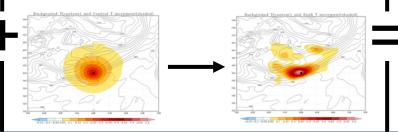
Observations and Impacts (Eric James)



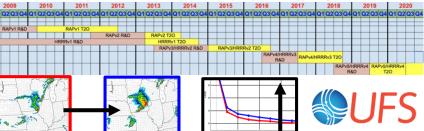




Analysis Development (Terra Ladwig)



Improved Weather Prediction (Steve Weygandt)



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# Observations & Their Impacts

Eric James
Senior Research Associate, Assimilation and Verification Innovation Division

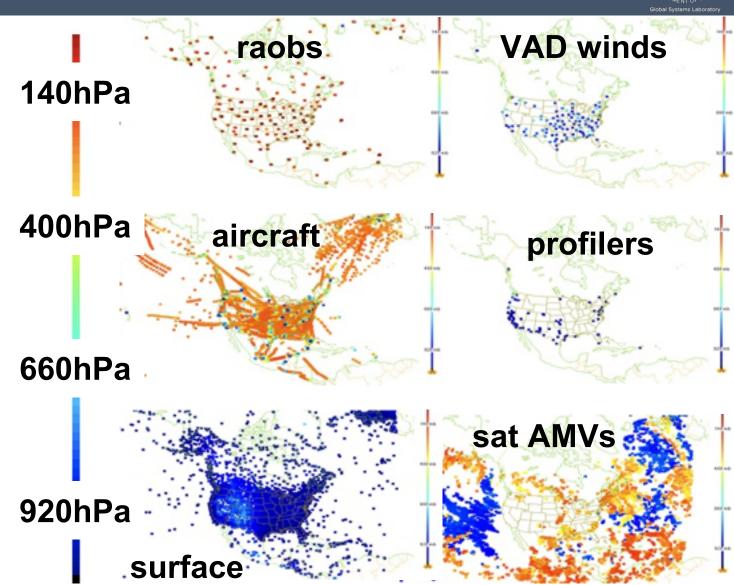




# Weather Observations for Analysis/Fcst Systems



- Effective rapidly-updating analysis/forecast systems depend critically upon optimal use of observations:
  - Verification
  - Data Assimilation
- Quantifying obs impacts ensures an effective data assimilation system, and identifies opportunities for expansion.



# Why run rapidly updating NWP models?



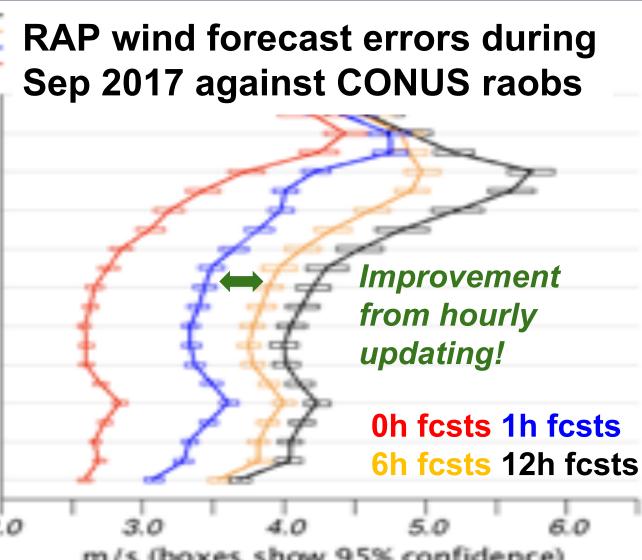
#### Frequent updates allow:

- Use of the most recent weather observations for improved forecasts.
- Updated information for decision makers.

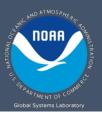


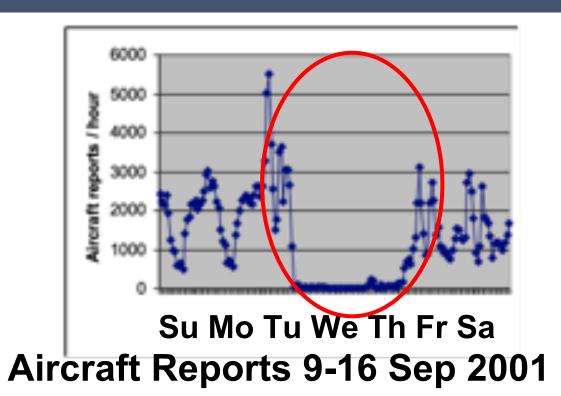






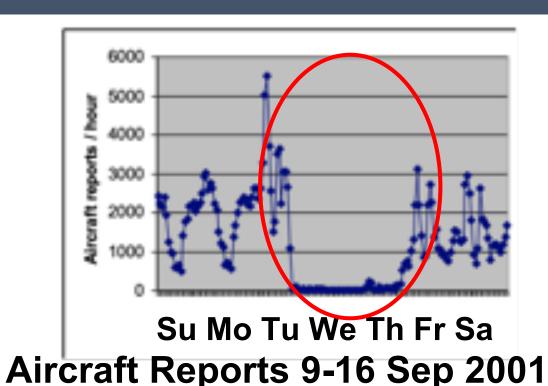
### Why run rapidly updating NWP models?





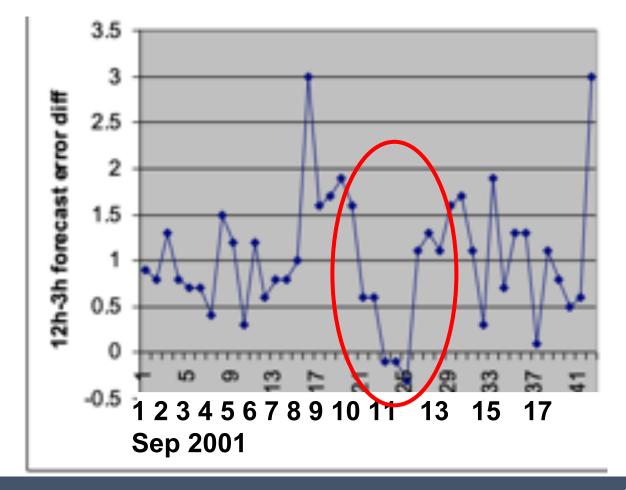
### Why run rapidly updating NWP models?





Without any aircraft reports, 3h forecast errors are just as large as 12h forecast errors!

250hPa wind forecasts: difference in error: 12h fcst err - 3h fcst err



### Monitoring and Quality Control of Observations



Observation quality monitoring carried out via realtime **O minus B** statistics (**O**bservation minus **B**ackground)

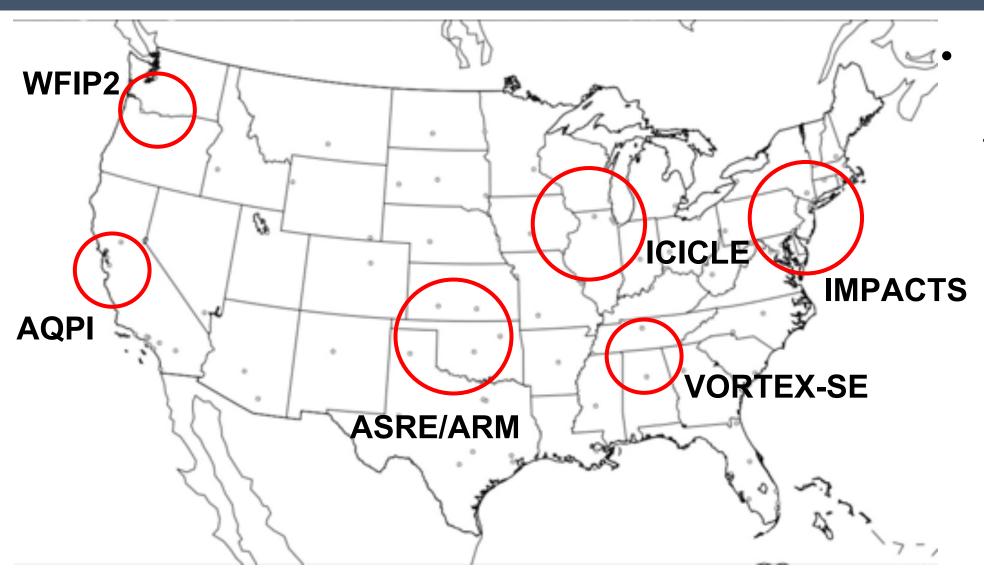
Differences over time are used to construct dynamic lists for use or rejection of observations, for surface mesonet and aircraft observations.

**Example:** temperature bias (bs\_T) exceeding 2.0 K warrants rejection:

aircraft temperature bias > threshold: tail temp reject obs from this aircraft. Number bias ;tail bs\_T Std\_T bs\_S\_\_cd\_S bs\_D std\_D std\_W rms\_W bs\_RH **FSL MDCRS** Ν errors 0000250 T - - 7103 ----- 1032 0.0 4.5 00000251 - - 6752 ----- 211 0.0 CNJCA117 W - 7124 -----0.0 16.1

### Collaborating for Targeted Observations





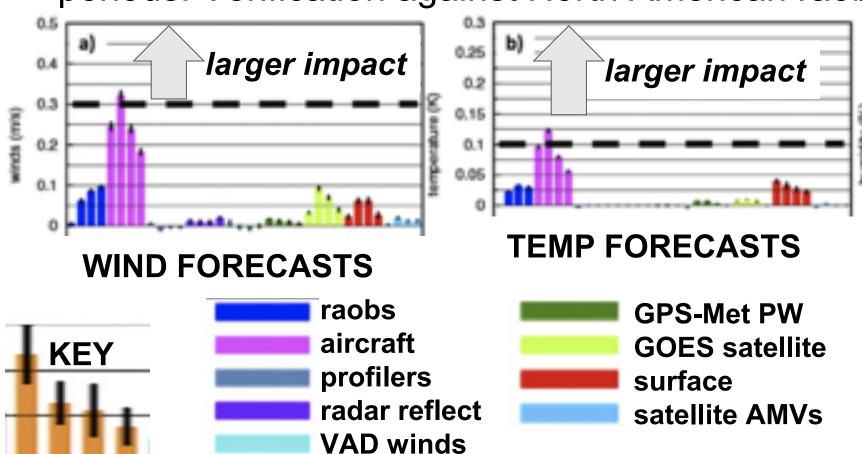
Field campaigns since 2015 have focused on:

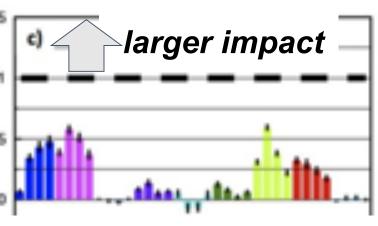
- Wind Energy
- Landfalling
   Pacific storms
- Aviationhazards(convection / icing)
- Severe storms

# Observing System Experiments (OSEs)



 RAPv3 data denial experiments carried out for three multi-season 10-day periods. Verification against North American raobs.





Dashed lines: 25% error

**RH FORECASTS** 

reduction

**Significance:** 1 standard error (67% significance)

James and Benjamin 2017

9 12h forecasts

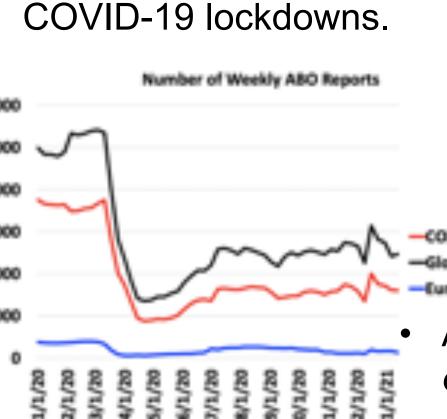
# What was the impact of the COVID-19 decrease in aircraft observations?



 We designed a set of partial data denial experiments which exclude ~80% of aircraft observations, similar to what happened at the peak of

winds (m/s) RMSE

0.1



Wind forecast impacts for winter 0.3 (Feb 2019) 0.2

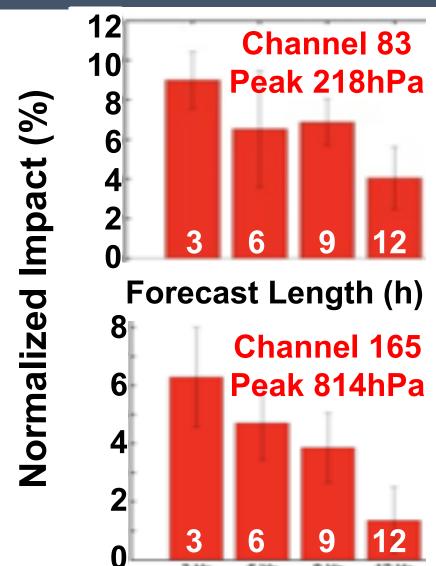
> raobs all aircraft 80% of aircraft obs

Across seasons / variables, we find 30% higher errors without any aircraft obs and 12% higher errors without 80% of aircraft obs

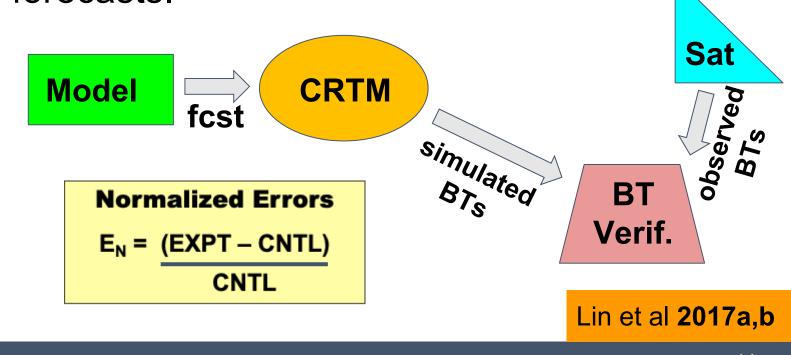
James et al 2020

### Satellite Radiance Direct Broadcast Impacts in RAP





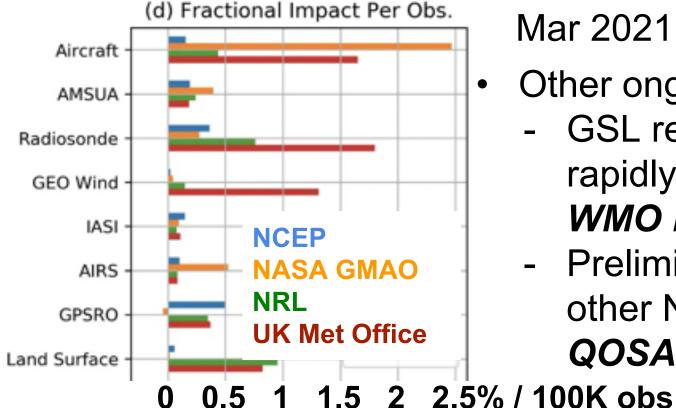
- 4-week retro test 1-28 Sep 2017.
- Hourly verification of brightness temp forecasts against *CrIS* observations reveals significant radiance impacts in RAP for short range forecasts.



# Ensemble Forecast Sensitivity to Obs Impacts



- OSEs (data denial experiments) quantify impact of one selected change upon forecast skill; FSOI/EFSOI quantifies impact of all assimilated obs.
- GSL was involved with EFSOI installation in GFSv16, implemented on 22



Mar 2021 (collaboration with *NCEP/EMC*)

- Other ongoing collaborations:
  - GSL represents a unique perspective with rapidly-updating regional obs impacts at *WMO Impacts Workshop* every 4 years.
  - Preliminary coordination meetings with other NOAA labs within the *NOAA QOSAP program* (recommendation C4.6).

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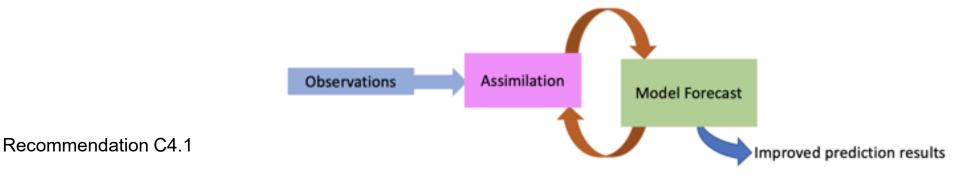
# Development of Analysis Systems

Terra Ladwig, Chief, Data Assimilation Branch & Research Scientis Assimilation and Verification Innovation Division





- Short-term hazardous weather forecasts rely heavily on the quality of the initial conditions.
- In order to capture high temporal observations rapid assimilation is required, as demonstrated in the previous sub-section.
- GSL has developed and deployed two types of data assimilation techniques.
  - a. Advanced data assimilation, specifically 3DEnVAR and EnSRF
  - b. Non-conventional state or tendency specification



### 3D Ensemble-Variational Data Assimilation



3DEnVar uses both static and ensemble background error covariances.

Analysis increments for a single observation near a frontal boundary highlight how ensemble flow dependent background error covariances can improve the distribution of observational information.

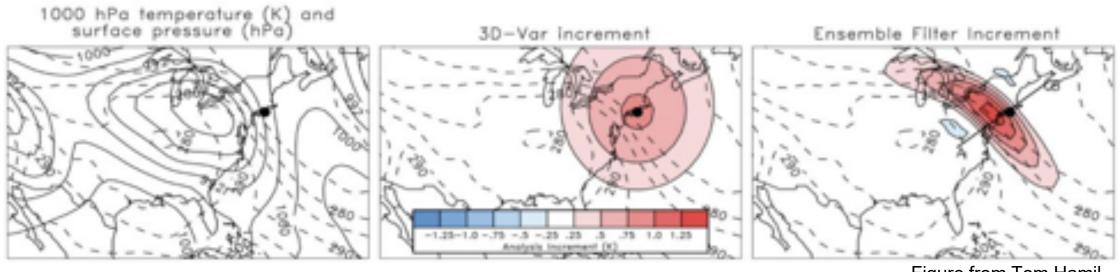
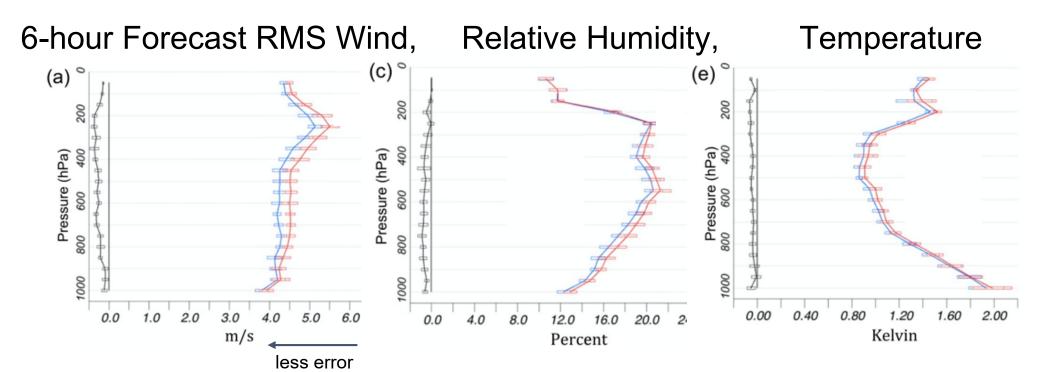


Figure from Tom Hamil

### 3D Ensemble-Variational Data Assimilation



- 3DEnVar is very successful for regional hourly data assimilation in RAP/HRRR and has superior skill over purely a 3D-variational analysis (shown below).
- Ensemble background error covariance data from the global system, GDAS, is effectively used with static errors in the RAP 3DEnVar.



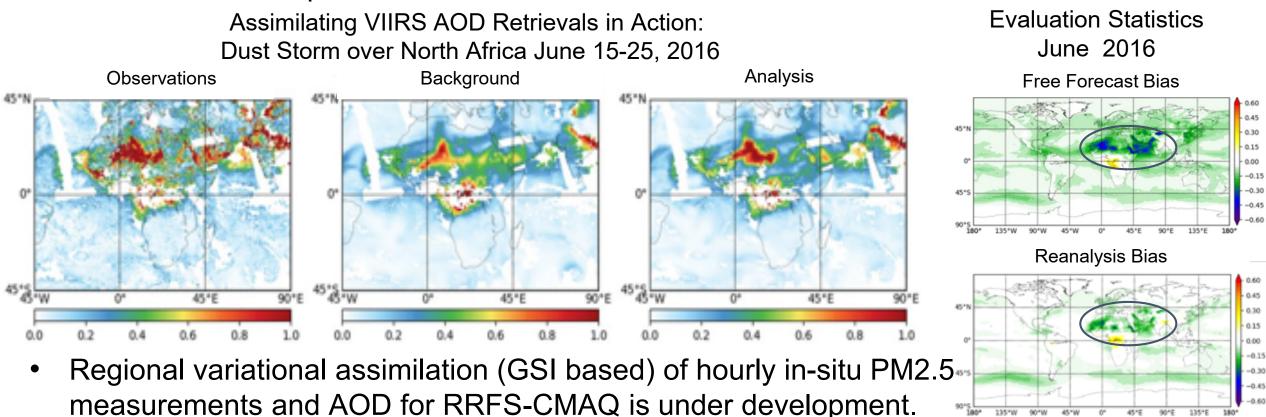
3DVar 3DEnVar Difference

Hu, M., et al., 2017.

### Assimilation System for Aerosol Analysis and Forecasting



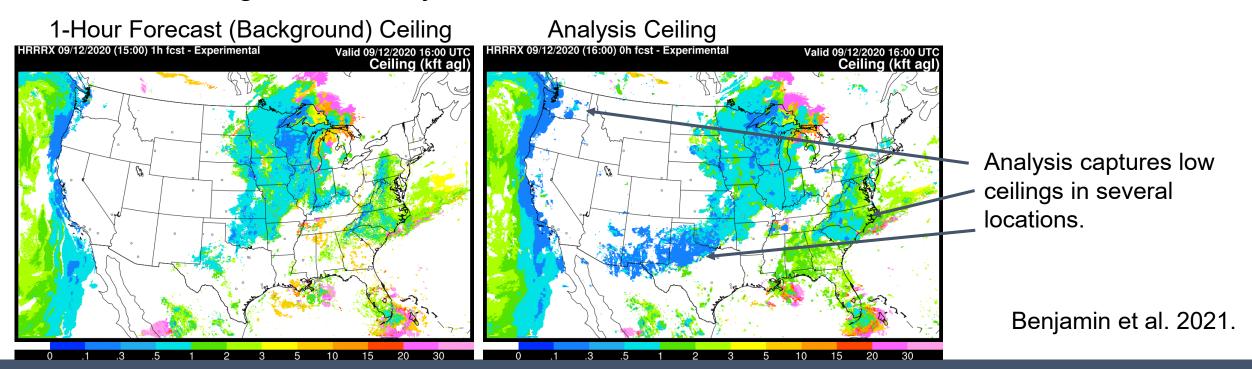
- Satellite based observations of Aerosol Optical Depth (AOD) at 550 nm are assimilated with JEDI software and enable aerosol prediction.
- High AOD indicates large aerosol concentrations that can impact radiation, cloud and rain formation, and air pollution.



### Stratiform Cloud-Hydrometeor Assimilation



- Accurate cloud and precipitation initialization is fundamental to short-range prediction systems such as RAP & HRRR.
- The stratiform cloud-hydrometeor assimilation improves retention of observed cloudy and clear 3D volumes in subsequent model forecasts.
- Cloud ceiling and visibility forecasts have more skill due to the cloud assimilation.

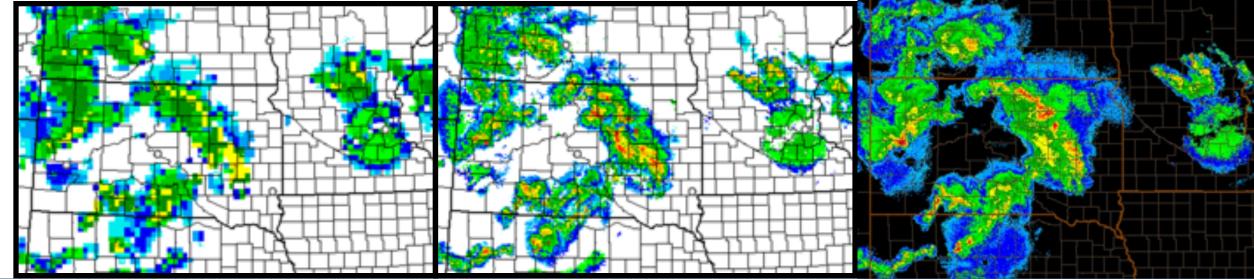


# Latent-Heating Based Precipitation Assimilation



- Assimilation of precipitation related observations, especially for convection is essential to the success of HRRR/RAP short-term forecasts.

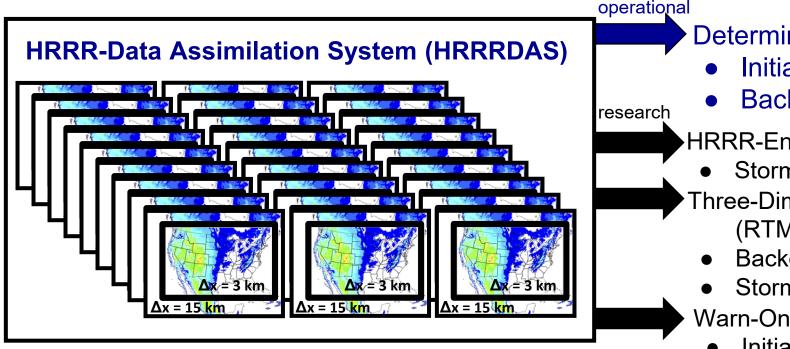
Reflectivity Analysis
Without Latent-Heating With Latent Heating Observations



### Convective-Scale Ensemble Based Assimilation



- Ensemble covariances provide temporal and flow dependent information, which is especially important near small scale nonlinear phenomena.
- Convective-scale ensemble assimilation has improved initial conditions for the HRRR and is a foundation for future implementations.



Deterministic HRRRv4

- Initial conditions
- Background Error Covariances

HRRR-Ensemble (HRRRE)

- Storm-Scale Ensemble Forecasts
- Three-Dimensional Real-Time Mesoscale Analysis (RTMA-3D)
  - Background Error Covariances
  - Storm-Scale Analysis Uncertainty

Warn-On-Forecast (WoF) System

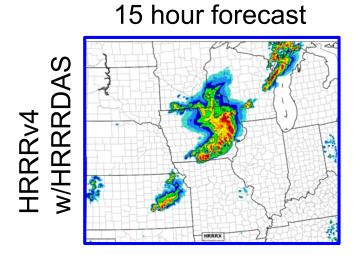
- Initial Conditions
- Boundary Conditions

Recommendation C4.2

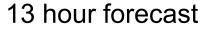
### HRRRDAS Initial Conditions Improves HRRRv4

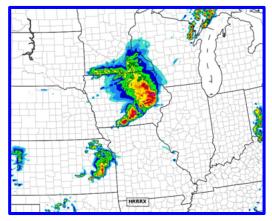


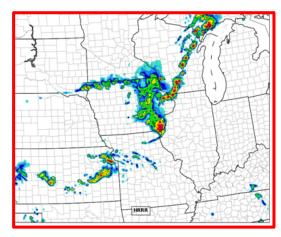
#### Case Study 18 UTC 10 August 2020



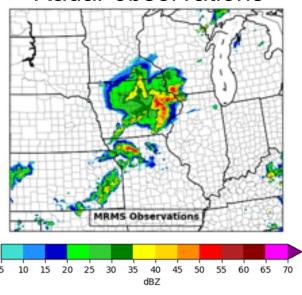
HRRR<sub>v3</sub>







#### Radar observations

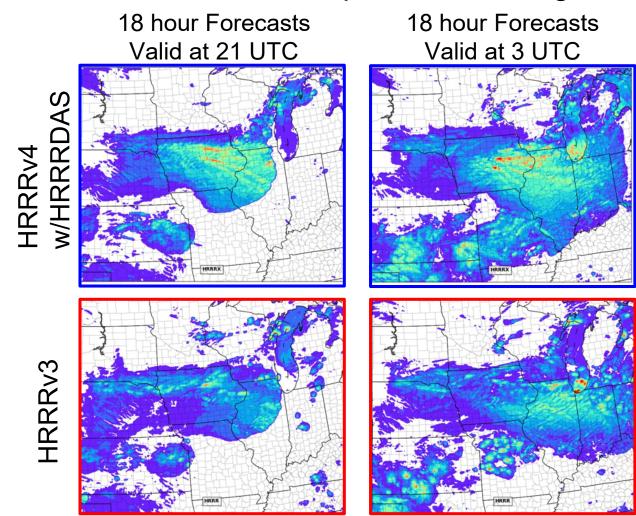


- Challenging forecast of very damaging derecho
- HRRRv4
   consistently shows a
   more organized and
   accurate convective
   system due to
   improved initial
   conditions from
   HRRRDAS.
- HRRRDAS uses hourly assimilation cycles to incorporate both conventional and radar observations and the ensemble mean provides convective scale motions and thermodynamics to HRRRv4.

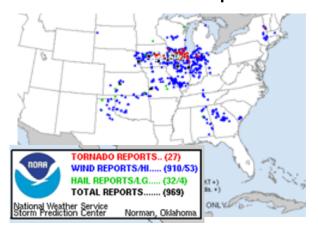
### HRRRDAS Initial Conditions Improves HRRRv4



#### 10m Max Wind Speed 10-11 August 2020



#### **SPC Storm Reports**



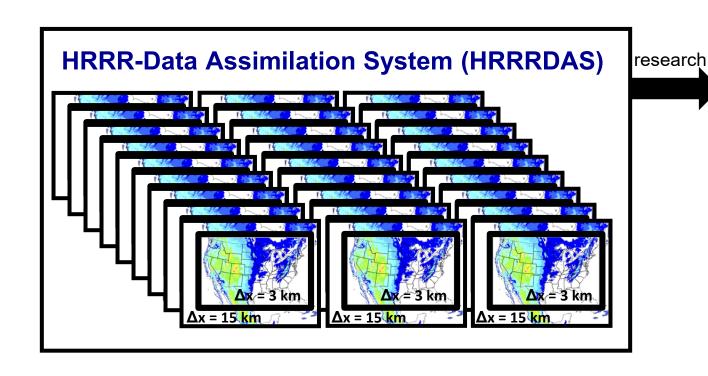
The HRRRv4 wind solutions correctly have:

- stronger overall winds and more coverage further east
- stronger winds starting earlier in the system's life over central lowa
- narrow corridors of very intense winds over eastern lowa

### Ensemble Forecast Challenge: Spread vs Error

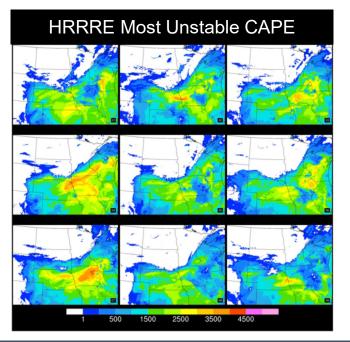


Ensemble forecasts can provide the range of possibilities that might occur and a measure of uncertainty in a forecast.



HRRR-Ensemble (HRRRE)

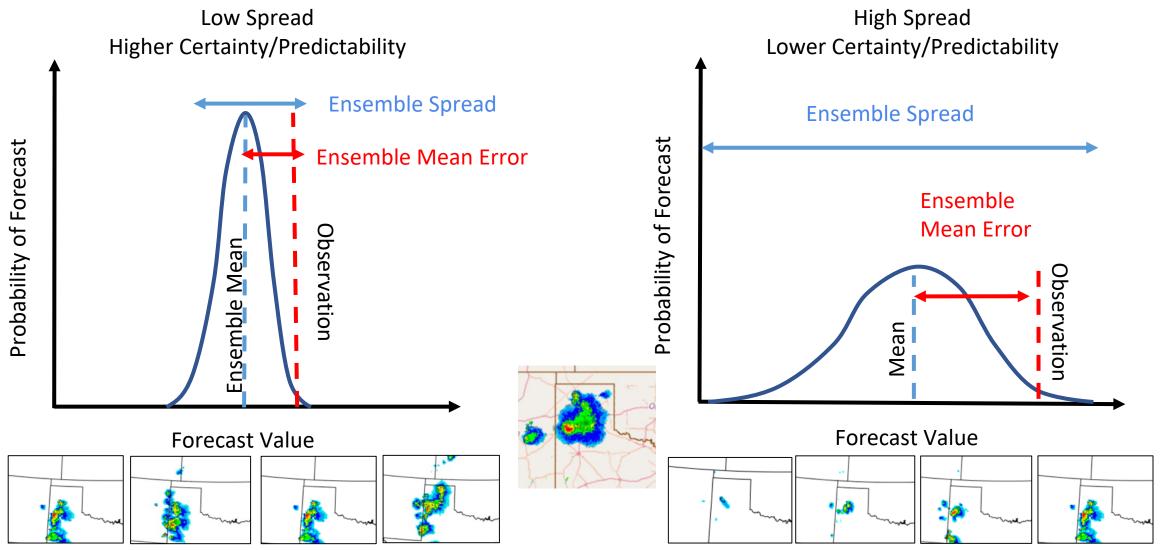
Storm-Scale Ensemble Forecasts



Recommendation C4.2

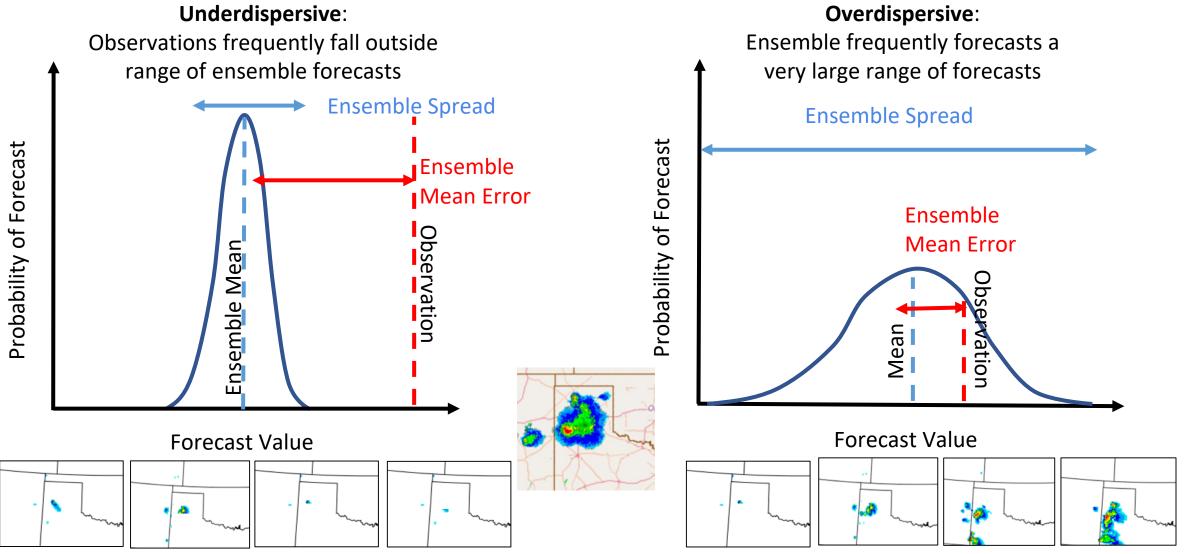
### Ensemble Forecast Challenge: Spread vs Error



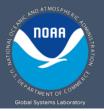


### Ensemble Forecast Challenge: Spread vs Error

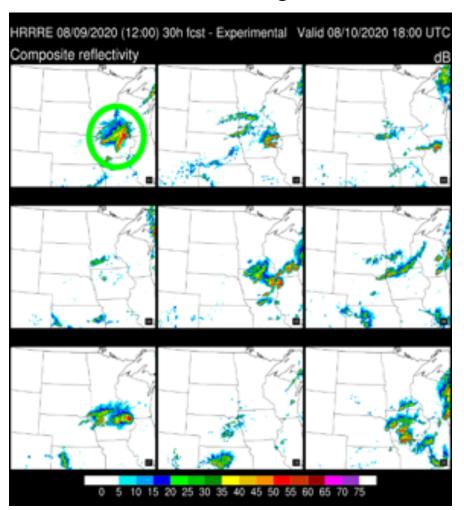




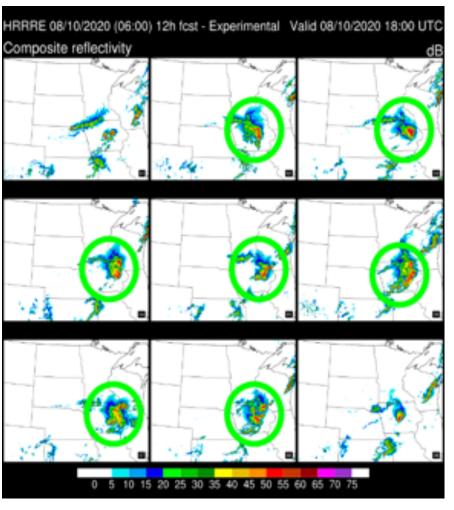
### Ensemble DA & Forecasting ⇒ Better Probabilities



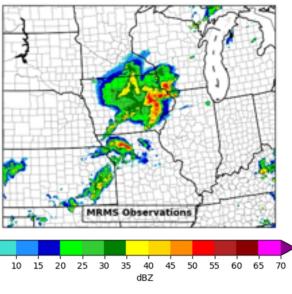
30-h lead-time: get 1 hit



12-h lead-time: 7 of 9 hits



#### Radar observations



- Longer forecasts show low likelihood for a large impact event
- Shorter forecasts decreased spread, increased confidence
- HRRRE includes stochastic physics to create spread

### NOAA Unified Analysis: RTMA-3D Nowcast/Analysis of Record

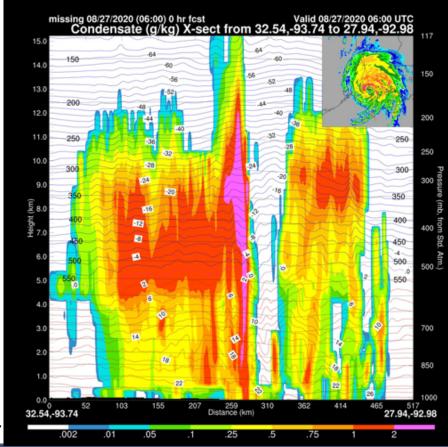


#### GOAL: Record the best estimate of the convective-scale Earth system state.

- Collaborative development with EMC (JTTI & UFS R2O)
- RTMA-3D is not constrained by the need to initialize a forecast (close fit to obs; model behavior not considered)
- RTMA-3D provides 15 min gridded analysis of 3D atmospheric fields with application in multiple areas:
  - General analysis and reanalysis applications
  - Severe weather; targeting replacement of Mesoanalysis
     System (SPC)
  - Aviation parameters; targeting 3D cloud coverage/ceiling grids (AWC)
  - Hydrodynamical modeling input and water in all forms snow cover, soil moisture, lake forecasts, heavy rainfall
  - Model verification 'truth' dataset
- Operational implementation planned for 2024

Recommendation C4.7

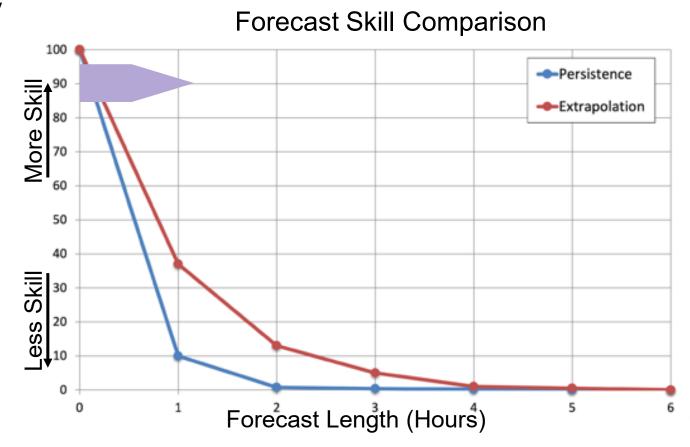
Experimental RTMA-3D during Hurricane Laura 27 Aug 2020





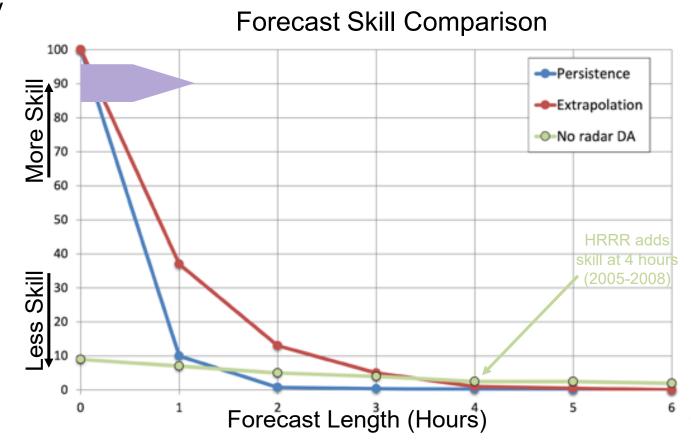
#### GSL is a World Leader in Rapidly Updating DA

 The RTMA-3D analysis system fits obs very closely and is valuable for nowcasting.



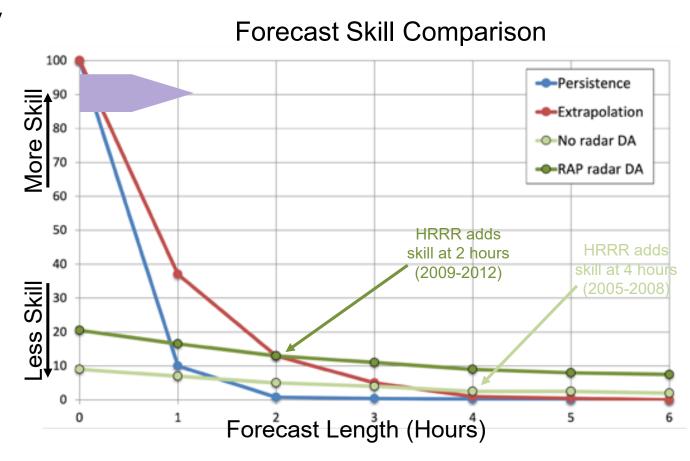


- The RTMA-3D analysis system fits obs very closely and is valuable for nowcasting.
- Convective-scale forecasts lead to useful predictions.



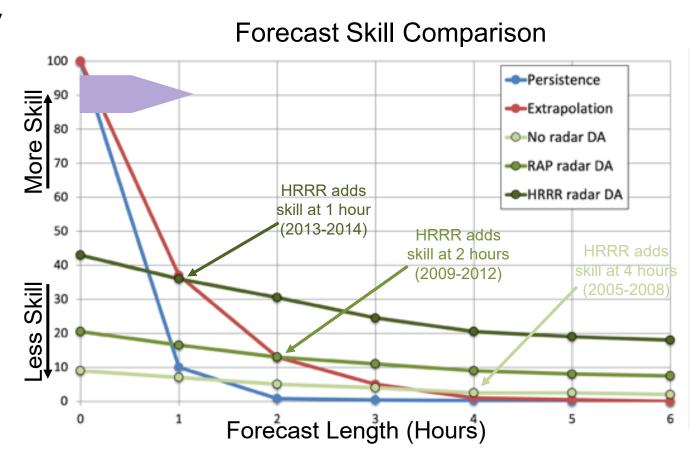


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- Convective-scale forecasts lead to useful predictions.
- Improved mesoscale initial conditions push forecast value to shorter time scales.



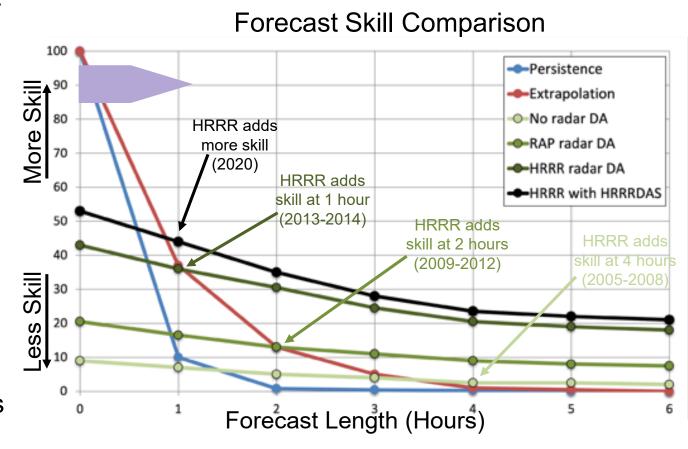


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- Improved mesoscale initial conditions push forecast value to shorter time scales.
- Latent-heating radar assimilation push forecast value to nowcasting time scales.





- The RTMA-3D analysis system fits obs very closely and is valuable for nowcasting.
- Convective-scale forecasts lead to useful predictions.
- Improved mesoscale initial conditions push forecast value to shorter time scales.
- Latent-heating radar assimilation push forecast value to nowcasting time scales.
- Convective-scale ensemble with direct reflectivity assimilation provides initial conditions that increase forecast skill across lead-times.



### NOAA Global Systems Laboratory







### Making Forecasts Better with Data Assimilation Advances



#### Development of analysis systems has advanced forecast skill

- Built on the success of hybrid variational/ensemble approaches and specification techniques
- Transitioning towards use of only advanced techniques for increased observation information retention
- Expanding ensemble assimilation and forecast development for future success

#### Extensive experience with conventional, targeted and novel observations

- Applying observations to model data assimilation and verification applications
- Documenting relative impacts on predictive skill
- Using various sensitivity techniques (OSEs, FSOI)

#### World Leader in R&D of high-resolution rapidly updating models

 Working with many collaborators to improve skill, accumulating innovations for many stakeholders, and finishing with operational transitions

# Thank you!



